

WenQuanYi Micro Hei [Scale=0.9]WenQuanYi Micro Hei Mono song-
WenQuanYi Micro Hei sfWenQuanYi Micro Hei "zh" = 0pt plus 1pt

FleetX
åŘŚåŸČ 0.1.0.beta

PaddlePaddle

2020 åŹŤ 11 æIJĹ 24 æŮě

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- æñcéŁŒĆłǼŒæşléčđæąłŁŁĖăŸĈăijRèő■żĈiijŃæŁŚăznăŸŃæIJZèĈ;ăŸőăŁŁ'æfRăŸĂăŸłĉŤłæŁŭètřăŸŁ

CHAPTER 1

æȚʼäǰŞăžŇčž■äÿŎǻẸǼǻőžæęĆèğŁ

1.1 æŋcēŒŌăĖşæşlād'gègĎælaæuśăżęăęăzăæŁĂæljŕ

ēġſā■āāzt' æīēiŋŋāuſāžčā■ēāžāæĹÄæĲrāy■æŨ■āĹuæŨrēğEğĹ'āÄæĞĲDŭér■ēĹÄāÄAæér■ēſſāÄAæĤ

1.2 éċđæǻłǻŁĚǻŷĈǻijŘěő■çŻĈæŘŘăĴŻçŻĎæǻŷǻŁĚĈǻżŭǻĀij

1. æʒRèĠläžgäyŽāōðeũtçŽDçzRéĤNijŽ
- éćđæaĺçŽDāLEāyČaijRèō■çzČæLĀæIJræžRèĠtçŽĶāžęçŽDäyŽāLāāōðeũtiijNæYřczRèĤGèũĒād' gëğD
 - éćđæaĺāLEāyČaijRèō■çzČçzRèĤGāōðeũtæcĀēĤNçŽDāžTçTlécĒāššāNĒæNñēĠtçDūēr■ēĪĀād' DçREijj
2. āōNād' ĠçŽDāžūēāNāĶāijRijŽ
- æTṛæ■ōāžūēāNijŽēŚĹāržāžgäyŽçTṚNæIJAāyÿçTlçŽDæTṛæ■ōāžūēāNāĶāijRijNéćđæaĺēŚĹāržāōðéŽĒ
 - ætAært' çžĶāžūēāNijŽēĪcāRŚāijČæđDçañāžūijNætAært' çžĶāžūēāNēČĶād' šārĒāĶāādNēōaçōŪēČĪāLEā
 - æĶāādNāžūēāNijŽāržāžŌēũĒād' gëğDæĶāāLEçsžēŪōécYriijNéćđæaĺæRŘäĶZēōaçōŪäyŌā■YāČĪāRĶNæŪ
3. ēĪcāRŚāžŚçnrāIJzæŽřçŽDāžūēāNēō■çzČçzDāžūijŽ
- éćđæaĺēŚĹāržēŽĒççĶd' çĶŚçzIJçŌřácČāĀAçañāžūēōĶād' ĠærTēĶČäĶŌēĒ■çŽDāIJzæŽræRŘäĶZād' Žçğ■ā
 - ēŚĹāržāžŚçnrçōŪāĶZāĒūæIJL' āijzæĀğçŽDçL' zçČžijNéćđæaĺāžšāğNçZĶāIJāŌócť' cāijzæĀğæũsāžęā

1.3 áijĀāğNä;ăçŽĐăĹĚăŷČăijRěő■çžČăžNăĽĚ

- æTt'ä;ŠăĚăőžiižæĹSăžnăŌĹē■ŘæĆĭçŽt' æŌěæăžæ■őăŷzéăŷiižNăĽ'çĚğçnăĚĹCéăžăžRéĂŔăŷĹăŷĹRě
- FAQiižŽăŕăžăžŌénŸécŠăĜžçŎŕçŽĐéŮőécŸiižNăĹSăžnăiižŽăőŽăĹJšæTt' çŘĚçŽŷăĚšăĚăőžăĹŔFAQ
- âĤnéĂšăŷĹăĹNiižŽăĉCăđĹJăČšăĹĹĂă;ŎăĹŔăĹJnçŽĐăžĚĚğčĉđăăĭçŽĐăĹĚăŷČăijRěő■çžČiižNăĹS
- GPUăđ'ŽăĹJžĚő■çžČiižŽăĉCăđĹJăĆĭăŷšçžŔăiižĀāğNă;ĤçŤĭGPUĚĤŽăăNăđ'ŽăĹJžăđ'Žă■ăĚő■çžČiižNăĹS
- âŔCăĤŕăĹJ■ăĹăăŽĭiižŽăĤăăĤăĤăĤ' căĂăăŌĹē■ŘçšžçžšĉĉĚăššăŷŷçŤĭçŽĐăžŷĚăNĚő■çžČăŮžăiižŔiiž
- âĚnăĹJĹ'ăžŠçŎŕăĉČăőđĚŮŷiižŽăĉCăđĹJăĆĭăĹĹăĚnăĹJĹ'ăžŠăŷĹĚŮŠĚĜĭăŷšçŽĐGPUăđ'Žă■ăžžăĹăĹiižNă
- âijžăĂğĚő■çžČiižŽăĉCăđĹJăŕăžăĉCă;ŤăĹĹ'çŤĭăžŠçŕăiižăĂğĚĹDăžŔĚĤŽăăNăđ'ğĚğĐăĹăĚšŷĉĚRěő■çžČăĹJ■ăĹăăđNăiižăĂğĚšŷĉĚŔ

1.4 RoadMap

- æĹSăžnăžšăiižŽăŌĹēĂăăđ'ğĚğĐăĹăăŷăžăă■ăžăăĹĂăĹŔĉĉĚăššăĹĹĂăĹ■ăĤçŽĐăĹĂăĹŔăĹŔĚĤŽĚĉ
 - ĚĤSăĹJšiižŽă■ČăžĚĚğĐăĹăăĹăăđNăŔCăĤŕçŽĐGPUăđ'ŽăĹJžăđ'Žă■ăĚő■çžČiižNăĹŤnĚŮăĹJšăĹĚ

CHAPTER 2

Installing Paddle and FleetX

2.1 Paddle

Paddle is a deep learning framework that can be installed on a GPU or CPU. This section shows how to install Paddle on a GPU.

- To install Paddle on a GPU, run the following command:
- To install Paddle on a CPU, run the following command:
- To install Paddle on a GPU with CUDA 11.0, run the following command:

```
pip install paddlepaddle-gpu
```

After installation, you can verify the installation by running the following command:

2.2 FleetX

FleetX is a distributed training framework that can be installed on a GPU or CPU. This section shows how to install FleetX on a GPU.

- To install FleetX on a GPU, run the following command:
- To install FleetX on a CPU, run the following command:
- To install FleetX on a GPU with CUDA 11.0, run the following command:

```
pip install fleet-x==0.0.7
```


æŁŰèĀĚä;ŁçŤlæŁŚäžňäÿžçŤlæŁuæŘŘă;ŽăžĚăũšçzŔçijŮèŕŚăě;çŽĎăŏL'èĉĚăŇĚïjŇăŔřăžěäÿŇè;ĵăĹŕă

```
# python2
wget --no-check-certificate https://fleet.bj.bcebos.com/fleet_x-0.0.
  ↪7-py2-none-any.whl
pip install fleet_x-0.0.4-py2-none-any.whl
# python3
wget --no-check-certificate https://fleet.bj.bcebos.com/fleet_x-0.0.
  ↪7-py3-none-any.whl
pip3 install fleet_x-0.0.4-py3-none-any.whl
```

CHAPTER 3

éiŽæĀAāŽ;āLĒāyČāijRèő■çzČāĒnéĀšāijĀāgŃ

årzāžŌād'gēČlāLĒçTlāLūæIēēōšijNæTṛæ■ōāzūēāNèő■çzČāšžæIJñāRfāzēēgčāEšāōđéŽĒāyŽāLāāy■çŽ
paddle.distributed.fleetēfZēāNæTṛæ■ōāzūēāNèő■çzČāĀČāIJlāTṛæ■ōāzūēāNæŪzāijRāyNīijNéA
TrainingīijL'āŠNāRCæTṛæIJ■āLāāŽlèő■çzČīijLParameter Server Train-
ingīijLīijNæŌēāyNæIēçŽDä;Nā■RāijŽāzēāRŃæāüçŽDālāādNæIēērt'æYŌāyđ'çg■æđūæđDçŽDæTṛæ■ōāzūē

3.1 çL'ĒæIJñèēAæśĆ

- paddlepaddle-2.0.0-rc-cpu / paddlepaddle-2.0.0-rc-gpuāRŁāžēāyŁ

3.2 ælāqāđNæRRèĚř

äyžāžĒæŪzä;fèrt'æYŌīijNæLŠāzñéGĠçTlāyđ'āsČāĒlēđæŌēç;ŠçzIJçŽDāLĒçšžælāqāđNīijNāzūā;ŁçTlā
py

```
import paddle
import paddle.static.nn as nn

paddle.enable_static()
def mnist_on_mlp_model():
    train_dataset = paddle.vision.datasets.MNIST(mode='train')
    test_dataset = paddle.vision.datasets.MNIST(mode='test')
    x = paddle.data(name="x", shape=[64, 1, 28, 28], dtype='float32
↪')
    y = paddle.data(name="y", shape=[64, 1], dtype='int64')
    x_flatten = paddle.reshape(x, [64, 784])
```

(äyNéatçzğçz■)

(çz■äÿŁéął)

```
fc_1 = nn.fc(input=x_flatten, size=128, act='tanh')
fc_2 = nn.fc(input=fc_1, size=128, act='tanh')
prediction = nn.fc(input=[fc_2], size=10, act='softmax')
cost = paddle.fluid.layers.cross_entropy(input=prediction,
↪label=y)
acc_top1 = paddle.fluid.layers.accuracy(input=prediction,
↪label=y, k=1)
avg_cost = paddle.fluid.layers.mean(x=cost)
return train_dataset, test_dataset, x, y, avg_cost, acc_top1
```

3.3 éĜĜçŦĠPUåd'ŽæĲžåd'Žā■àè£ŽèąŃăŔŅæ■ěèő■çžČ

collective_trainer.py

```
import os
import paddle
import paddle.distributed.fleet as fleet
from model import mnist_on_mlp_model

train_data, test_data, x, y, cost, acc = mnist_on_mlp_model()
place = paddle.CUDAPlace(int(os.environ.get('FLAGS_selected_gpus', 0)))

train_dataloader = paddle.io.DataLoader(
    train_data, feed_list=[x, y], drop_last=True,
    places=place, batch_size=64, shuffle=True)
fleet.init(is_collective=True)
strategy = fleet.DistributedStrategy()
#optimizer = paddle.optimizer.Adam(learning_rate=0.01)
optimizer = paddle.fluid.optimizer.Adam(learning_rate=0.001)
optimizer = fleet.distributed_optimizer(optimizer, strategy=strategy)
optimizer.minimize(cost)

exe = paddle.static.Executor(place)
exe.run(paddle.static.default_startup_program())

epoch = 10
step = 0
for i in range(epoch):
    for data in train_dataloader():
        step += 1
        loss_val, acc_val = exe.run(
            paddle.static.default_main_program(),
            feed=data, fetch_list=[cost.name, acc.name])
```

- å■TæIJžāZžā■ąèő■çzČāŘřāŁāŚ;äzd'

parameter_server_trainer.py

```
import paddle
import paddle.distributed.fleet as fleet
from model import mnist_on_mlp_model

paddle.enable_static()

train_data, test_data, x, y, cost, acc = mnist_on_mlp_model()

fleet.init()
strategy = fleet.DistributedStrategy()
strategy.a_sync = True
optimizer = paddle.fluid.optimizer.Adam(learning_rate=0.001)
optimizer = fleet.distributed_optimizer(optimizer, strategy)
optimizer.minimize(cost)

if fleet.is_server():
    fleet.init_server()
    fleet.run_server()
else:
    place = paddle.CPUPlace()
    exe = paddle.static.Executor(place)
    exe.run(paddle.static.default_startup_program())
    fleet.init_worker()

train_dataloader = paddle.io.DataLoader(
    train_data, feed_list=[x, y], drop_last=True, places=place,
    batch_size=64, shuffle=True)

epoch = 1
for i in range(epoch):
    for data in train_dataloader():
        cost_val, acc_val = exe.run(
            paddle.static.default_main_program(),
            feed=data, fetch_list=[cost.name, acc.name])
        print("loss: {}, acc: {}".format(cost_val, acc_val))
    fleet.stop_worker()
```

```
fleetrn --worker num 2 --server num 2 parameter server trainer.py
```

CHAPTER 4

ǎŁíæĀȦǎŽǃǎŁĘǎŸČǎijRèő■çžČǎ£néĀşǎijĀǎğŊ

PaddleöYæŨzæŨGæqäy■ārZāLlæĀAāZĭiijLāŚ;āzd'āijRçijŨćlŊiijL'āAžZāEærTē;ČèrēçzEçŽDžzŊçz
paddle.distributed.fleetæŖēāRčāzŖPaddle 2.0-RCĽ'LēIJñāijĀāğNæŤræNĀāLlæĀAāZĭāLēāyČ
paddle.distributed.fleetæŖēāRčēfZēaŊNāLlæĀAāZĭāLēāyČāijRēō■çČāĀCæŖēäyNæIēāLŚāžñ

æʃliijZçZôâL■paddle.distributed.fleetâRřfâĹâĹĹāĂĀăZĭ;âĹEăȳČaijŘěô■čžCăzĚăŤrăŇAé.
CommunicationiijL'əlaaijRiiŋNäy■ăŤrăŇAâRĆăŢrăIJ■ăŁaăŻĲiijLP
Parameter-ServeriijL'əlaaijRâĂĆăIJnăŮĞçd'žă;NäyžéZEârĹĹéĂŽăŁăəlaaijRăzzăŁăăĂĆ

4.1 çL'ŁæIĴñèęAæśĆ

- paddlepaddle 2.0-rc-gpu

4.2 å■Tæljžå■Tå■æèó■çZČ

äyÑéÍcæYräyÄäyléÍdäyÿcôÄä■TçŽĐäLÍæÄÄäZ;ä■TæIJžä■Tä■açÍÑäžRäÄĆç;ŚçzIJäRÍæIJL'äRÍæIJL'2

```
# -*- coding: UTF-8 -*-
import paddle
import paddle.nn as nn

# åŒžăžŁ'âĖĹèĹđăŒŒěç;ŚçżIJiijŇÉIJĂçžğæŁ'ĹèĞĹnn.Layer
class LinearNet(nn.Layer):
    def __init__(self):
        super(LinearNet, self).__init__()
        self._linear1 = nn.Linear(10, 10)
```

(äyÑeåtçzğçz■)

(çzäyLéat)

```

self._linear2 = nn.Linear(10, 1)

def forward(self, x):
    return self._linear2(self._linear1(x))

# 1. äijÄäRäLíæÄÄäZçæÍääijR
paddle.disable_static()

# 2. äöžázL'ç;ŠçzIJärzéšäiijNä■Säd'säG;æTřäŠNäijYäNŮäZÍ
layer = LinearNet()
loss_fn = nn.MSELoss()
adam = paddle.optimizer.Adam(
    learning_rate=0.001, parameters=layer.parameters())

for step in range(20):
    # 3. æL'ğèäNäL'■äRŠç;ŠçzIJ
    inputs = paddle.randn([10, 10], 'float32')
    outputs = layer(inputs)
    labels = paddle.randn([10, 1], 'float32')
    loss = loss_fn(outputs, labels)

    print("step:{}\tloss:{}".format(step, loss.numpy()))

    # 4. æL'ğèäNäR■äRŠèóäçôŮäŠNäRĆæTřæŽt'æŮř
    loss.backward()
    adam.step()
    adam.clear_grad()

```

ärEäzëäyLäzčçäAäfIä■Yäyžtrain_single.pyüijNëfRëaÑpython
train_single.pyüijNäCíärEçIJNäLřæYçd'žæCäyNæŮëäŮäfææAřüijŽ

```

step:0   loss:[1.2709768]
step:1   loss:[0.7705929]
step:2   loss:[2.2044802]
step:3   loss:[1.6021421]
step:4   loss:[2.0286825]
step:5   loss:[0.7866151]
step:6   loss:[1.926115]
step:7   loss:[0.3647427]
...

```

4.3 ä■TæIJžäd'Žä■æèö■çzČ

ä;fçTíFleetæŮëärčèfZëäNäLíæÄÄäZçäLEäyČäijRèö■çzČäEüäódeIdäyçŽDçóÄä■TüijNäRléIJÄäfóæ

1. äřijäEëpaddle.distributed.fleetäNĚ

```
from paddle.distributed import fleet
```

2. áĹĭăġŇăŇŮfleetçŎŕăčČ

```
fleet.init(is_collective=True)
```

3. éĀŽèĹĢfleetèŎŭăŔŮăĹĚăŸČăĭjŔăĭjŸăŇŮăŽĹăŠŇăĹĚăŸČăĭjŔăĭjăđŇ

```
strategy = fleet.DistributedStrategy()
adam = fleet.distributed_optimizer(adam, strategy=strategy)
dp_layer = fleet.distributed_model(layer)
```

èŕŕæŸŎĭjŽçŽŏăĹ■éĹŽæĀĀăŽçDistributedStrategyăŸŇçŽĎăĹĚăŸČăĭjŔç■ŮçŦæ■céĀŔæ■ěăĹ
æăžæ■ŏăĹŖăžŇæĪĀăĭjĀăġŇæŔŔăçŽçŽĎă■ŦæĪJžă■Ŧă■ăăžççăĀçđ'žăçŇĭjŇăĒ■æăžæ■ŏăŖççèŕĀč

```
# -*- coding: UTF-8 -*-
import paddle
import paddle.nn as nn
#ăĹĚăŸČăĭjŔstep 1: âŕĭjăĚĕpaddle.distributed.fleetăŇĚ
from paddle.distributed import fleet

# âŏŽăžĹ'ăĚĹèĹđăŎĚç;ŖçžĪJĭĭjŇéĪJĀçžġăĹ'ĹèĢĭnn.Layer
class LinearNet(nn.Layer):
    def __init__(self):
        super(LinearNet, self).__init__()
        self._linear1 = nn.Linear(10, 10)
        self._linear2 = nn.Linear(10, 1)

    def forward(self, x):
        return self._linear2(self._linear1(x))

# 1.ăĭjĀăŔŕăĹăĹăăŽçăĹăăĭjŔ
paddle.disable_static()

# âĹĚăŸČăĭjŔstep 2: áĹĭăġŇăŇŮfleet
fleet.init(is_collective=True)

# 2. âŏŽăžĹ'ç;ŖçžĪJăŕžèŖăĭjŇæ■Ŗăđ'ŖăĢç;æŦŕăŖŇăĭjŸăŇŮăŽĹ
layer = LinearNet()
loss_fn = nn.MSELoss()
adam = paddle.optimizer.Adam(
    learning_rate=0.001, parameters=layer.parameters())

# âĹĚăŸČăĭjŔstep 3:
→éĀŽèĹĢfleetèŎŭăŔŮăĹĚăŸČăĭjŔăĭjŸăŇŮăŽĹăŠŇăĹĚăŸČăĭjŔăĭjăđŇ
strategy = fleet.DistributedStrategy()
adam = fleet.distributed_optimizer(adam, strategy=strategy)
dp_layer = fleet.distributed_model(layer)
```

(ăŸŇéăŦçžġç■)

(çzäyLéat)

```

for step in range(20):
    # 3. æL'ğèaÑâL'■âŘŚç;ŚçzIJ
    inputs = paddle.randn([10, 10], 'float32')
    outputs = dp_layer(inputs)
    labels = paddle.randn([10, 1], 'float32')
    loss = loss_fn(outputs, labels)

    print("step:{}\tloss:{}".format(step, loss.numpy()))

    # 4. æL'ğèaÑâŘ■âŘŚèóaçÕŮâŠÑâŘĆæŤræŽt'æŮř
    loss.backward()

    adam.step()
    adam.clear_grad()

```

ârEäzèäyLäzčçäAäfiä■YäyŽtrain_fleet.pyiijNâAĞèöçèçAèfRèaÑ2â■açŽDäzzâLajijNéCczLâR

```
fleetrn --gpus=0,1 dygraph_fleet.py
```

æCíârEçIJNâLræYçd'žæCäyNæŮèâfŮäfæAfiijŽ

```

----- Configuration Arguments -----
gpus: 0,1
ips: 127.0.0.1
log_dir: log
server_num: None
servers:
training_script: dygraph_fleet.py
training_script_args: []
worker_num: None
workers:
-----

INFO 2020-0X-XX 08:33:30,247 launch.py:441] Run collective gpu mode.
→ gpu arguments:['--gpus'], cuda count:8
INFO 2020-0X-XX 08:33:30,247 launch_utils.py:430] Local start 2_
→ processes. First process distributed environment info (Only For_
→ Debug) :

→ +=====
|                                     Distributed Envs                                     |
→ Value                               |                                                                 |
+-----+
→ -----+
|                                     PADDLE_CURRENT_ENDPOINT                                     | 127.0.
→ 0.1:59664                           |                                                                 |
|                                     PADDLE_TRAINERS_NUM                                           |
→ 2                                     |                                                                 |

```

(äyNéatçzğçz■)

(çzäyLéat)

```

|                                FLAGS_selected_gpus                                |
↪0                                |
|                                PADDLE_TRAINER_ENDPOINTS                        127.0.0.
↪1:59664,127.0.0.1:48993        |
|                                PADDLE_TRAINER_ID                                |
↪0                                |
|
↪+=====
step:0  loss:[1.3279431]
step:1  loss:[2.5023699]
step:2  loss:[3.3197324]
step:3  loss:[2.6869867]
step:4  loss:[2.6306524]
step:5  loss:[1.9267073]
step:6  loss:[1.2037501]
step:7  loss:[1.1434236]
...

```

ãŃæŦ' 2ãacŽDæŮeafŮafæAřazšãRřãIJĹ. /log/çŽŃajŦäyNæšçIJNãĀĆazEğğcæŽt'ãd'Žfleetrun
ãRřãĹãĹEäyČãijRäzzãŁaãĀĆ

4.4 4d'ŽæIJžãd'ŽãæèõçžČ

äzŌãŦæIJžãd'ŽããĹřãd'ŽæIJžãd'ŽãæèõçžČijNãIJläžčçãAäyŁãžũäyēIJĀèçAãAŽãzzã;ŦæŦžãĹij

```
fleetrun --ips="xx.xx.xx.xx,yy.yy.yy.yy" --gpus=0,1 dygraph_fleet.py
```

ãIJ2ãRřæIJžãŽläyŁãĹEãĹnëfRëãNäžëäyŁãRřãĹãš;äzd'ijNfleetrunãRĹãIJläRŌãRřãĹEãĹnãRřãĹ
æĆĹãRĹãIJlipäyžxx.xx.xx.xxçŽDæIJžãŽläyŁçIJNãĹřãš;äzd'ãRřè;ŠãĠžæŮeafŮafæAřijŽ

```

----- Configuration Arguments -----
gpus:  None
ips:  xx.xx.xx.xx,yy.yy.yy.yy
log_dir:  log
server_num:  None
servers:
training_script:  dygraph_fleet.py
training_script_args:  []
worker_num:  None
workers:
-----
INFO 2020-0X-XX 21:29:41,918 launch.py:434] Run collective gpu mode.
↪ gpu arguments:['--ips'], cuda count:2
INFO 2020-0X-XX 21:29:41,919 launch_utils.py:426] Local start 2
↪ processes. First process distributed environment info (Only For
↪ Debug) :

```

(äyNëatçžçz)

ä|£çŦífleetrunăŘřăĺăĹĖăÿČăijŘăzzăłă

PaddleæRŘä;ZäŚ;äzd'ëaÑäRřäŁlāŚ;äzd'fleetrunĭijÑëĚ■āRŁPaddleçŽDāLEāyČāijRénYçžgAPIpac
distributed.fleetā■śāRřë;æİ;āRřāŁlPaddleēZEāRŁēĀŽāfæāİāijRæLŪāRĆāTřæIJ■āŁāZlāİāijf
fleetrunāIJléİZēĀāZ;āŚNāŁlāēĀāZ;āIJzēŽřāyNāİGāRřā;ŁçTlāĀĆ

5.1 aEĖaŏžarijeL

1. ä;£çŦłēAæšĆ
2. ä;£çŦłērt'æŸŎ
 - 2.1. éŽEāŘĹéĀŽä£qèő■çžČ
 - 2.2. āŘĆæŦræIJ■āŁqāŽĹèő■çžČ
3. fleetrunăŜ;ăžď'āŘĆæŦrăžŇč■
4. ä;£çŦłfleetrunē£ŽeqŇGPUăď'Žă■qèő■çžČăőďü;Ň

5.2 ä;£çŦíèèAæśĆ

ä;£çŦífleetrunåŦ;äzd'çŽĐèęAæśĆ:

- paddlepaddle 2.0-rc

5.3 ä;£çŦlèrt'æŸŦ

fleetrunajŁçTlāIJzæŽřäyzeèAāĹEäyžēŽEāŘĹéĂŽăfăèő■czČijĹCollective
TrainingüijĹāŠŇāŔĈæŦræIJ■āŁāāŽĹēő■czČijĹParameter Server TrainingüijĹāĂĈ

éŽĚāŔĹéĀŽāŋāēō■czČäyĀēĹŋāIJGPUēō;āđ'ĜäyĹēŔĚāŇijŇāŽāē■d'ēĹSāznāŔĚāžŇcz■GPUā■TāIJā■Tā
āŔCāŤŕāIJ■āLāāZĹēō■czČāŇĚāŔŋāēIJ■āLāēĹCçCzāĀāēō■czČēĹCçCzāēāŔĹāijCāđDēō■czČēĹCçCzçŽĎ
āŽāē■d'ēĹSāznāŔĚāžŇcz■āIJCPUéŽĚç;đ'āĀGPUéŽĚç;đ'äyĹāŠŇāijCāđDēŽĚç;đ'äyĹāēCā;Tā;ŋçTĹfle
āĀCfleetruneŤŕāŇāāIJçŽ;āžēāĒŋāŔyāĒēČlāžSPaddleCloudäyĹēŔĚāŇāĹēāyČāijŔāžžāLāijŇāēŌē

ā;āāžšāŔŕāžēā;ŋçTĹ python -m paddle.distributed.launch
āēāŔŕāĹēō■czČāžžāLāijŇāžŇāōđäyĹijŇ fleetruneŤŕāĹēĀĒçŽĎāŋāē■ūāŮžāijŔāĀČ

5.3.1 éŽĚāŔĹéĀŽāŋāēō■czČ

- GPUā■TāIJā■Tāāēō■czČ

ā■TāIJā■TāāēIJĹäyđ'çģ■āŮžāijŔijŽäyĀçģ■āŔŕçŽt'ēŌēā;ŋçTĹpythonāĹġēāŇijŇāžšāŔŕāžēā;ŋç
āĀŔāŮžāŋçTāyĀāŠçŽt'ēŌēā;ŋçTĹpythonāĹġēāŇ

```
export CUDA_VISIBLE_DEVICES=0
python train.py
```

āĀŔāŮžāŋçTāžŇāĀŠā;ŋçTĹfleetruneĹġēāŇ

```
fleetrune --gpus=0 train.py
```

āšĹijŽāçCāđIJāŇĜāōŽāžĒexport CUDA_VISIBLE_DEVICES=0
ijŇāĹŽāŔŕāžēçŽt'ēŌēā;ŋçTĹijŽ

```
export CUDA_VISIBLE_DEVICES=0
fleetrune train.py
```

- GPUā■TāIJāđ'Žāāēō■czČ

ēŇēāŔŕāĹā■TāIJžāāçŽĎāžžāLāijŇāŔĹēIJāēĀŽēŋç--gpusāŇĜāōŽçĹžēŮšçŽĎ4āijāā■ā■šāŔŕāĀ

```
fleetrune --gpus=0,1,2,3 train.py
```

āšĹijŽāçCāđIJāŇĜāōŽāžĒexport CUDA_VISIBLE_DEVICES=0,1,2,
3ijŇāĹŽāŔŕāžēçŽt'ēŌēā;ŋçTĹijŽ

```
export CUDA_VISIBLE_DEVICES=0,1,2,3
fleetrune train.py
```

- GPUāđ'ŽāIJāđ'Žāāēō■czČ

[çđ'žā;ŇäyĀ] 2āIJž8ā■ā (āŔŕāyĹēĹCçCzāā■ā)

```
fleetrune --ips="xx.xx.xx.xx,yy.yy.yy.yy" --gpus=0,1,2,3 train.py
```

āšĹijŽāçCāđIJāŔŕāŔŕāIJāŽĹāĪĜāŇĜāōŽāžĒexport CUDA_VISIBLE_DEVICES=0,
1,2,3ijŇāĹŽāŔŕāžēçŽt'ēŌēāIJāŔŕāŔŕēĹCçCzāyĹāŔŕāĹijŽ

```
export CUDA_VISIBLE_DEVICES=0,1,2,3
fleetrune --ips="xx.xx.xx.xx,yy.yy.yy.yy" train.py
```

[čd'žä;NäžŇ] 2æIJž16ā■āijŁæfRāyŁēŁĆčĆž8ā■āijŇāAĞēō;æfRāRřæIJžāZÍlāĜæIJL'8ā■āRřä;ččŤlīij

```
fleetrn --ips="xx.xx.xx.xx,yy.yy.yy.yy" train.py
```

5.3.2 āRČæŤræIJ■āŁāāZlēō■čžČ

āIJCPUēZEč;čd'èŁRēāŇāRČæŤræIJ■āŁāāZÍ

- āRČæŤræIJ■āŁāāZlēō■čžČ - ā■ŤæIJžæłæŇšāŁEāyČāijRēō■čžČ

1āRřæIJžāZÍēĀŽēŁĜād'ŽēŁZčÍŇāłæŇšāŁEāyČāijRēō■čžČīijŇ1āyŁæIJ■āŁāēŁĆčĆžæŘēĒ4āyŁēō■čžČ

fleetrnāRřāŁīæŪūāRlēIJāŇĜāōŽæIJ■āŁāēŁĆčĆžæŤr--server_numāŠŇēō■čžČēŁĆčĆžæŤr-

```
fleetrn --server_num=1 --worker_num=4 train.py
```

- āRČæŤræIJ■āŁāāZlēō■čžČ - èĜlāōŽāzŁ'ād'ŽæIJžēō■čžČ

fleetrnāRřāŁīæŪūāRlēIJāŇĜāōŽæIJ■āŁāēŁĆčĆžčŽDipāŠŇčŇrāRčāŁŪēāł--servers
āŠŇēō■čžČēŁĆčĆžčŽDipāŁŪēāł--workers īijŇā■šāRřēŁZēāŇād'ŽæIJžēō■čžČāĀČ
āyŇāŁŪčd'žä;Näy■īijŇxx.xx.xx.xxāžčēāłæIJžāZÍlīijŇyy.yy.yy.yyāžčēāłæIJžāZÍlīijŇ6170āžčēāłčŤlāŁūæŇČ

```
# 2āyŁservers 8āyŁworkers
fleetrn --servers="xx.xx.xx.xx:6170,yy.yy.yy.yy:6171" --workers=
→ "xx.xx.xx.xx,xx.xx.xx.xx,xx.xx.xx.xx,xx.xx.xx.xx,yy.yy.yy.yy,yy.
→ yy.yy.yy,yy.yy.yy.yy,yy.yy.yy.yy" train.py
```

--workersāRČæŤrāRřāzēāžĒæŇĜāōŽipāŁŪēāłīijŇæ■d'æŪūfleetrnāRřāĪijZāIJlāRřāŁlēō■čžČāž

```
# 2āyŁservers 8āyŁworkers
fleetrn --servers="xx.xx.xx.xx:6170,yy.yy.yy.yy:6171" --workers=
→ "xx.xx.xx.xx:6172,xx.xx.xx.xx:6173,xx.xx.xx.xx:6174,xx.xx.xx.
→ xx:6175,yy.yy.yy.yy:6176,yy.yy.yy.yy:6177,yy.yy.yy.yy:6178,yy.yy.
→ yy.yy:6179" train.py
```

āIJGPUēZEč;čd'èŁRēāŇāRČæŤræIJ■āŁāāZÍ

- āRČæŤræIJ■āŁāāZlēō■čžČ - ā■ŤæIJžæłæŇšāŁEāyČāijRēō■čžČ

1āRřæIJžāZÍēĀŽēŁĜād'ŽēŁZčÍŇāłæŇšīijŇ2āyŁæIJ■āŁāēŁĆčĆžæŘēĒ4āyŁēō■čžČēŁĆčĆžīijŇāfRāy

```
# 2āyŁserver 4āyŁworker
export CUDA_VISIBLE_DEVICES=0,1,2,3
fleetrn --server_num=2 --worker_num=4 train.py
```

1āRřæIJžāZÍēĀŽēŁĜād'ŽēŁZčÍŇāłæŇšīijŇ 2āyŁæIJ■āŁāēŁĆčĆžæŘēĒ2āyŁēō■čžČēŁĆčĆžīijŇāyŁd'ā

```
# 2äŸlserver 2äŸlworker
export CUDA_VISIBLE_DEVICES=0
fleetrun --server_num=2 --worker_num=2 train.py
```

• aRČæTŕæIJ■aŁaŹlëo■czČ - èĠlăoŽăzL'ăd'ŽæIJžèo■czČ

fleetrunăRŕăLlăUŭăRlăIJăæŇĠăoŽăæIJ■aŁaèŁĆćĆzčŽDipăŠŇġŕăRčăLŪeăl--servers
ăŠŇ èo■czČèŁĆćĆzčŽDipăŠŇġŕăRčăLŪeăl--workers iijŇăŇăŕŕèĤZeaŇăd'ŽæIJžèo■czČăĂČ

ăžëăŸŇčđ'žăĴŇăŸ■iijŇxx.xx.xx.xxăžčëălăIJžăŽl1iijŇyy.yy.yy.yyăžčëălăIJžăŽl2iijŇ6170ăžčëălăĴlăLŭă

```
#_
→2ăRŕăIJžăŽl1iijŇăRŕăRŕăIJžăŽlăĠăIJL'1ăŸlăIJ■aŁaèŁĆćĆzčiiijŇ1ăŸlëo■czČèŁĆćĆz
# 2äŸlserver 2äŸlworker
# æŕŕăRŕăIJžăŽlăĠăŇĠăoŽăžEăRŕčŦlëoĴăd'Ġ GPU:0
export CUDA_VISIBLE_DEVICES=0
fleetrun --servers="xx.xx.xx.xx:6170,yy.yy.yy.yy:6171" --workers=
→"xx.xx.xx.xx:6172,yy.yy.yy.yy:6173" train.py
```

ăžëăŸŇčđ'žăĴŇăŸ■iijŇfleetrunăRŕăLlăŇăIJ2ăRŕăIJžăŽlăŸLăRŕăLl1ăŸlăIJ■aŁaèŁĆćĆziiijŇ4ăŸlëo■czČ

```
#_
→2ăRŕăIJžăŽl1iijŇăRŕăRŕăIJžăŽlăĠăIJL'1ăŸlăIJ■aŁaèŁĆćĆzčiiijŇ4ăŸlëo■czČèŁĆćĆz
# 2äŸlserver 4äŸlworker
# æŕŕăRŕăIJžăŽlăĠăŇĠăoŽăžEăRŕčŦlëoĴăd'Ġ GPU:0,1,2,3
export CUDA_VISIBLE_DEVICES=0,1,2,3
fleetrun --servers="xx.xx.xx.xx:6170,yy.yy.yy.yy:6171" --workers=
→"xx.xx.xx.xx:6172,xx.xx.xx.xx:6173,xx.xx.xx.xx:6174,xx.xx.xx.
→xx:6175,yy.yy.yy.yy:6176,yy.yy.yy.yy:6177,yy.yy.yy.yy:6178,yy.yy.
→yy.yy:6179" train.py
```

ăijČæđĐÉZĚčĴđ'èĤŕëăŇăRČæTŕæIJ■aŁaŹl

• aRČæTŕæIJ■aŁaŹlëo■czČ - aŇTæIJžælăæŇŝăLĚăŸČăijŕëo■czČ

1ăRŕăIJžăŽlăĂŽëĠăđ'ŽëĤŽĴlŇălăæŇŝiijŇ2ăŸlăIJ■aŁaèŁĆćĆzăŕëĤ■2ăŸlëo■czČèŁĆćĆzăžëăŕL2ă

```
# 2äŸlserver 4äŸlworker
export CUDA_VISIBLE_DEVICES=0,1
fleetrun --server_num=2 --worker_num=2 --heter_worker_num=2 train.py
```

5.4 fleetrunăŠĵăzd'aRČæTŕăžŇcz■

• CollectiveălăăijŕčŽăăŝăRČæTŕ:

- ips iijŇLŕiijŇăRŕëĂŇiijŇŹ æŇĠăoŽăĂŇăŇl'ăŝăžZëŁĆćĆzIPëĤZeaŇëo■czČiijŇézŸëođ'ăŸžă
ăŇăĵăĵŽăIJlăIJŇăIJŕăL'ġëăŇăŇTæIJžăŇTăăăLŪăđ'Žăăăëo■czČăĂČ

- ## 5.5 ä;£çŦífleetruneè£ŽèaŦGPUåd'Žă■àèő■çžČăődă,Ŧ

```
import os
import time
import paddle
import paddle.distributed.fleet as fleet
import paddle.static.nn as nn
import paddle.fluid as fluid

def mnist_on_mlp_model():
    train_dataset = paddle.vision.datasets.MNIST(mode='train')
    test_dataset = paddle.vision.datasets.MNIST(mode='test')
    x = paddle.data(name="x", shape=[64, 1, 28, 28], dtype='float32
    ↪')
    y = paddle.data(name="y", shape=[64, 1], dtype='int64')
    x_flatten = fluid.layers.reshape(x, [64, 784])
    fc_1 = nn.fc(input=x_flatten, size=128, act='tanh')
    fc_2 = nn.fc(input=fc_1, size=128, act='tanh')
    prediction = nn.fc(input=[fc_2], size=10, act='softmax')
    cost = fluid.layers.cross_entropy(input=prediction, label=y)
```

(çzäyLéa)

```

acc_top1 = fluid.layers.accuracy(input=prediction, label=y, k=1)
avg_cost = fluid.layers.mean(x=cost)
return train_dataset, test_dataset, x, y, avg_cost, acc_top1

paddle.enable_static()
train_data, test_data, x, y, cost, acc = mnist_on_mlp_model()
place = paddle.CUDAPlace(int(os.environ.get('FLAGS_selected_gpus', 0)))
train_dataloader = paddle.io.DataLoader(
    train_data, feed_list=[x, y], drop_last=True,
    places=place, batch_size=64, shuffle=True)
fleet.init(is_collective=True)
strategy = fleet.DistributedStrategy()
#optimizer = paddle.optimizer.Adam(learning_rate=0.01)
optimizer = fluid.optimizer.Adam(learning_rate=0.001)
optimizer = fleet.distributed_optimizer(optimizer, strategy=strategy)
optimizer.minimize(cost)

exe = paddle.static.Executor(place)
exe.run(paddle.static.default_startup_program())

epoch = 10
for i in range(epoch):
    total_time = 0
    step = 0
    for data in train_dataloader():
        step += 1
        start_time = time.time()
        loss_val, acc_val = exe.run(
            paddle.static.default_main_program(),
            feed=data, fetch_list=[cost.name, acc.name])
        if step % 200 == 0:
            end_time = time.time()
            total_time += (end_time - start_time)
            print(
                "epoch: %d, step:%d, train_loss: %f, total time_
cost = %f, speed: %f"
                % (i, step, loss_val[0], total_time,
                   1 / (end_time - start_time) ))

```

5.5.1 5.5.1

ärEäyLèçřäzççäAäfiäYäIJltrain.pyäzççäAäyijNäTæIJžäTäæèöçzÇäAäLÈçŽĐçóÄäTijNä

```

export CUDA_VISIBLE_DEVICES=0
python train.py

```


āŕŕāzēçIJŇēğAçZŁçñŕäyŁæL'SāŕæŮčāŁŮāŁæAŕiijŽ

```
epoch: 0, step:200, train_loss: 0.424425, total time cost = 0.
→000947, speed: 1055.967774
epoch: 0, step:400, train_loss: 0.273742, total time cost = 0.
→001725, speed: 1285.413423
epoch: 0, step:600, train_loss: 0.472131, total time cost = 0.
→002467, speed: 1347.784062
epoch: 0, step:800, train_loss: 0.445613, total time cost = 0.
→003184, speed: 1394.382979
epoch: 1, step:200, train_loss: 0.512807, total time cost = 0.
→000681, speed: 1468.593838
epoch: 1, step:400, train_loss: 0.571385, total time cost = 0.
→001344, speed: 1508.199928
epoch: 1, step:600, train_loss: 0.617232, total time cost = 0.
→002034, speed: 1449.310297
epoch: 1, step:800, train_loss: 0.392537, total time cost = 0.
→002813, speed: 1283.446756
epoch: 2, step:200, train_loss: 0.288508, total time cost = 0.
→000796, speed: 1256.155735
epoch: 2, step:400, train_loss: 0.448433, total time cost = 0.
→001531, speed: 1360.461888
epoch: 2, step:600, train_loss: 0.593330, total time cost = 0.
→002292, speed: 1314.005013
...
```

5.5.2 āŕŕāzēçIJŇēğAçZŁçñŕäyŁæL'SāŕæŮčāŁŮāŁæAŕiijŽ

āzŌāŕŕāzēçIJŇēğAçZŁçñŕäyŁæL'SāŕæŮčāŁŮāŁæAŕiijŽ
pyāzēççāAŕiijŇāŕŕāzēçIJŇēğAçZŁçñŕäyŁæL'SāŕæŮčāŁŮāŁæAŕiijŽ

```
export CUDA_VISIBLE_DEVICES=0,1,2,3
fleetrun train.py
```

ēŌāŕŕāzēçIJŇēğAçZŁçñŕäyŁæL'SāŕæŮčāŁŮāŁæAŕiijŽ
çZŁçñŕäyŁæL'SāŕæŮčāŁŮāŁæAŕiijŽ

```
----- Configuration Arguments -----
gpus: 0,1,2,3
ips: 127.0.0.1
log_dir: log
server_num: None
servers:
training_script: train.py
training_script_args: []
worker_num: None
workers:
-----
INFO 202X-0X-0X 06:09:36,185 launch_utils.py:425] Local start 4
→processes. First process distributed environment info (Only for debug):
→Debug):
```

(çzäyLéa)

Distributed Envs	Value
↪-----	
PADDLE_CURRENT_ENDPOINT	127.0.0.1:33360
PADDLE_TRAINERS_NUM	4
FLAGS_selected_gpus	0
PADDLE_TRAINER_ENDPOINTS	... 0.1:11330,127.0.0.
↪1:54803,127.0.0.1:49294	
PADDLE_TRAINER_ID	0
=====	
epoch: 0, step:200, train_loss: 0.306129, total time cost = 0.	
↪001170, speed: 854.759323	
epoch: 0, step:400, train_loss: 0.287594, total time cost = 0.	
↪002226, speed: 947.009257	
epoch: 0, step:600, train_loss: 0.179934, total time cost = 0.	
↪003201, speed: 1025.752996	
epoch: 0, step:800, train_loss: 0.137214, total time cost = 0.	
↪005004, speed: 554.582044	
epoch: 1, step:200, train_loss: 0.302534, total time cost = 0.	
↪000975, speed: 1025.752996	
epoch: 1, step:400, train_loss: 0.375780, total time cost = 0.	
↪001934, speed: 1042.581158	
epoch: 1, step:600, train_loss: 0.247651, total time cost = 0.	
↪002892, speed: 1043.878547	
epoch: 1, step:800, train_loss: 0.086278, total time cost = 0.	
↪003845, speed: 1049.363022	
.....	

CHAPTER 6

FleetXå£néÄšäijÄğŃ

6.1 FleetXæŸřăžĂăžĹiijſ

FleetXæŘŘă;ŽæŢĹçŎĞæIJĂénŸçŽĎăĹĚăŸČăijŘăĹăđŃécĎěő■çzČăĹšèČ;iiĵŃăőČăŘřăžă;IJăŸpa
distributed.fleetçŽĎăĹ'ăſŢēŢŽēăŃēĚ■ăŘĹă;ŢçŢĹăĂĆ

6.2 æŘŘă;ŽăſłăžŽăĹšèČ;iiijſ

- çſ■ăžçăĂăőŽăžĹ'écĎěő■çzČăĹăđŃ
- écĎç;őçzŘăĚŸăĹăđŃçŽĎăĚŃăijĂěő■çzČăŢřă■ő
- çŢĹăĹăăŘřă;ŎăĹŘăIJŃăŽăæ■céĜăIJĹ'æŢřă■őéŽĚ
- éĹăăŘſăřăŸăĹăđŃçŽĎăIJăă;ſăĹĚăŸČăijŘěő■çzČăőđèùŢ

6.3 äŸĹæĹ'Ńçđ'žă;Ń

ăžăŸŸŃéĂŽèŢĜăŽ;ăČŘăĹĚçſzResnet50çŽĎă;Ńă■ŘiijŃěřŢ æŸŎăēČă;Ţă;ŢçŢĹFleetXçŽĎăŎăŘçèŢŽè

- âřijăĚă;ĹèŢŮ
- æđĎăžžăĹăđŃ
- âőŽăžĹăĹĚăŸČăijŘç■ŮçŢě
- âijĂăğŃèő■çzČ

ăŸžăžĚçőĂăŃŮăĹăđŃăőŽăžĹ'çŽĎēŢĜçĹŃiijŃăĹſăžŃăIJăăŘŎéĹççŽĎăŮĜăăçăŸ■ăijŽăř;éĜŘă;ŢçŢĹFl

6.3.1 1. ĀrijāĒēāzĭlētŪ

FleetXāzĭlētŪPaddle 1.8.0āRĻāzŅāRŌčŽĎčLĻāIJñāĀCērŭçāōēōd'āūsāōL'ēčĒā■čçāōčŽĎ-PaddlečLĻāIJñijŅāzŭāŅL'čĒģāzēāyŅāŪzāijRārijāĒēPaddle āRĻ FleetXāĀČ

```
import paddle
import paddle.distributed.fleet as fleet
import fleetx as x
```

6.3.2 2. æĎĎāzžæĭāđŅ

ēĀŽēfGFleetXæRĪāzĭŽčŽĎX.applications æŌēāRčĭijŅčŦĭāLŭāRfāzēāzĭčŦĭāyĀēāŅāzčçāAāLāē

```
paddle.enable_static()
configs = X.parse_train_configs()

model = X.applications.Resnet50()
downloader = X.utils.Downloader()
local_path = downloader.download_from_bos(fs_yaml="https://xxx.xx.xx.
→xx/full_imagenet_bos.yml", local_path='./data')
loader = model.get_train_dataloader(local_path, batch_size=32)
```

6.3.3 3. āōŽāzL'āLĒāyČāijRč■ŪčŦē

āIJāōŽāzL'āōŅā■ŦæIJžēō■čžČç;ŠçzIJāRŌĭijŅčŦĭāLŭāRfāzēāzĭčŦĭpaddle. distributed.fleet.DistributedStrategy() æŌēāRčāōŽāzL'āLĒāyČāijRč■ŪčŦēĭijŅārĒāĭāđŅ

```
# āzĭčŦĭpaddle.distributed.fleetēfŽēāŅcollective training
fleet.init(is_collective=True)
# āōŽāzL'DistributedStrategy
dist_strategy = fleet.DistributedStrategy()
# ēčĒēēřā■ŦæIJžoptimizerāyžāLĒāyČāijRoptimizer
optimizer = paddle.distributed.fleet.distributed_
→optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)
```

6.3.4 4. āijĀāgŅēō■čžČ

āRfāzēāzĭčŦĭFleetXāĒēç;ōčŽĎēō■čžČāZĭēfZēāŅāfŅēĀšēō■čžČĭijŅāŪzāzĭčōŪæšŦāūēčĭŅāyLāfŅē

```
trainer = X.MultiGPUPUTrainer()
trainer.fit(model, loader, epoch=10)
```

čŦĭāLŭāzžāRfāzēēGčŦĭPaddleāŌšçŦŦšçŽĎAPIēfZēāŅēō■čžČætAçĭŅčŽĎāōŽāzL'ĭijŅāzčçāAāēČāyŅ

```

exe = fluid.Executor(place)
exe.run(fluid.default_startup_program())

for epoch_id in range(5):
    step_id = 0
    for data in loader:
        cost_val = exe.run(paddle.default_main_program(),
                           feed=data,
                           fetch_list=[model.loss.name])
        if step_id % 100 == 0:
            print("worker index: %d, epoch: %d, step: %d, train_
↪loss: %f"
                  % (fleet.worker_index(), epoch_id, step_id, cost_
↪val[0]))

```

azPaddle 2.0 rcL'LaIJnaijAagNijNaeL'Sazncz\$ayAeGcTlfleetruneZeaNad'Za■aeo■czCzZDaRraL

```

fleetrn --gpus 0,1,2,3 resnet_app.py

```

aEsazOfleetrnaSjazd'ijNaeZt'ereczEczZDa;fcTlert'aeYOfuaRCeAcfleetrn

ä|£çŦfleetsubæŦŦäžd'éŽ£ç¿d'äzzåŁą

7.1 fleetsubæYrázÄäZL

a̟ʃəCíáoL'ècĚäZēfleet-xaŕŔŌijNăꞤáRřázë;ɬçTífleetsubaIjléZEç;d'äyŁæRŘāzd'alÉayČajRa

7.2 ä;£çŤíèęAæśĆ

ä;ɸçTÍfleet subåŚ;äzd'çŽDëçAæśĆiijŽăoL'èçĚfleet-x

- ãÄŖæŮzæşŦäyÄãÄŚäzŎpipæžŖåŏL'èčĚ

```
pip install fleet-x
```

- ăĂŖæŮzæʂTăžŇăĂŠäyNèj;whlăŇĚázúăIĬăIĬňăIĬrăŎL'èčĚ

```
# python2
wget --no-check-certificate https://fleet.bj.bcebos.com/fleet_x-0.0.
↪4-py2-none-any.whl
pip install fleet_x-0.0.4-py2-none-any.whl
# python3
wget --no-check-certificate https://fleet.bj.bcebos.com/fleet_x-0.0.
↪4-py3-none-any.whl
pip3 install fleet_x-0.0.4-py3-none-any.whl
```

7.3 ä;£çTlèrt'æYÖ

åIJlæRRäzd' äzzåLååL' ■ijNçTlæLüéIJÀèçAåIJlyamlæÜGäzüäy■éË■ç;öäzzåLaçZÿåËşçZDäfaæArijNäç
éçÜåËLçIJNäyÄäylyamlæÜGäzüçZDæåüä;NäÄCäZäyžäfaæAraöL' äËlçZDäÖşåZäyamlæÜGäzüäy■ç;

```
num_trainers: 4
num_cards: 8
job_prefix: bert_base_pretraining
image_addr: ${image_addr:-"dockerhub.com/paddlepaddle-public/paddle_
➔ubuntu1604:cuda10.0-cudnn7-dev"}
cluster_name: v100-32-cluster
group_name: k8s-gpu-v100-8
server: paddlecloud.server.com
log_fs_name: "afs://xx.fs.com:9902"
log_fs_ugi: "ugi_name,ugi_passwd"
log_output_path: "/xx/yy/zz"
file_dir: "./"

whl_install_commands:
- pip install fleet_x-0.0.5-py2-none-any.whl
- pip install paddlepaddle_gpu-0.0.0-cp27-cp27mu-linux_x86_64.whl

commands:
- fleetrn bert_base.py --download_config=bert.yaml
```

å■ÜæöüåR■çgř	å■ÜæöüåRñäzL'	çşzåđN
num_trainers	èö■çzÇèLÇçCzçZDæTřéGR	INT
num_cards	å■TèLÇçCzäyLçTřèrûçZDGPUå■æTř	INT
job_prefix	äzzåLååR■åL'■çijÄ	STRING
image_addr	éTlJåČRåIJraIA	{STRING}
cluster_name	éZEçç;d'åR■	STRING
group_name	çç;d'çzDåR■	STRING
server	éZEçç;d'masterèLÇçCzæIJ■åLååR■	STRING
log_fs_name	äzzåLæÜèåfÜå■YæTççZDæÜGäzüçşçzşåR■	STRING
log_fs_ugi	äzzåLæÜèåfÜå■YæTççZDæÜGäzüçşçzşUGI	STRING
log_output_path	äzzåLæÜèåfÜå■YæTççZDçZöæåGæÜGäzüçşçzşSARINIA	STRING
file_dir	æRRäzd' äzzåLæIJÀèçAäyLäijäçZDæÜGäzüçZöåS	STRING
whl_install_commands	åöL'èçËåRDçg■wheelåNËçZDåS;äzd'	Repeated Command Line
commands	èfRèaÑäzzåLæL'gèaNçZDåRDçg■åS;äzd'	Repeated Command Line

7.4 äzzåŁæŘŘäzd'

åóŽázL'áoNäyLè£řèĎŽæIJñåŘŮijNčTíæLũ■şåRfä;£çTífleetsubåŚ;äzd'åŘŚPaddleCloud
æŘŘäzd' äzzåŁæžEijŽ

```
fleetsub -f demo.yaml
```

7.5 ä;£çTíæăũă;N

ăĚüä;ŞçŽĎä;£çTíèft'æYŎåRLæăũă;NăžčăAęérûåRCèĀčăyNéİççŽĎWIKI

ä|£çŦíFleetè£ŽèąŇăŔĆæŦřæ|J■åŁąăŻíèő■çžČ

āIJlād' gæTṛæ■ōætḥæ;ōçŽDæŌlāLīäyNiiJNæIjL' æāGç■;èö■çzČæTṛæ■ōçŽDëgDælaaRŪa; ŪāžEēčđéĀšç
 Speech 2çszçzçšaj;ŁçTlāžE11940ārRæUūçŽDēr■ēsşæTṛæ■ōāzēāRŁeūĒēfG200äyGaRēēalēfrālēēō■çzČēr■ēs

äyžăẼæRṚénYæÍaǻđNçZĐèő■çzCæȚLçŎĞiijNăLÊăȳČăijRëő■çzČăzTetĚŘěĂNçȚșiiȳNăĚŭăy■ășzăžŎ

- **ēō■ꝛꝥČēŁĆꝥČzīijŽēēēēŁĆꝥČzēťšēťčāōNæŁRæTŕæ■ōēŕzāRŪāĀAāL'■āŘSēōaꝥōŪāĀAāR'■āŘSæcŕāžē**
- **æIJ■āŁaēŁĆꝥČzīijŽāIJlæTūāŁŕæL'ĀæIJL'ēō■ꝛꝥČēŁĆꝥČzāijælēčŽDæcŕāžēāŘŌīijNēēēēŁĆꝥČzāijŽā**
æāžæ■ōāŕCæTŕæŽťæŪŕčŽDæŪzāijRāy■āŔNīijNāŔŕāžēāŁEāyžāŔNæ■ēāŠNāijCæ■ēāyđ'çg■īijŽ
- **āŔNæ■ēēō■ꝛꝥČīijŽāIJlāŔNæ■ēāŕCæTŕæIJ■āŁaāŽlāŁEāyČāijRēō■ꝛꝥČäy■īijNæL'ĀæIJL'ēō■ꝛꝥČēŁĆꝥČ**
- **āijCæ■ēēō■ꝛꝥČīijŽāyŌāŔNæ■ēēō■ꝛꝥČäy■āŔNīijNāIJlāijCæ■ēēō■ꝛꝥČäy■āzza;Tāyđ'āylēō■ꝛꝥČēŁĆꝥČ**
āyNélcæŁSāznārEēĀŽēŁGā;Nā■ŔīijNāyžæCīāzNčz■āŔNæ■ē/āijCæ■ēēō■ꝛꝥČČāIJlFleetāy■čŽDāōđČŌŕā
āIJlāijĀagNāzNāL'■āŁSāznēēŪāĒŁÉIJĀēēAāyNē;ēō■ꝛꝥČäy■æL'ĀéIJĀēēAçŽDæTŕæ■ōīijŽ

```
# äYÑè;;āzűëğčāŌNāTřā■ōiijÑèč■čžČřā■ōèōšä£İā■YèGşāŘ■äyž raw_data_
→čžDăŮĠĞäzűăđ'ž
wget --no-check-certificate https://fleet.bj.bcebos.com/ctr_data.
→tar.gz
tar -zxvf ctr_data.tar.gz
```

8.1 ăĩđçŦĩæăũă;Ŧ

äyÑéÍcæŁsăznæİēāzŃcz■āęĆä;TçTİFleetæŌēāRćiiJŃăoŃNæŁŔăRĆæTŕæIJ■āŁaāZİŁŁEāyČaijRèő■çzÇŕ

8.1.1 áRíjāĚěäĹèŧŮ

```
import paddle
import os
import fleetx as X
import paddle.fluid as fluid
import paddle.distributed.fleet as fleet
import paddle.distributed.fleet.base.role_maker as role_maker
```

8.1.2 áŖŽāZĹ'āĹĚāŸČāijRāēāijRāzŮāĹiāġNāŇŮ

éĀŽèĹGX.parse_train_configs() æŖēāRčĭijŇčŤĹāĹāRřāzēāŖŽāZĹ'èŏ■čžČžŸāĚšçŽĎāRCæŤ
init() æŖēāRčāŖŽāZĹ'āžĚāĹĚāŸČāijRāēāđNĭijŇinit() æŖēāRčézŸèŏđ'ā;ĹčŤĹāRCæŤřāIJ■āĹāŽĹāēā
āZĹ'āžzā;ŤāRCæŤřāĀČ

```
paddle.enable_static()
configs = X.parse_train_configs()
role = role_maker.PaddleCloudRoleMaker()
fleet.init(role)
```

8.1.3 āĹāēĭĭāēāđNāRĹæŤřæ■ŏ

čŤĹāĹāRřāzēéĀŽèĹGX.applicationsæŖēāRčāĹāēĭĭāĹSāznécĎāĚĹāŖŽāZĹ'āē;çŽĎāēāđNāĀČā
DNNāēāđNĭijŇāRŇæŮŸčŤĹāĹāRřāzēāŸzāēāđNāŏŽāĹŮčŽĎdata_loaderæŖēāRčāĹāēĭĭæŤřæ■ŏ.

```
model = X.applications.MultiSlotCTR()
loader = model.load_criteo_from_file('./train_data')
```

8.1.4 áŖŽāZĹ'āRŇæ■ēēŏ■čžČ Strategy āRĹ Optimizer

āIJĹFleet APIāŸ■ĭijŇčŤĹāĹāRřāzēā;ĹčŤĹfleet.DistributedStrategy() æŖēāRčāŖŽāZĹ'ēġĹāŮ
āĚŮāŸ■a_syncéĀĹēāžčŤĹāžŖāŖŽāZĹ'āRCæŤřāIJ■āĹāŽĹčžŸāĚšçŽĎç■ŮčŤēĭijŇā;ŖāĚŮēčēŏ;āŖŽāŸ

```
dist_strategy = fleet.DistributedStrategy()
dist_strategy.a_sync = False

optimizer = fluid.optimizer.SGD(learning_rate=0.0001)
optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)
```

8.1.5 āijĀāġNēŏ■čžČ

āŖŇāĹRāēāđNāRĹēŏ■čžČç■ŮčŤēāzēāŖŖŏĭijŇāĹSāznārsāRřāzēāijĀāġNēŏ■čžČāēāđNāžĚāĀČāZāāŸ

`fleet.init_server()` to start the server and `fleet.init_worker()` to start the worker. The `fleet` module also provides a `run` function to run the server and worker in a single process. The `run` function takes the following arguments:

```

if fleet.is_server():
    fleet.init_server()
    fleet.run_server()
else:
    fleet.init_worker()
    trainer = X.Trainer(fluid.CPUPlace())
    trainer.fit(model, loader, epoch=10)
  
```

8.1.6 Running FleetX

To run FleetX, you need to install the `fleetx` package. You can install it using `pip` or `conda`. The following command will install the latest version of `fleetx`:

```
fleetrun --server_num=1 --worker_num=2 ctr_app.py
```

ä;£çŤInMemoryDataset/QueueDatasetè£ŽèąÑěő■çžČ

9.1 æşíaěŘ

æIŋæTʒɹlNçZõAŁ■äy■æTræÑAåLíæÄAåZjiiñNäZĖæTřæÑAåIłpaddleéÍZæÄAåZ;ælaaijRäyNä;ɬçTł.

```
paddle.enable_static()
```

9.2 çŒÄžŇ

äyžāẼēĈ;énŸéĀšēƒRēāÑæĭāāđNĉŽĐēő■čȝCĭijÑæĹŚāžñä;ƒçTĬInMemoryDataset/
QueueDatasetAPIēƒZēāNénŸæĀğēĈ;ȝŽDIOĭijÑāĒüā;ŞāžNĉȝ■āRřāžēāŔĈēĀĈæŨĜæāçInMemoryDat
āŠŇ QueueDataset, āžēäyŇĉōĀçğŕDatasetāĀĈDatasetæŸřäyžād'ŽčȝƒĈĬNāŔĹāĒĭāijĈæ■ēāŨžāijŔēĜŔēž
ād'ŽæŭĹēŕ'zēĀĒçŽĐæĭāāijŔĭijÑāijŽæđĀāđ'ğçŽĐāĹæĀŸæĹŚāžñçŽĐæĭāāđNēő■čȝĈāĀĈ

æIjnæŨĜäzèèö■çžČword2vectorælaadŇäyžä;ŇiiĴŇâIJlèö■çžČäy■aiĴTâĖĖâšžäžŒDataset
APIèržâRŨlèö■çžČæTṙæ■öçŽDæŨžaiĴRiiĴŇâĴSäzñçŽTṙæŒĖâĴäè;ĴFleetxéçDâĖĴĴâŒžäZĴĴä;ççŽDword2vector

9.2.1 ĀijTāĒdataset

1. `dataset = paddle.distributed.InMemoryDataset()`
`dataset = paddle.distributed.QueueDataset()`
2. `dataset.set_filelist()`
3. `dataset.init()` `api` `dataset` `init()`

- āŕĖæĹŚāznāōŽāzĹ'āē;çŽDæŦŕæ■ōē;ŚāĒēāiĵāiĵŔāiĵāçzŽDataset,
éĀŽēfGuse_varēĒ■ç;ōāĀĆ
- æŊĜāōŽæĹŚāznçŽDæŦŕæ■ōērzaŔŪæŪzāiĵŔiĵŊçŦśmy_data_generator.
pyāōđçŌŕæŦŕæ■ōērzaŔŪçŽDēgDāĹZiĵŊāŔŌēicārĖāiĵŽāzŊçz■ērzaŔŪēgDāĹZçŽDāōđçŌŕ,
éĀŽēfGpipe_commandēĒ■ç;ōāĀĆpipe_commandæŦŕDatasetçĹ'zāIJĹçŽDēĀŽēfGçōāēAŚ
- æŊĜāōŽæŦŕæ■ōērzaŔŪçŽDbatch_sizeiĵŊēĀŽēfGbatch_sizeēĒ■ç;ōāĀĆ
- æŊĜāōŽæŦŕæ■ōērzaŔŪçŽDçzĹçĹŊæŦŕiĵŊāyĀēĹŊērēçzĹçĹŊæŦŕāŚŊēō■çzÇçzĹçĹŊāzŦāĹæŊā

```
dataset = paddle.distributed.InMemoryDataset()
batch_size = config.config["batch_size"]
thread_num = config.config["thread_num"]
dataset.init(use_var=model.inputs, pipe_command="python my_data_
→generator.py", batch_size=batch_size, thread_num=thread_num)
dataset.set_filelist([config.config["train_files_path"]])
```

9.2.2 æĆäȚæŃǦăőŽæȚræ■øérzàRÚèğĐǻŁŻ

aIJlāyLæŨĜæĹSāznæRŘăĹrāžEçŤsmy_data_generator.
 pyāōđçŎŖăĒüă;ŞçŽDæŤræ■ōçōăĀŞşērZăRŨēĜĐăĹZīijNéCčázĹīijNæĂŎăăüăyždatasetăĹZăžzăŤræ■ōērZăĹ
 äžăyNæŸrmy_data_generator.pyçŽĐăĒĹéĹĹăžçăĀīijNăĒüă;ŞætĀçĹNăēCăyNīijŽ

1. `ēēŨāĒĹāĒŚāznēIJāĒēēAāijTāĒēēdata_generatorčŽDčsziiJNā;āžŌpaddle.
distributed.fleet.data_generatorāĀĆ` 2. `āčrāŸŌāyĀāžZāIJĹāTřrā■ōērzāRŪāy■āijŽčTĹāĹrčŽD`
3. `āĹZāžžāyĀāyĹā■ŘčszWord2VecReaderīijNčžgāĹĤfleet.
data_generatorčŽDāšžčszīijNāšžčszāIJĹāđ'Žčg■ēĀĹāēNĹ'īijNāēČādIJāŸřād'Žčg■āTřrā■ōčszādNāu`
4. `čžgāĹĤāžŭāōđčŌřāšžčszāy■čŽDgenerate_sampleāĠ;āTřīijNēĀŘēāNēřzāRŪāTřrā■ōāĀČēřēāĠ;āT
nce_reader() īijNāĒŚāžnāōŽāZĹāTřrā■ōērzāRŪčŽDēĀžē;ŚāĀČā;NāēČāržāžēēāNāyžā■Tā;■čŽDāTřrā■`
5. `āIJĹēĤZāyĹāŘřāžēēĤ■āžččŽDāĠ;āTřāy■īijNāēČčđ'žā;NāžččāAāy■čŽDdef
nce_reader() īijNāĒŚāžnāōŽāZĹāTřrā■ōērzāRŪčŽDēĀžē;ŚāĀČā;NāēČāržāžēēāNāyžā■Tā;■čŽDāTřrā■`
6. `āIJĀāŘŌīijNāĒŚāžnēIJāĒēēAārEāTřrā■ōāTt'čŘēāyžčĹ'žāōŽčŽDbatchčŽDāīijāijRīijNāĹ■ēČ;ād'šēč
āāžā■ōēō;āōŽčŽDāĀŽbatch_sizeāĀŽ, ēřēāĠ;āTřāijZāIJĹgenerator_sampleāĠ;āTřāžgčTšāāuā
local_iter() āĀĆ`
7. `čŌĀā■TāĹēēřt'īijNāTřrā■ōčŽDē;ŠāĠžēāžāžŘāyŌāĒŚāžnāIJĹ;ŚčzIJāy■āĹZāžžčŽDinputsāĤĒēāžāē
[value]), ('true_label', [value]), ('neg_label', [value])`

```
import sys
import io
import os
import re
import collections
import time
import config
import logging

import paddle
import numpy as np
import paddle.distributed.fleet as fleet
```

(äyÑeātçzğçz■)

(çz■ăŸŁéął)

```

logging.basicConfig(format='% (asctime)s - %(levelname)s -
→%(message)s')
logger = logging.getLogger("fluid")
logger.setLevel(logging.INFO)

class NumpyRandomInt(object):
    def __init__(self, a, b, buf_size=1000):
        self.idx = 0
        self.buffer = np.random.random_integers(a, b, buf_size)
        self.a = a
        self.b = b

    def __call__(self):
        if self.idx == len(self.buffer):
            self.buffer = np.random.random_integers(self.a, self.b,
→len(self.
buffer))
            self.idx = 0

        result = self.buffer[self.idx]
        self.idx += 1
        return result

class Word2VecReader(fleet.MultiSlotDataGenerator):
    def init(self,
            dict_path,
            nce_num,
            window_size=5):

        self.window_size_ = window_size
        self.nce_num = nce_num

        word_all_count = 0
        id_counts = []
        word_id = 0

        with io.open(dict_path, 'r', encoding='utf-8') as f:
            for line in f:
                word, count = line.split()[0], int(line.split()[1])
                word_id += 1
                id_counts.append(count)
                word_all_count += count

        self.word_all_count = word_all_count
        self.corpus_size_ = word_all_count
        self.dict_size = len(id_counts)
        self.id_counts_ = id_counts

```

(ăŸŊéąłçzçz■)

(čŹăŸŁéął)

```

        logger.info("corpus_size:", self.corpus_size_)
        self.id_frequencies = [
            float(count) / word_all_count for count in self.id_
↪counts_
        ]
        logger.info("dict_size = " + str(self.dict_size) + " word_
↪all_count = " + str(word_all_count))

        self.random_generator = NumpyRandomInt(1, self.window_size_
↪+ 1)

    def get_context_words(self, words, idx):
        """
        Get the context word list of target word.
        words: the words of the current line
        idx: input word index
        window_size: window size
        """
        target_window = self.random_generator()
        start_point = idx - target_window # if (idx - target_
↪window) > 0 else 0
        if start_point < 0:
            start_point = 0
        end_point = idx + target_window
        targets = words[start_point:idx] + words[idx + 1:end_point_
↪+ 1]
        return targets

    def generate_batch(self, samples):
        def local_iter():
            np_power = np.power(np.array(self.id_frequencies), 0.75)
            id_frequencies_pow = np_power / np_power.sum()
            cs = np.array(id_frequencies_pow).cumsum()
            result = [[], []]
            for sample in samples:
                tensor_result = [("input_word", []), ("true_label",
↪[]), ("neg_label", [])]
                tensor_result[0][1].extend(sample[0])
                tensor_result[1][1].extend(sample[1])
                neg_array = cs.searchsorted(np.random.sample(self.
↪nce_num))

                tensor_result[2][1].extend(neg_array)

            yield tensor_result
        return local_iter

```

(ăŸŇéąłčŹčŹł)

9.3 安装

安装 FleetX 0.1.0.beta:

```
trainer = X.CPUTrainer()
trainer.fit(model, loader, epoch=10)
```

安装 FleetX 0.1.0.beta 需要安装以下依赖包，安装命令如下：

```
import paddle
import paddle.fluid as fluid
import paddle.distributed.fleet as fleet
import config
# 安装 paddle 依赖包
paddle.enable_static()

fleet.init()

model = X.applications.Word2vec()

"""
need config loader correctly.
"""

loader = model.load_dataset_from_file(train_files_path=[config.
    config["train_files_path"]], dict_path=config.config["dict_path"])

dist_strategy = fleet.DistributedStrategy()
dist_strategy.a_sync = True

optimizer = fluid.optimizer.SGD(learning_rate=0.0001)
optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)

if fleet.is_server():
    fleet.init_server()
    fleet.run_server()
else:
    place = paddle.CPUPlace()
    fleet.init_worker()
    exe = paddle.static.Executor(place)
    default_startup_program = paddle.static.Program()
    default_main_program = paddle.static.Program()
    scope1 = fluid.Scope()
    with fluid.scope_guard(scope1):
        exe.run(model.startup_prog)

dataset = paddle.distributed.QueueDataset()
```

(安装 FleetX 0.1.0.beta)

(çzäyLéat)

```

batch_size = config.config["batch_size"]
thread_num = config.config["thread_num"]
dataset.init(use_var=model.inputs, pipe_command="python my_data_
→generator.py", batch_size=batch_size, thread_num=thread_num)
dataset.set_filelist([config.config["train_files_path"]])

with fluid.scope_guard(scopel):
    exe.train_from_dataset(model.main_prog, dataset, scopel,
→debug=False, fetch_list=[model.loss], fetch_info=["loss"], print_
→period=10)

fleet.stop_worker()

```

æIJĀŘŌæüzāLäyLēřäzččāAä;ŁçŤlčŽĐéĚ■ç;őæŮĜäzŮconfig.py

```

config = dict()

config["dict_path"] = "thirdparty/test_build_dict"
config["train_files_path"] = "demo_train_data/part_1"
config["batch_size"] = 1000
config["nce_num"] = 5
config["thread_num"] = 12

```

éŽžŁĜäžäyŁčŏĀæt' AčŽĐäzččāAĭijNā■şāŔřäzēāóđčŎřword2vectoræłāđNčŽĐāđ'ŽçŁčlNāžŮāŔSèó

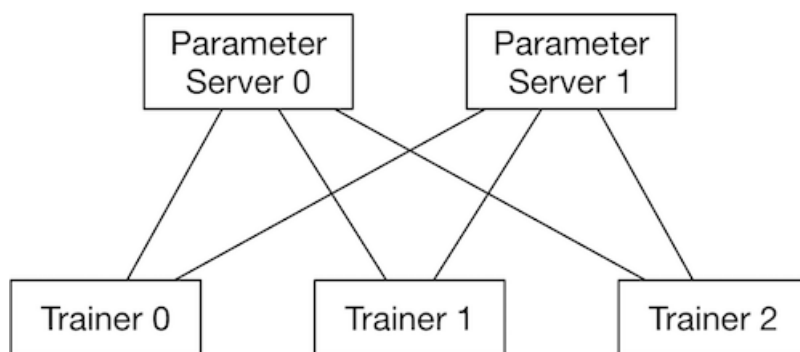
Collective Communication Primitives

10.1 All-to-All Communication

Figure 10.1 illustrates the All-to-All communication primitive. In this scenario, every process (Trainer) must receive data from every other process (Trainer). This is achieved by having each Trainer send its data to a central Parameter Server (PS) and then receive the aggregated data back from the PS. The PS acts as a central hub for the data exchange.

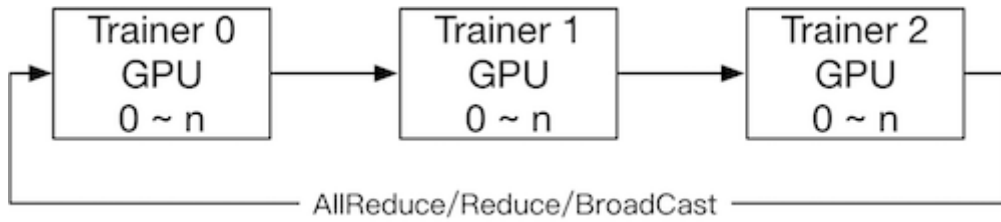
The diagram shows three Trainers (Trainer 0, Trainer 1, Trainer 2) and two Parameter Servers (Parameter Server 0, Parameter Server 1). Each Trainer sends its data to both Parameter Servers. The Parameter Servers then aggregate the data and send the result back to each Trainer. This ensures that every Trainer receives data from every other Trainer.

- All-to-All communication is a fundamental primitive in distributed systems. It is used in various applications, including data aggregation, model averaging, and consensus algorithms. The diagram illustrates how this primitive can be implemented using a central Parameter Server (PS) and multiple Trainers.



- All-to-All communication is a fundamental primitive in distributed systems. It is used in various applications, including data aggregation, model averaging, and consensus algorithms. The diagram illustrates how this primitive can be implemented using a central Parameter Server (PS) and multiple Trainers.

aṚṢäyĈ eäcražēāLræL' ÄæIJL' trainersiiĴNæIJÄāRÖæfRäyĴ trainer
ä;ŁçTlāRṢäyĈ eäRÖçZĎæcražēçNñçñNāōNæL'RāRĆæTṛæZt' æŰrāĈ



çZyāzĎ' äžÖäijCæ eēōçZĈ, aṚṢäyĈ eēōçZĈZĎçZĎäijYāŁāIJlāžÖLossāRrāžēæfTè;ČçlšāōZçZĎäyN
āZäæd' āIJlèōçZĈē;Čäyžād' æIČçZĎælaadNæŰüijNāšælaadNèōçZĈēŁGçlNäyçēdçZṚç;ŠçZIJèōçZĈē

Fleetäy PServerälaaijRä;ŁçTl gRPC éÄžäfaaijNCollectiveälaaijRä;ŁçTl NCCL2
éÄžäfaāĈ

äyNæŰGärEçTšäyL' éČlāŁEçZĎæL'RiiĴ

- äžNçZ Fleet aṚṢäyĈ eēōçZČäy äyçTlçZĎāGäyŁçŰçTē āÄÄaijYāNŰ
- çZŠāRĬäyŁèfṛäyçTlāijYāNŰüijNçZŽāGžäyÄäyŁāIJl 4èŁČçZ 32 V100 éŽEç;Ď' èōçZĈ ResNet50çZĎçĎ' žä;NāžčçāA
- āōNæTt' Fleet aṚṢäyĈ eēōçZČāRĆæTṛçŰçTēäžNçZ

10.2 Fleet Collective aṚṢäyĈ eēōçZČäijYāNŰ

Fleet æTṛæNāāIJl GPU (CUDA çL'ŁæIJñ >= 7.5)
æIJāŁāāZlÉZEç;Ď' äyŁāōNæL'RénYæÄgèČ;āŁEäyČaijRèōçZČāĈ

çTlāŁāRrāžēéÄžēŁG fleet.DistributedStrategy

èōç;ç;ōèōyād' ŽäyÖèōçZČæÄgèČ;çŰçTēçZyāEšāRĆæTṛāĈçZōāL Fleet

äyžēŁZāžŽāRĆæTṛæRṚä;ŽāžEäyÄäyŁè;ČéÄžçTlézYēōĎ' āÄijiiĴNçTlāŁāRrāžēäyāÖžèČæTt' āĈä;Eāç

āIJlèŁZèāNæÄgèČ;aijYāNŰæŰüijN æČæšēæfRéazaijYāNŰçZāžüēlNērAāržāžTæRṚāGüijNæIJÄç
äyÄäyŁçōÄāTçZĎélnērAā;ŠāLçZĎèōçZČlNāžRæYṛāRçéIJÄēAēŁZäyÄæäijYāNŰæÄgèČ;ZĎæŰ
æYṛæšēçIJNGPUçZĎèōaçōŰāL'çTlçÖGüijNéÄžäyçTl nvidia-smi āš;äžĎ' æšēçIJNāĈ
āçČæĎIJGPUāL'çTlçÖGè;Čä;ÖüijNāŁZāRfèČ;āYāIJlè;Čād' gçZĎäijYāNŰçl'žēŰt' āĈ

äyNæŰGärEäžNçZārfæÄgèČ;ā;šāšē;Čād' güijNèōç;ç;ōéçŠçÖGæfTè;ČénYçZĎāGäyŁāRĆæTṛiiĴNērē

æšlæĎRiiĴ ä;ŁçTlNCCL2ælaaijRāŁEäyČaijRèōçZČæŰüijNéIJÄēAçāōāŁāfRäyŁèŁČçZèōçZČçL

- éŽRæIJžéGĜæuäyÄäžZæTṛæōüijNæāēāŁāŁEēĈāLrè;ČārSæTṛæōçZĎèŁČçZäyŁāĈüijLæŌlèR
- āIJlpythonäžčçāÄäyüijNæfRäyŁèŁČçZæfRäyŁpassāRlèōçZČāžžāōZçZĎbatchæTṛiiĴNæČæĎIJēŁZäy

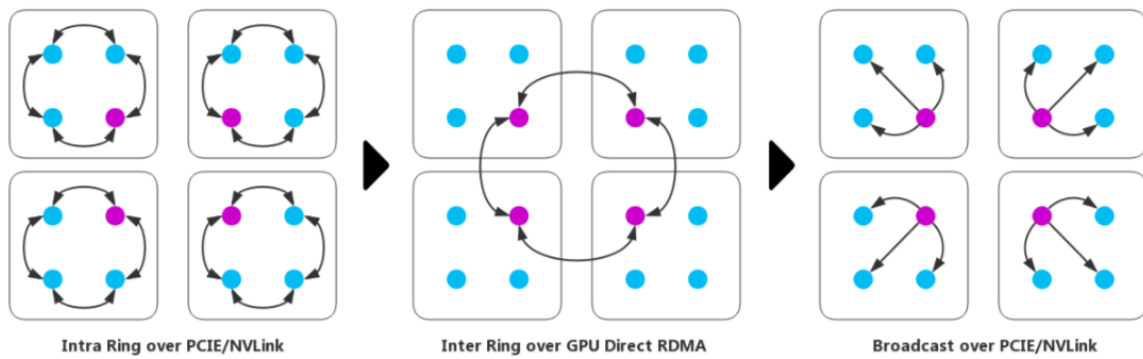
10.2.1 OPèĎāRĬ

ārEälaadNç;ŠçZIJäyēāžāžRæL'gēāNçZĎād'ŽäyŁOPsèŁZèāNèĎāRĬèČ;ād' šāGRārSOP
ērČāžççZĎäijÄéTāüijNæRṚāGèōçZČéÄšāžēāĈçZōāL Fleet
äyæTṛæNāāçCäyN3çgçZĎOP èĎāRĬüijŽ

- éĀŽāyvä; fçTlëfZāžZc■ŨçTëĈ; äiJŽā; fæTt'ä; ŠæL'gëaÑëfĠçlNæZt'āfñāĀĆ

æ■d' ad' ŪiijNäyžæTræŇAæZt' ad' gçšŠāžęçŽĐāŖĆæŦræcŕāžęęđ■āŖĹiijŇFleet
æŖŖä; ŽāžEäžęäyŇäy'd' äyléĀŖ' éāzīiijŇcŦlāŖLūāŖŕāžęāIJlő■czCclŇāžŖęŖŖēāŇāŖ■āIJlDistri

- æslæĐŘijŽ AllReduceēđ■āŘĹçŻōāL■äy■æŤræŃAşparseāŘĆæŤræćrāžẽāĈĆ



```
dist_strategy = fleet.DistributedStrategy()
dist_strategy.use_hierarchical_allreduce = True
dist_strategy.hierarchical_allreduce_inter_nranks = 8
```

10.2.4 ä;£çŦlâŦŦæ■ěAllreduce

Fleet ä;£çŦlâŦŦæ■ěAllreduce uses NCCL2 to perform collective operations. It is configured by setting the following parameters:

```
dist_strategy.sync_nccl_allreduce=True
```

10.2.5 ëŦç;ŦâŦŦæ■ěAllreduce

ncclAllreduce uses nccl_comm_num to specify the number of communication links between GPUs.

```
dist_strategy = fleet.DistributedStrategy()
dist_strategy.nccl_comm_num = 2
```

10.2.6 ëŦç;ŦâŦŦæ■ěAllreduce

PaddlePaddle Fluid uses the following parameters to configure the CPU threads used for Allreduce:

```
dev_count = 4
num_threads = 2
dev_count * num_threads = 8
```

```
dist_strategy = fleet.DistributedStrategy()
dist_strategy.thread_num = 3
```

10.2.7 æRŘénYç;ŚçzIJçŽDāRđāRŘ

ad'ŽèLĆçCzèõ■çzCæUūç;ŚçzIJçŽDāyēāō;āyāyāyæLŘäyžèõ■çzCçŽDçŚúécĹāĀCæĹSāznāIJlāōđætNāy
socket çŽDēĀŽāĤææUzāijRāŕEæĹRäyžèõ■çzCēĀšāžççŽDçŚúécĹiijN
 ä;Ĥad'ŽèLĆçCzèõ■çzCæUāæšTāĒĒāĹEāĹPçTĪFLeet æūūāRĹçš;āžçèōaçōUāyēæĪēçŽDēĀšāžæRŘā■
 āIJlāĹSāznāōđætNāy■ā;ĤçTĪ: 100Gb ç;Sā■āijNRDMA[7] āšN
 InfiniBand[8]æĪēæRŘā■Gç;ŚçzIJāyēāō;ijNā;Ĥç;ŚçzIJāijæē;šāy■āijZæĹRäyžèōaçōUēĀšāžççŽDçŚúécĹ
 āIJlāijĀāgNèõ■çzCāL■iijNēIJĀēæAæ■ççāōēōç;ōāžēāyN NCCL
 çŌřácCāRŸéGRä;ĤāřzāžTçañāzūēōç;ōçTšæTĹiijŽ

Env Name	Description
NCCL_SOCKET_IFNAME	The RDMA device, e.g. eth2
NCCL_P2P_DISABLE	Set to 1 to disable P2P transfer between GPUs
NCCL_IB_DISABLE	Set to 1 to disable using RDMA
NCCL_IB_CUDA_SUPPORT	Set to 1 to enable GPU Direct if supported
NCCL_DEBUG	Set debug level: VERSION, WARN, INFO

10.2.8 écDāĒĹāĹEēĒ■ēūšad'sçŽDæYç;ā■Y

éĀŽèĤGçŌřácCāRŸéGR FLAGS_fraction_of_gpu_memory_to_use=0.7
 èōç;ōécDāĒĹāĹEēĒ■çŽDæYç;ā■Yā■āærTāĀC çTšāžŌCUD-
 AāŌšçTšçŽDæYç;ā■YāĹEēĒ■cuMallocāšNēĜĹæTç;cuFreeæš■ā;IJāĪGæYŕāRŸæ■ēæš■ā;IJiijNēĪdāyēēĀUā
 éĀŽèĤG èōç;ō FLAGS_fraction_of_gpu_memory_to_use æĹRäyĀäyĹē;Čad'gççŽDāĀijijNærTāçC0.7iijNāR
 0.7 æYŕæNĜ 70%çŽDæYç;ā■YāijŽécDāĒĹāĹEēĒ■āĀCèōç;ōçŽDēNČāZŕ æYŕ0.0~1.0āĀC

```
os.environ['FLAGS_fraction_of_gpu_memory_to_use'] = "0.98"
```

10.2.9 éŽ■ä;Ōscope dropécŚçŌĜāšNfetchécŚçŌĜ

āGRārŚscope dropāšNfetchécŚçŌĜiijNāRŕäzēāGRārŚécŚçzAççŽDāRŸéGRāĒĒā■YçTšèŕuāĀæĜĹæTç;
 āzŌēĀNæRŘā■GæĀĝèČ;āĀC

```
# æŕR 30 batch äžNāRŌæyĒçRĒäyĀāñāyŕt'æUūāRŸéGR
dist_strategy = fleet.DistributedStrategy()
dist_strategy.BuildStrategy = {'num_iteration_per_drop_scope': 30}

# éŽ■ä;ŌfetchécŚçŌĜiijNærR 30 batch fetch äyĀāñāēō■çzCē;šāĜž
for pass_id in xrange(PASS_NUM):
    batch_id = 0
    while True:
        if batch_id % 30 == 0:
            fetched = exe.run(fetch_list)
        else:
            exe.run([])
```

10.2.10 áćďăď'gbatch_size

áĽĚäyČäijRäRŇæ■ēēő■čzČriijŇeũĽĽCćCzéĂžăĚæĽŮăď'ŽæĽŮăřSäijŽăyēæĽæĂğēČ;ă;ăS■iijŇăćďăđ
ăŔřăžěăĽæŇĂĚĂžăĚăăijĂéŤĂăy■ăŔŸčŽDăĈĚăĚăyŇiijŇăćďăď'ğēőăçőŮăŔďăŔŔăžŌēĂŇéŽ■ă;ŌéĂžăĚăă

10.2.11 ä;ĚčŤí DALI reader

æŤŕæ■őēřăŕŮčŽďäijŸăŇŮăĬĬGPUēő■čzČăy■ēĜşăĚşéĜ■ēēAriijŇăřď'ăĚŮăĬĬăy■æŮ■ăćďăĽăbatch_size
ăŔŕēČ;ăĽŔăyžēő■čzČéĂşăžēčŽďçŞüéĽăĂĆ Fleet äy■ăŔřăžěă;ĚčŤí Nvidia DALI6
ăĬĬăyžæŤŕæ■őloader. ä;ĚčŤíDALIčŽďäijŸčCžæĬĬ'riijŽ

- ä;ĚčŤíGPUăőŇăĽŔēČĽăĽĚæŤŕæ■őéćĐăď'ĐçŔĚriijŇăĽăĂşæŤŕæ■őēřăŕŮēĚĜĽŇriijŇăĜŔăŕS
CPU ēť şæŇĚăĂĆ
- DALI æŔŔă;ŽéćĐăŔŮēŸşăĽŮriijĽperfetchequeueiijĽăĽşēČ;riijŇēőĽ'æŤŕæ■őéćĐăď'ĐçŔĚăŞŇăĽăđŇă

```
import fleetx as X
model = X.applications.Resnet50()
loader = model.load_imagenet_from_file("/path/to/imagenet/train.txt",
↪ use_dali=True)
```

10.2.12 ä;ĚčŤíæũũăŔĽčş;ăžēēő■čzČ

V100 GPUæŔŔă;ŽăžĚ Tensor Core äŔřăžěăĬĬæũũăŔĽčş;ăžēēőăçőŮ
ăĬĬæŽŕăďĂăď'ğçŽďăŔŔă■ĜăĂğēČ;ăĂĆă;ĚčŤíæũũăŔĽčş;ăžēēőăçőŮčŽďă;Ňă■ăŔřăžěăŔĆēĂĆæŮĜăă
' <<https://todo/>>'__

čŽőăĽ'■PaddleăŔŔă;ŽăĬĬăyď'ăyĽăĽăđŇriijĽResNet, BERTriijĽčŽďăũũăŔĽčş;ăžēēőăçőŮăőđçŌŕăž
loss scalingriijŇăĚŮăžŮăĽăăđŇă;ĚčŤíæũũăŔĽčş;ăžēăžş äŔřăžěăŔĆēĂĆăžēăyĽčŽďăőđçŌŕăđŇăĽŔēĽŇăĽă

10.3 ResNet50ēő■čzČčď'žă;Ň

ēŕŤēŇăijĂăğŇăĽ'■ăĽŤăžŇăũşçžŔăĬĬGPU éŽĚč;ď'ăy■æŔŔăĽ'■éĚ■č;őăē; RDMA äŞŇ
InfiniBandriijŇăĜŔăŕSç;ŞçzĬĬéĂžăĚăçŽďçŞüéĽriijŇēĚ■č;őçzĚēĽĆăŞŇăĚŮă;ŞçăŇăžűčŽăăĚşriijŇăŔřăžěăŔĆ
x] <<https://community.mellanox.com/s/article/what-is-rdma-x>>'__

10.3.1 ēő;ç;ő AllReduceēďăŔĽč■ĽăŔĆæŤŕ

æćŕăžēēďăŔĽăy■čŽď16 äŞŇ 50 æŸŕăĽŤăžŇăăžăæ■őēĜĽēžŇç;ŞçzĬĬăŇăžűăŞŇResNet50
ēő■čzČēŕŤēŇă;ŮăĜççŽďçžŔēŇăăĬriijŇçŤĽăĽăŔřăžěăăžăæ■őēĜĽēžŇçăŇăžűăŞŇăĽăđŇăēĚžēăŇēŕČæŤŕăă
0.7 æŸŕăyžăžĚčž DALI loader æŔŔăĽ'■éćĐçŤŹæŸ;ă■ŸçĽ'žēŮŕ'ăĂĆ


```
import os
os.environ['FLAGS_fuse_parameter_memory_size'] = "16"
os.environ['FLAGS_fuse_parameter_groups_size'] = "50"
os.environ['FLAGS_fraction_of_gpu_memory_to_use'] = "0.7"
```

10.3.2

```
import os
import fleetx as X
import paddle
import paddle.fluid as fluid
import paddle.distributed.fleet.base.role_maker as role_maker
import time
import paddle.distributed.fleet as fleet
```

10.3.3

èŁŽéĜŇæĹŚäzñä;ŁçŤĪDALI reader äĜŖärŤCPU æŤŕæ■ōädŤĎçŖĒètŤšæŇĒäŤŇæŤŕæ■ōēržäŖŮçŤüécĹä

```
paddle.enable_static()
configs = X.parse_train_configs()
fleet.init(is_collective=True)

model = X.applications.Resnet50()
downloader = X.utils.Downloader()
local_path = downloader.download_from_bos(
    fs_yaml='https://fleet.bj.bcebos.com/test/loader/small_imagenet.
    ↪yaml',
    local_path='./data')
batch_size = 32
loader = model.get_train_dataloader(local_path, batch_size=batch_
    ↪size)
```

10.3.4

èŁŽéĜŇæĹŚäzñä;ŤäijÄäŖräyŁæŮĜäy■æŖŖäŖŁŖçŤĎäŖĎéäzēō■çžCäijŸäŇŮç■ŮçŤëijŇäçŤriijŽèĜĹäŁ
èĎ■äŖĹç■ŁäĀĀ

```
dist_strategy = fleet.DistributedStrategy()

# distributed strategy
dist_strategy.sync_nccl_allreduce = True
dist_strategy.nccl_comm_num = 2
dist_strategy.fuse_all_reduce_ops = True
```

(äŸŇéäŤçžĝç■)

(çzäyLéa)

```

cost_val = exe.run(model.main_prog,
                    feed=data,
                    fetch_list=[model.loss.name])

end_time = time.time()
print(
    "worker_index: %d, step%d cost = %f, speed: %f"
    % (fleet.worker_index(), i, cost_val[0], batch_size / (end_
    ↪time - start_time)))

```

10.3.6 Fleetrun äÄéTõäŘřáÍ

"xx.xx.xx.xxâĀĬ ç■LæYřāZŽäyĭlèŁĆćĆzipsüijŇæřRäyĭlèŁĆćĆz 8 āijă GPUă■āijŇŇ āĚś 32 GPU āzūēāŇēō■çzČăĀĆ

```

fleetrun --ips="xx.xx.xx.xx, yy.yy.yy.yy, aa.aa.aa.aa, bb.bb.bb.bb"
↪--gpus=0,1,2,3,4,5,6,7 example_collective.py

# worker_index: 0, step0 cost = 7.147776, speed: 34.481360
# worker_index: 0, step1 cost = 7.151375, speed: 408.405991
# worker_index: 0, step2 cost = 7.025396, speed: 509.624355
# worker_index: 0, step3 cost = 6.501647, speed: 533.641315
# worker_index: 0, step4 cost = 6.759287, speed: 520.999193
# worker_index: 0, step5 cost = 6.266363, speed: 536.729215
# worker_index: 0, step6 cost = 6.243353, speed: 522.510241
# worker_index: 0, step7 cost = 6.923586, speed: 519.478763
# worker_index: 0, step8 cost = 7.607512, speed: 534.919526
# worker_index: 0, step9 cost = 7.111218, speed: 508.371600

```


10.4 Fleet èó■çzČç■ŮçŤě

10.4.1 DistributedStrategy

Dist ribut- ed- Strat- egy	čsžāčēYēēōžāŮL	čsžāčēYēēōžāŮL	čsžāčēYēēōžāŮL
auto	bool	False	èĠlāLl āNŮæaEædūāRČæŤrāijYāNŮ
a_sync	bool	True	æŇĠçd'ž æŸrāRēā;ŁçŤlāijČæ■ēSGD èŁZēāNāRČ æŤræŽt' æŮrīijNāzĚāIJlPS erverælāāijRāy■çŤšæŤL
sync _nccl_allreduce	bool	True	æŇĠçd'žæŸr āRēāIJlæfRāyĹēĀZāŁçžŁ çlNāy■āy■ā;ŁçŤlāRŇæ■ē allre duceīijNāzĚāIJlColle ctiveælāāijRāy■çŤšæŤL īijNēĀZāyāIJlā;ŁçŤlāRŇ æ■ēallreduceāRŌçšž çžšçŽDāijĀēŤĀāijŽēZ■ā;Ō
nccl_comm_num	int	1	ncclēĀZāŁçāZlæŤrēĠR. ncclēĀZāŁçāZlæŤrēĠR nccl_comm_num āRfāzēāLāāfnGPUāžN éŮt'çŽDēĀZāŁçæŤLçŌĠīijN āžžēōōā■ŤæIJžēō;ç;ōāyž 1īijNād'ŽæIJžēō;ç;ōāyž2 āĀČēŚLāržCPUçžŁçlNæŤr num_threads īijNāžžēōōā■ŤæIJžēō;ç;ō āyž1īijNād'ŽæIJžēō;ç;ōāyž nccl_comm_num + 1
use_hierarchical_allreduce	bool	False	āLŁçžgāijRallred uceīijNāržāžŌād'ŽæIJžæĹā āijRīijNēŚLāržārRæŤræ■ō éĠRçŽDēĀZāŁçāīijNŖing AllReduc ēēĀZāŁçæŤLçŌĠā;ŌīijNēĠĠ çŤlHierarchical AllReduceāRf āžžēgčāEšēfēēŮōēēYāĀČ
hierarchical_allreduce_inter_ranks	int	1	āIJl āLŁçžgāijRallred eīijNā;ŌāsČžggroups āy■çŽD r ankæŤrāĀČāyĀēLŇç■L'āžŌ ā■ŤāyĹGPUēLČçČžāy■çŽD GPUæŤr
sync_batch_norm	bool	False	ēāŁçd'žæŸrāRēā;Ł çŤlāRŇæ■ēçŽDæL'žæ■čāLZ āNŮrīijNā■sāIJlēō■çzČēYū æōŁēĀZēŁGād'ŽāyĹēō;ād'Ġ āRŇæ■ēāĪGāĀijāŠNæŮžāūō āĀČā;ŠāL■çŽDāōđçŌrāy■ æŤræNāFP16ēō■çzČāŠŇC PUāĀČāžūāyŤçŽDāL■* āžĚæŤræNā*āžĚāIJl āyĀāRræIJžāZlāyĹēŁēZēāN āRŇæ■ēāijRæL'žæ■čāLZāĀČ
fuse_all_reduce_ops	bool	True	ēžYēōd' æČĚāEġāyNāijŽ āRēāRŇāyĀlayerāy■āRČ æŤrçŽDæčfāžççŽDAllR educeæš■ā;IJāRlāžūāLŖ āyĀāyīijNærŤāçCāržāžŌ fluid.layers.fc āy■æIJl' WeightāŠŇBi- asāy'd' āyĹāRČæŤrīijNæL'ŠāijĀēfē éĀL'ēāžāžNāRŌīijNāŌšæIJŇ ēIJĀēēĀāy'd'æñāAllRed uceæš■ā;IJīijNçŌfāIJlāRl çŤlāyĀæñāAllRe- duce æš■ā;IJāĀČ
fuse_ grad_size_in_MB	int	32	æfRāyĹAllReduceæš■ ā;IJçŽDæčfāžæ■ŮēLČæŤr
fuse_gradt _size_in_TFLOPS	int	20	æŇĠ āōŽæfRæñāAllReduc eæš■ā;IJçŽDæIJĀād'ġāsČæŤr īijNā■sāLŖē;ēfēāsČæŤr āRšēŁZēāNAllReduce
cudnn_exclusive_search	bool	True	ēāŁçd'žæŸr āRēā;ŁçŤlçl'ūāyĹæRlJçt'č æŮžæšŤæĹēēĀL'æNl'ā■ūçġr ēŌĹfæšŤāĀČāIJlcuDNNāy■ æIJl'āy'd'çġ■æRlJçt'čæŮžæšŤ īijNāRfāRŠāijRæRlJçt'čāŠŇ çl'ūāyĹæRlJçt'čāĀČçl'ūāyĹ æRlJçt'čārĹēŤæL'ĀæIJl'cu DNNçŌŮæšŤāžēēĀL'æNl'āĚū āy■æIJlāfŇçŽDçŌŮæšŤāĀČ æd'æŮžæšŤēĹdāyvēĀŮæŮū

10.4.2 BuildStrategy

Build-Strategy			
enable_sequential_execution	bool	False	<p>True if the operations should be executed sequentially. If False, the operations will be executed in parallel.</p>
fuse_elementwise_add_activation_ops	bool	False	<p>True if the elementwise add and activation operations should be fused. If False, the operations will be executed separately.</p>
fuse_batch_norm_activation_ops	bool	False	<p>True if the batch norm and activation operations should be fused. If False, the operations will be executed separately.</p>
fuse_relu_depthwise_conv	bool	False	<p>True if the relu and depthwise conv operations should be fused. If False, the operations will be executed separately.</p>
fuse_broadcast_ops	bool	False	<p>True if the broadcast operations should be fused. If False, the operations will be executed separately.</p>
fuse_all_optimizer_ops	bool	False	<p>True if all optimizer operations should be fused. If False, the operations will be executed separately.</p>
enable_inplace	bool	False	<p>True if the operations should be executed in-place. If False, the operations will be executed out-of-place.</p>
enable_backward_optimizer_dependencies	bool	True	<p>True if the backward pass should depend on the optimizer operations. If False, the backward pass will be executed independently.</p>
cache_runtime_context	bool	False	<p>unkown</p>

10.4.3 ExecutionStrategy

Execution-Strategy	cszãdŁŁYēōōāāŁ		
num_threads	1		<p>èāłçd'žā;ŠāL■ Executor çŽĐçžŁłŃæšā(thread pool)çŽĐād'gārŔ, æ■d'çžŁłŃæšāāŔŕçŦlæİēāžūāŔŠ æL'gēāŃprogramäy■çŽĐ operatorīijŁçōŮā■Ŕ īijŃēŁŔçōŮīijL'āĀĆāēĆāēIJ num_threads=1 īijŃāŁZæL'ĀæIJL' çŽĐoperatorāŔEäyĀ äyŁæŌēäyĀäyŁāIJŔæL'gēāŃ īijŃā;EāIJläy■āŔŃçŽĐpro graméG■ād'■āŚlæIJ§(it era-tions)äy■æL'gēāŃ ēāžāžŔāŔŕēČ;äy■āŔŃāĀĆ</p>
num_iteration_per_drop_scope	10		<p>èŕēēĀL'ēāžēāł çd'žēŮŕ'ēŽŦād'ŽārŠæñāēŁ■ äžčāžŃāŔŌæyĒçŔŔēäyĀæñā äyt'æŮūāŔŸéĠŔāĀĆæĪāādŃ ēŁŔēāŃēŁGłŃāy■īijŃçŦš æŁŔçŽĐäy■ēŮŕ'äyt'æŮūāŔŸ éĠŔāŔEēēŃæŦ;āŁŕlocal execution scopeäy■īijŃäyžāžE ēAŁāĒ■āržäyt'æŮūāŔŸéĠŔ ēēŚçžAçŽĐçŦšēŕūäyŌéĠŁ æŦ;īijŃēĀŽäyŷārEāĒūēō; äyžē;Čād'gçŽĐāĀīijLæŕŦ āçČ10æLŮēĀĒ100īijL'āĀĆ</p>
num_iteration_per_run	3	int	<p>āōČēĒ■ç;ōāžEā;ŠçŦlæŁū āIJlpythonēĐŽæIJñäy■ èŕČçŦlpe.run()æŮū æL'gēāŃāŽlāijŽæL'gēāŃçŽĐ èŁ■āžčæñāæŦŕāĀĆExecu toræŕŔæñāēŕČçŦlīijŃāijŽ èŁŽēāŃnum_iterat ion_per_runæñāēŌ■ çžČīijŃāōČāijŽā;ŁæŦŕ'ä;Š æL'gēāŃēŁGłŃæŽŕ'āŁŃāĀĆ</p>
use_thread_barrier	bool	False	<p>ā;Šā;ŁçŦŦ PServer æĪāāijŔæŮūäyž True</p>

ẽĠłŁłæũũăŘĹčš; ăžęçžČăŁăéĂşăĹĒăŸČăĭjŘěő■čžČ

11.1 čŎĂăžŇ

ăĬĹă;ŁčŤĹăŤŕă■őăžũëăŇăĹĒăŸČăĭjŘěő■čžČžĎăŘŇăŮũ,
 æĹŚăžňëŁŸăŘŕăžëăĭjŤăĚëëĠăĹăũũăŘĹčš;ăžę(Auto Mixed Precision)
 æĹëëŁžăŸĂă■ëăŘŕă■Ġëő■čžČžĎéĂşăžę.

ăŸžăŤĂčžĎčëđčžŘč;ŚčžĬăĹăđŇéĂžăŸŸă;ŁčŤĹă■Ťčš;ăžę single-precision
 (FP32) æŤŕă■őăăĭjăĭjŘăĹëă■ŸăĹăĹăđŇăŘČăŤŕăĂăëŁžăăŇëő■čžČăŖŇéčĎăŤŇ.
 ăĬĹăŸŁëŁŕčŎŕëĹČăŸă;ŁčŤĹă■Łčš;ăžę half-precision (FP16) æĹëăžčăŽăă■Ťčš;ăžę.
 ăŘŕăžëăŸęăĹëăžëăŸŇăë;ăđ'Ď:

1. ăĠŔăŕŚăŕžGPU memory čžĎéĬĂăśČ: GPU æŸ;ă■ŸăŸăăŕŸăČĚăĒăŸăŸŇ,
 æŤŕăŇĂăŽŤ'ăđ'ġăĹăđŇ / batch size
2. éŽă;ŎăŸ;ă■ŸŕžăĒžăŮũčžĎăŸęăŏ;ăŎŇăĹž
3. ăĹăéĂşGPU æŤŕă■ëëŁŕčŎŮéĂşăžę (éĬĂëëĂGPU æŤŕăŇĂ[1])
4. GPUăŸŁ FP16 ăŘđăŘŕăŸŕFP32 čžĎ 2 - 8 ăĂ■[2]

Paddle æŤŕăŇĂëĠăĹăũũăŘĹčš;ăžęëőăçőŮ, ăžũăőđčŎŕăžĒ
 èĠăĹĹčžŤ' æĹđ' FP32 ăĂăFP16ăŘČăŤŕăĹ' ŕăĬŇ, Dynamic
 loss scaling, opéžŚčž;ăŘăă■Ť ç■Łč■ŮčŤăĹëéĂăĒă ăžă FP16
 ăĹăăĂăëŇČăžŤ'ë;ČăŕŕëĂŇăŸęăĹëčžĎăĹăđŇăĬĂčžĹčš;ăžęă■şăđ' şăĂČ Fleet
 ă;ĬăŸžPaddleéĂžčŤĹčžĎăĹăŸČăĭjŘěő■čžČĂĬăŕŕă;ŽăžĒčŎĂă■ŤăŸşčŤĹčžĎăŎëăŕč,
 čŤĹăĹăăĹăĬĂëëĂăũăăĹăăĠăăŸăăžčăĂ ăŕşăŕŕăŕĒëĠăĹăũũăŘĹčš;ăžęăžŤčŤĹăĹăŎşăĬĹčžĎăĹăŸČă

ăŸŇăŮĠăŕĒéĂžëŁĠăŸăŸăŸčŎĂă■Ťă;Ňă■ŕăžŇčžăăëČăëČă;ŤéĂžëŁĠ
 FleetăŕĒăőđčŎŕăũũăŘĹčš;ăžęčžĎăĹăŸČăĭjŘěő■čžČ, ăŕĒăđ'ŮčžžăĠžăĹŚăžňă;ŁčŤĹ
 Fleet ëŁžăăŇăŕŇă■ëëő■čžČăĹăéĂşčžĎăđëũăĂČ

11.1.1 èŕŦélŊæŦŁædIJ

çŖŕŕŕČ: 4 æIJž 32â■a V100-32GB

imagenet	â■Ŧâ■a batch size	éĀšāžę img/s	top1
‘VGG16-FP32	32	4133	55.4%
‘VGG16-AMP	32	7238	54.6%

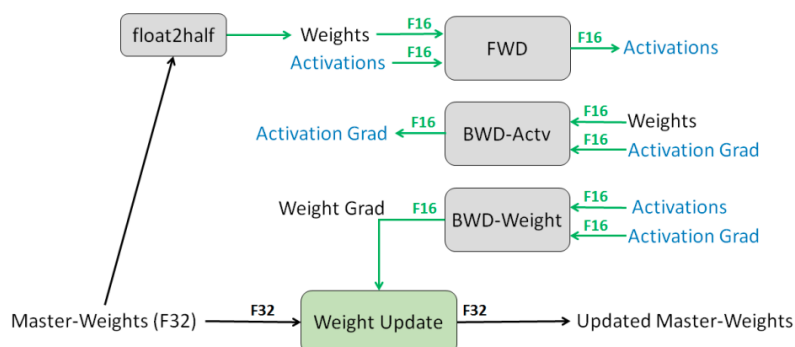
imagenet	â■Ŧâ■a batch size	éĀšāžę img/s	top1
‘Resnet50-FP32	128	8410	76.3%
‘Resnet50-AMP	128	25591	76.0%
‘Resnet50-FP32	256	OOM	OOM
‘Resnet50-AMP	256	29440	76.0%

11.2 AMP âĖnéĀšâijĀğŦ

èĖŽéĖŊžēâIJĀ■ŦæIJžâd’Žâ■äÿŁēő■çzČResnet50 äÿžçōĀâ■ŦäĴŊâ■ŖžâžŊçz■Fleet äÿ■AMPçŽĎÇŦĹæšŦ.

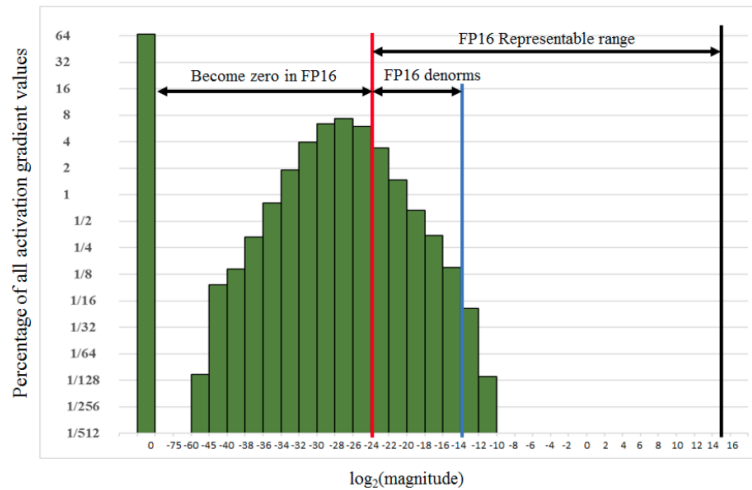
11.2.1 èĖĹâĹĹæuâĹĹçšĴâžęâŖšçŖĖ

FP32 âŖČæŦŕâĹŕæIJŋâŖĹæŽŦæŦŕ



âçCäÿŁâŽĴæĹĀçd’ž, âIJAMP äÿ■, æĹââĎŊâŖČæŦŕ weight ,
 âĹ■âŖŠäÿ■éŦŦçŽĎÇzšædIJactivation, âŖ■âŖŠçŽĎgradient éČĴäžèFP16
 âĴçâijŖâ■ŦŦâĴĴ, çŦŦsæ■d’âŖŕäžēâĖŖâŖšæĹââĎŊâ■çŦĴçŽĎæŦĴâ■ŦçĴŦŦijŊâŖŊæŦŦæŖŖŖénŦēōaçōŦâšŊ
 PaddleæaEæĎūâijŽäÿžæŦŖäÿĀäÿĴweight çžŦæĹd’äÿĀäÿĴFP32âĹŕæIJŋ,
 çŦĹâžŖŖČæŦŕæŽŦæŦŕ.

Loss scaling



æÇÄyŁāZŁæL'Äçd'ž, åóðéŽĚæČĚåEŁäy■æŁāāđNëō■çZČäy■çŽDæšŘāžZāRÝéĠR,
æŕTāēČgrad (çL'žāLñæYŕ activation çŽD grad), āŕŕēČāijZāZāārRāžŌ
FP16çŽDçšŁāžēä;ŌēĀNāRÝæLŔ0;

āŕēäyĀæŰžēÍčāIJFP16 çŽDēāŁçd'žēNČāŽt' çŽDäy■æIJL'āŁLād'ğçŽDäyĀéČlāLE(āžŌæIJĀād'ğāĀijā;Ā
ā■t'æšqæIJL'ēčnāLl'çTlāLŕ.

āržgradient āAŽäyĀäyŁæTŕ'ä;šçŽDæTŁād'ğ, ēČlād'šæŽt'āĒĒāLEçŽDāLl'çTlāFP16
çŽDēāŁçd'žēNČāŽt'.

Fleet AMP āijZāIJlāR■āŔšāijĀāğNāL'■ārž loss ēŁZēāŃ up scaling,
āžŰāIJlāL'ğēāNāžzā;TæčŕāžççŽyāĒšæš■āIJ(e.g. gradient-clip, update) āžNāL'■ārž gre-
dient ēŁZēāŃ down scaling æAčād'■āŌšælēçŽDād'ğārŔ.

scaling factor çŽDēōŁçŁōæYŕ Lossing scaling çŽDāĒšéTō, Fleet AMP æŔŔä;Ž
Dynamic loss scaling iijLēzYēōđ'ijl' āšŃ Constant loss scaling äyđ'çğ■s-
caling ç■ŰçTē:

- Constant loss scaling: ēōŁçŁō use_dynamic_loss_scaling = False āšŃ
init_loss_scaling (float)
- Dynamic loss scaling: scaling äy■ēlčäyŕ'çŽDēŰōēčYæYŕā;šscaling
up äy■ēūšæŰū, āž■āijZæIJL'ēČlāLEēŁČārRāRÝéĠRāijZēčnēāŁçd'žæLŔ
ŌēĀNā■šād'šçšŁāžē; ā;šscaling up ēŁGāžēæŰū,
āRÝéĠRēūĒēŁGFP16ēāŁçd'žēNČāŽt' āGžçŎŕ nan or inf,
āŔNæūēĀāæLŔçšŁāžēæ■šād'š. ænd'ç■ŰçTēēĠGçTlēĠlāLl' gradient
āĀijæčĀætNçŽDæŰžāijŔ:

– ā;šēŁđçz■incr_every_n_steps(int) äyŁbatch äy■æL'ĀæIJL'çŽD-
gradient ēČlāIJFP16 çŽDēāŁçd'žēNČāŽt', āŕEscaling factor
āčđād'ğincr_ratio(float) āĀ■;

– ā;šæIJL'ēŁđçz■decr_every_n_nan_or_inf(int) äyŁbatch äy■gradient
ēĠNāGžçŎŕ nan / infæŰū, scaling factor çijl'ārŔ decr_ratio(float) āĀ■.

– äyŁēčŕāZŽäyŁāŔČæTŕFleet æŔŔä;ŽçŽDēzYēōđ'āĀijāŔŕāžžæzæūšçzĪād'ğēČlāLEēŁæAæšĆ,

çTlæLûéĀŽāyŷāy■éIJĀèçAāfōæTz.

æÇäyNāZ;æL'Āçd'žāIJÍ Dynamic loss scaling äy■ijNæaEæđūāIJlæfRāyĀäyĭ iteration
éÇ;āijŽā;Iæ■ōā;ŠāL■ gradients æYrāRçāGžçŌr nan or inf èfYæIJL'çTlæLûèð;ç;ōçŽD Dy-
namic loss scaling āRÇæTŕæIēāLlæĀĀèrÇæTŕ' loss scaling factor çŽDād'gārRijNārEgradient
ār;éGRāfIæNĀāIJÍ FP16 çŽDēāIçd'žèNČāZŕ'āzNāEĒāĀĆ



OP ézŠçŽ;āR■ā■T

æIqādNāy■çŽDæŠRāžZOperation (OP) āRfèÇ;āržçš;āžçè;ÇäyžæTŕæDš,
äyžāEçāōāIAMP äy■çš;āžçæUāæ■\$, āRfäzēēĀŽèfĠOP ézŠçŽ;āR■ā■TāržāEūā;ŠOP
æŠ■ā;IJçŽDçš;āžçæĀŽæNĠāōŽ.

- çŽ;āR■ā■T: OP æŠ■ā;IJāIJÍFP16çš;āžçæyNèfZèāN, input: æÇædIJäy■æYŕFP16
āijŽècñéŰāĒLcast æLŔFP16āRŌāE■è;ŠāĒĒOP. output: FP16
- ézŠāR■ā■T: OP æŠ■ā;IJāIJÍFP32çš;āžçæyNèfZèāN, input: æÇædIJäy■æYŕFP32
āijŽècñéŰāĒLcast æLŔFP32āRŌāE■è;ŠāĒĒOP. output: FP32
- çAŕāR■ā■T: æL'ĀæIJL'äy■āIJlézŠæLŰçŽ;āR■ā■TéGŦçŽDOP. āzĒā;ŠOP æL'ĀæIJL'
inputs éÇ;æYŕ FP16çš;āžçæUū, æŠ■ā;IJæL'■āIJÍFP16çš;āžçæyNèfZèāN, āRççIĀāžèFP
32èfZèāN. input / output: āŠNāŌšāgNè;ŠāĒĒäy■çŽDæIJĀénYçš;āžççŽyāRŦ

Fleet āušçžRécDèð;āžEäyĀäyĭèÇ;ād'šèEççŽŰçziād'gād'ŽæTŕæIqādNŌPçŽDézŠçŽ;āR■ā■T,
éĀŽāyŷæÇĒāEĭäyNçTlæLûāzūāy■éIJĀèçAāfōæTz, ā;EæYŕæÇædIJāzžāLāāržçš;āžçæIJL'çL'žæōLèçAæśĆ,
æLŰèĀĒäyNāIJZæŰŕāçðèĠāōŽāZL' OP, çTlæLûāRfäzēēĀŽèfĠ
paddle.distributed.fleet.DistributedStrategy.amp_configs äy■çŽD
custom_white_list āŠŦ custom_black_list èfZèāNæNĠāōŽ. āRŦæYŕ,
çTlæLûèfYāRfäzēēĀŽèfĠcustom_black_varnames, æIēāEūā;ŠæNĠāōŽPaddle
program æŠRāyĀäyĭ varāfĒēāzā;ççTÍFP32çš;āžç.

æLŠāznārEāIJlæŰĠæIJnçŽD appendixäy■ èfZäyĀæ■ēāzNçz■ Fleet
çŽDézŠçŽ;āR■ā■Tèð;ç;ōāRĒāĒūā;šāŠ■āĀĆ

11.2.2 āijĀāgNèō■çžČ

æūzāLāā;IèŰ

éçŰāĒLæLŠāznèçAārijaĒēā;IèŰāŠŦāōŽāZL'æIqādNāŠŦ data loader,
èfZäyĀæ■ēāŠŦFleet äyNāĒūāzŰāzžāLāāšžæIJnāyĀèĠt'.

```
import os
import fleetx as X
import paddle
import paddle.fluid as fluid
import paddle.distributed.fleet.base.role_maker as role_maker
import time
import paddle.distributed.fleet as fleet
```

Initial configuration

```
paddle.enable_static()
configs = X.parse_train_configs()
fleet.init(is_collective=True)
```

Model and data loader

```
model = X.applications.Resnet50()
downloader = X.utils.Downloader()
local_path = downloader.download_from_bos(
    fs_yaml='https://fleet.bj.bcebos.com/test/loader/small_imagenet.
    ↪yaml',
    local_path='./data')
batch_size = 32
loader = model.get_train_dataloader(local_path, batch_size=batch_
    ↪size)
```

AMP configuration

For more details, see [Loss scaling](#) in the [FleetX documentation](#).

Fleet AMP is a meta optimizer, which is used to wrap the inner optimizer. Fleet AMP uses the `paddle.optimizer.Adam` as the inner optimizer.

```
dist_strategy.amp = True
dist_strategy.amp_configs = {
    "init_loss_scaling": 32768,
    "decr_every_n_nan_or_inf": 2,
    "incr_every_n_steps": 1000,
    "incr_ratio": 2.0,
    "use_dynamic_loss_scaling": True,
    "decr_ratio": 0.5,
    "custom_white_list": [],
    "custom_black_list": [],
```

([FleetX documentation](#))

(çz■äyŁéął)

```
}

optimizer = fluid.optimizer.Momentum(learning_rate=0.01, momentum=0.
↪9)
optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)
```

ajĂăÑě■czĈ

èƒZäyĂéČláLEaŠŇFleet äy■ăĚűäzŰäzzaŁaą\$zæIJñçŻyăŘŇ:

```
place = fluid.CUDAPlace(int(os.environ.get('FLAGS_selected_gpus', 0)))
exe = fluid.Executor(place)
exe.run(fluid.default_startup_program())

for i, data in enumerate(loader()):
    start_time = time.time()
    cost_val = exe.run(model.main_prog,
                        feed=data,
                        fetch_list=[model.loss.name])

    end_time = time.time()
    print(
        "worker_index: %d, step%d cost = %f, speed: %f"
        % (fleet.worker_index(), i, cost_val[0], batch_size / (end_time - start_time)))
```

11.2.3 èŁŘàÑèő■čŻĈĎŽæłJň

äÿĂèàŇăŘřăĹĺă■TæIJžăd'Žă■ąăĹĒăÿČăijRěó■čzČiijŽ

```
fleetrn --gpus 0,1,2,3,4,5,6,7 --log_dir log example_amp.py
```

```
# worker_index: 0, step0 cost = 6.895311, speed: 12.192901
# worker_index: 0, step1 cost = 6.964077, speed: 412.116618
# worker_index: 0, step2 cost = 7.049311, speed: 433.850506
# worker_index: 0, step3 cost = 7.006689, speed: 358.400410
# worker_index: 0, step4 cost = 7.000206, speed: 398.210745
# worker_index: 0, step5 cost = 7.088611, speed: 462.322357
# worker_index: 0, step6 cost = 7.022367, speed: 425.185013
```

11.2.4 Fleet 0.1.0.beta

äyLæŮĜçöÄèçAäzŊçz■äzEFleet äy■ézŚçŻ;āŖ■ā■TçŽD API æŌëāŖcīijŊ
 äyŊæŮĜārEèfZäyĀæ■äzŊçz■Fleet äy■ézŚçŻ;āŖ■ā■TçŽDāōđçŌŕāŖŊāŖēç;ārēzēō■çzČēĀāēĻŖā;śāŖ■āĀ
 çŽōāĻ■ Fleet äy■ AMP çŽDēzŸēōd'ēzŚçŻ;āŖ■ā■TāçCāyŊīijŊ āĒūāzŮæIJāĻŮāĜçŽD op
 ēç;āsđāžŌçAŕāŖ■ā■TīijŽ

```
white_list = {
    'conv2d',
    'matmul',
    'mul',
}
black_list = {
    'exp',
    'square',
    'log',
    'mean',
    'sum',
    'cos_sim',
    'softmax',
    'softmax_with_cross_entropy',
    'sigmoid_cross_entropy_with_logits',
    'cross_entropy',
    'cross_entropy2',
}
```

0.1.0.beta

çŽ;āŖ■ā■Täy■āŖtæIJL'ā■ūçğŕāŖŊāzŸæŖTēŖçŌŮīijŊNēfZæūçŽDēōç;ç;ōēç;ād'şæzæūşād'ģēČlāĻEçŽL
 CV āIJzæŖçŽDēāāđŊāĻæĀŖīijLVgāĀĀResNetīijL'īijŊ āZāyŷā■ūçğŕēōāçŌŮā■āæ■ōēfZāzZāēāāđŊēōāç
 āĒūāzŮ ops çŽDāijĀēŤĀāŖtā■āā;ĻārŖāyĀēČlāĻEāĀČ āŕzāžŌ äyēèçAāijĀēŤĀāIJĻ RNN
 ēōāçŌŮçŽD NLP ælāāđŊīijŊçŽōāĻ■çŽD AMP āōđçŌŕāŖŖēĀŖāzūāy■æŸŕāĻæŸŌæŸĻāĀČ

ézŖāŖ■ā■Täy■çŽD op āŖŕāzēāĻEäyž3 ād'ğçşīijŽ *
 āŕzçş;āzēēĻdāyŷæŤŖæĐŖçŽD opīijŽ softmaxīijŊcross_entropy ç■ĻāĀČ
 * èç;ŖāĜççŽyārzāžŌèç;ŖāĒēæIJL'æZt'ād'ğāĻāĀĀēŊČāZt'çŽDopīijĻf(x) >>
 xīijL'īijŽexpīijŊsquare, log ç■ĻāĀČ * reduce çşzādŊçŽDopīijŽ meanīijŊsum ç■ĻāĀČ
 æĻĀāzēīijŊŊŤlāĻŮāyŊæIJZāĻd'æŮ■æŮŕçŽDēĜlāōžāzĻopæŸŕāŖēIJĀēçAāĻāāĒēēzŖāŖ■ā■TæŮīijŊŖāŖ

ēIJĀēçAæŖlæĐŖ: äyĀāzZāyŷçŤlçŽD op āēç BatchNormīijŊ
 poolingīijŊ relu āsđāžŌçAŕāŖ■ā■TīijŊNēfZæĐŖāŖŖçlĀēfZāzZ
 op çŽDæŤŖæ■ōçşzādŊĀEŖāōžāzŌāzŊāĻ■çŽD op çŽDçşzādŊīijŽ
 āŖēād'ŮāzūēāŊāĻEäyČāijŖēōāçŌŮā;ççŤĻ AMPāzŖāŖŖŌīijŊgradient-allreduce æŸŕāIJĻFP16
 äy■ēfZēāŊçŽDāĀČ

èĠlāLlāNŪop æRŠāĒĒ

āIJlēō■czČāijĀāgNāL■īijNæaEædūāijŽæāzæ■ōézSçŽ;āR■ā■TāIJlāL■āRŠāSŃāR■āRŠç;SçzIJèĠlāLlāæF
cast opīijN āēCīijŽ * āL■āRŠāy■æRŠāĒĒ FP32toFP16 castīijN ārE FP32 çŽDlayer param-
eter āL'raēIJn cast æLR FP16īijN èfŽēāN FP16 conv èoāçōŪāĀĆ * āR■āRŠāy■æRŠāĒĒ
FP16toFP32 castīijN ārEç■L'ālŕçŽD FP16 gradient cast æLR FP32īijN çDūāRŌæŽt'æŪr
FP32 çŽDparameter āL'raēIJnāĀĆ

cast op èŽ;çDūāijŽāyæælēēclād'ŪçŽDāijĀéTāīijN ā;EæYrāIJlèryāēĆ Vg-
gāĀAResNet ç■L'āyžēēAçT'séĠād'■çŽD conv layer āyšēāNçŽDèĀNæLR CV
ælaādNāy■īijN āRlēIJĀēēAcast input āšN ærRāyĀāsCçŽDparamīijNāzūāy■ēIJĀēēA-
cast ælaādNçŽDāy■ēŪt'çzSædIJīijNèfŽæāū cast æS■ā;IJāyæælēçŽDāijĀéTĀē;ČārS,
āōzæYšāĀ■ā■Lçš;āžēēōāçōŪāyæælēçŽDāLāéĀšēēEçŽŪīijZā;EæYrāēCædIJælaādNçŽDāyšēāN
layers āžRāLŪāy■ā■YāIJlē;Čād'ŽçŽDēzSāR■ā■T opīijLe.g. conv -> log -> conv -> square -
> convīijL'īijN èfŽæāūælaādNçŽDāy■ēŪt'çzSædIJēIJĀēēAèfŽēāNād'Žæñā FP32toFP16 āšN
FP16toFP32 castīijN cast āijĀéTĀāRēāijZæĀēāL'gācdād'gīijNāzŌēĀNæLtæūLā■Lçš;āžēāyæælēçŽDāLāéĀš

āRrēČ;āy■ēĀCçTÍ AMP āLāēĀšçŽDæČĒāEġ

- RNN āyžāyžçŽD NLP ælaādN
- ælaādNçzDç;Sāy■æIJL'è;Čād'ŽézSāR■ā■T op çŽDælaādN
- ārZæTŕæ■ōçš;āžēēTŕæDšçŽDāzZāLāīijLAdversarial Attacking in MLīijL'

āŽ;āČR Input Layout æāijāijR

CV ælaādNēō■czČæŪūāzEē;āL'raēIJĀā;šēĀšāžēīijNāy■āRŃāIJZæŽrāyNæŌlē■Rā;ççTlāy■āRŃāZ;āČF
LayoutīijŽ

- FP32īijŽNCHW
- èĠlāLlāæūāRĻçš;āžēīijŽ NHWC

```
# when build dataloader
loader = model.load_imagenet_from_file("./ImageNet/train.txt",
                                         batch_size=args.batch_size,
                                         data_layout="NHWC")

# when build model
if data_format == "NHWC":
    img_shape = [None, 224, 224, 3]
else:
    img_shape = [None, 3, 224, 224]
image = fluid.data( name="feed_image", shape=img_shape, dtype=
    ↪ "float32", lod_level=0)
conv = fluid.layers.conv2d(input=input, data_format= "NHWC")
```

11.3 æŒíë■ŘéŸĚèřž:

æĆæđIJéIJĚèAåržèĜlāŁlæuūāŘŁçš;āžæAŽāóŽāLúāŃŮäŁæŤž,æLŮæŽt' æůsāĚěčŘĚèğčAMPäy■āŒ

- Mixed Precision Training
- MIXED PRECISION TRAINING: THEORY AND PRACTICE
- Training With Mixed Precision

CHAPTER 12

Batch Normalization

12.1 Batch Normalization + Strategy

Batch Normalization is a technique used to normalize the inputs of each layer in a deep neural network. It helps to stabilize the learning process and allows for higher learning rates.

- It is applied to the input of each layer, except for the output layer.
- It involves calculating the mean and variance of the input for each layer, and then normalizing the input by subtracting the mean and dividing by the standard deviation.
- It also involves adding a small constant to the variance to avoid division by zero.

Batch Normalization is applied to the input of each layer, except for the output layer. It involves calculating the mean and variance of the input for each layer, and then normalizing the input by subtracting the mean and dividing by the standard deviation. It also involves adding a small constant to the variance to avoid division by zero.

- **Forward** **Recomputation** **Backpropagation**
The forward pass involves calculating the mean and variance of the input for each layer, and then normalizing the input. The backward pass involves calculating the gradients of the loss with respect to the input of each layer, and then propagating these gradients back through the network.
- **Gradient Merge** The gradient merge step involves combining the gradients from the forward and backward passes, and then applying them to the input of each layer.

12.2 Batch Normalization

12.2.1 Forward Recomputation Backpropagation

Batch Normalization is applied to the input of each layer, except for the output layer. It involves calculating the mean and variance of the input for each layer, and then normalizing the input by subtracting the mean and dividing by the standard deviation. It also involves adding a small constant to the variance to avoid division by zero.

- **Batch Normalization** The batch normalization step involves calculating the mean and variance of the input for each layer, and then normalizing the input by subtracting the mean and dividing by the standard deviation. It also involves adding a small constant to the variance to avoid division by zero.

- **aRnaRSèoaçõUñjZ** èfRèaÑaRnaRSçõUaRaeIèèoaçõUaRCætř(Parameter)çZDæcřažæĀĈ
- **aijYaNŮñjZ** ažTçTlāijYaNŮçõUașTāzèæZt æŮrāRCætřāAij āĀĈ

āIJlāLnaRSèoaçõUèfGçlNāyñijNāLnaRSçõUaRāijZèoaçõUaGžād' gēGRçZDāyēŮt' çZšædIJñjNçTš: SizeèŮLād' gñijNāyēŮt' çZšædIJāāçTlçZDāEĒāYāzšārseŮLād' gāĀĈéçðæaIæyāfĈæaEæđūaijZā;ŁçTl VariableāIēāYāClēfZāzZēZŖāsČZDāyēŮt' çZšædIJāĀĈā;ŠælaādNāsČætřāLāæuŝæŮñijNāEŮāyēŮt' çZ āāæōād' gēGRçZDāEĒāYāĀĈèZ;çDūéçðæaIæyāfĈæaEæđūçZDāY;āYāZðætŮæIJāLūaijZāRLæŮūæ ā;EæYřæIJLāzZāyēŮt' çZšædIJæYřāRnaRSèoaçõUèfGçlNāyçõUaRçZDè;ŠāĒēñjNēfZāzZāyēŮt' çZšæ

ārzažŌād' gārRāZžāōZçZDāEĒāYāIèèrt'ñijNāeČædIJçTlāLūāyNāIJZā;ŁçTlād' gBatch SizeçZDætřæōèfZèaÑèōçZČñjNāLZārEāñjēGt āTāyāyēŮt' çZšædIJāāçTlāEĒāYāčðād' gñijNéČcāzL

FRBæYřāEæuŝažæāēāzāç;ŠçzIJāLĠāLEāyžkāyIēČlāLEñjLsegmentsñijLāĀĈārzařRāyIsegmentèĀNēl

æLSāznæLāLĠāLEç;ŠçzIJçZDāRYēGRāRnāAžcheckpointsāĀĈ éČcāzLēŮōēçYāIēāzEñijNāeČā;TēĀL'æNl' checkpointsāŚñijšèĠāzŌFRBæŮzæșTāRŘāGžāzæIēñjNād' gē æLSāznçšēēAșæuŝažæāēāzāç;ŠçzIJēĀZāyāYāYřçTšāyĀāyIāyIāIŮāyšēĀTā;ŮāLřçZDñjNærTāeČResNet 50çTš16āyIblockāyšēĀTēĀNāLRñijN Bert-LargeçTš24āyItransformerāyšēĀTēĀNāLRñijNāzēāyđ' āyIāRēIā ārzažŌēIđāyšēĀTçZDç;ŠçzIJñjLærTāeČāRnæIJL'ād' gēGRshortcutçZšædDçZDç;ŠçzIJñjLñjNFRBāzšæTřæ āRlæYřāRfēČ;ād' ZēĀŮet' zāyAçČzāEĒāYñjLçTlāzŌāYāČIshortcutçZDVariableñijLāĀĈ

12.2.2 Gradient Merge

āyŌFRBçZyærTñijNGradient MergeāzūæšæIJL'āČRFR- BéČcæāuārzaEĒāYçZDā;ŁçTlāAžāGžād' gāLĀēYtæŮgēLñçZDæTžāLñijNāRlæYřāIJlèōçZČætAçlNāyLā SizeèōçZČætLædIJçZDçZōçZDāĀĈāEŮā;ŠæIèèrt'ñijNārsæYřā;ŁçTlēNēāzšāŌšæIJL'ād' gārRçZDBatchætř ç;ŠçzIJèoaçõUā;ŮāLřæcřažæĀĈāEŮēŮt' āijZæIJL'āyĀēČlāLEæY;āY/āEĒāYçTlāzŌāYæT;æcřažēñjNçTl

GradientMerge çŮçTēāIJlā;ŁçTlæŮzéIcāzšā;ŁçōĀāTñijNçTlāLūāRlēIJĀēēAāōZāzL'ārEād' ZārŠBatch

12.3 æŠā;IJāóðèùt

èřèçnáèLČāyæLSāznārEāšzāzŌBERTæIāādNçZDāōđçTlāæuā;NñijNāLEāLlānzèfZāyđ' āyIāčðād' gBat

- æužāLāèōçZČèDŽæIJnēfRèaÑæL'ĀāfEēāzçZDā;IèŮŮāNĒāĀĈ
- āōZāzL'āLEāyČāijRæIāaijRāzūāLIāgNāNŮāĀĈ
- āLāè;IāIāādNāRLæTřæōāĀĈ
- āōZāzL'èōçZČçŮçTēāŠñāijYaNŮāZñijNāIJlēfZāyĀæææLSāznārRāzēēĀL'æNl'ā;ŁçTlFRBæLŮēĀ MergeçŮçTēāIēāčðād' gBatchSizeāĀĈ

āyNēIcæLSāznæIēāLEāLlāzNçzFRBāŠNGradient Mergeāyđ' çgçŮçTēāL'ĀārzažTēDŽæIJñçZDçijŮ āRL bert_gradient_merge.py)āĀĈ

12.3.1 Forward Recomputation Backpropagation

æûzâŁäçlèŧŮ

éçŮăĚĹăĹŚăznéĪĂèçĀæûzâŁăèő■çzČăy■ăĹĀçŤĹăĹŕçŽĐpythonăĹăĹŮĹĹĹfleetx
ăĹŕăzèçŤĹăžŮăĹăè;;ăĹŚăznăyžçŤĹăĹăŕĂèçĚĐăŮăŕčăçĈĹĹžăĹăè;;ăĹăăđŊăŕĹăĹŤŕă■ōĹĹŊăĹăăđŊăő
distributed.fleet äy■ăőžăžĹ'ăžĒăyŕăŕŊçŽĐăĹăyČăĹĹŕç■ŮçŤăăçŽçŤĹăĹă;ŕçŤĹăĂĈ

```
# -*- coding: UTF-8 -*-
import paddle
import fleetx as X
import paddle.fluid as fluid
import paddle.distributed.fleet as fleet
```

ăőžăžĹ'ăĹĒăyČăĹĹŕăĹăĹĹŕăžŮăĹĹăğŊăŊŮ

éĂžèĹGX.parse_train_configs() æŮăŕčăĹĹŊçŤĹăĹăŕăžèăőžăžĹ'èő■çzČçŽăĚşçŽĐăŕčăĹŤ
init() æŮăŕčăőžăžĹ'ăžĒăĹĒăyČăĹĹŕăĹăăđŊĹĹŊăyŊăĹăžççăĂăy■çŽĐis_collective=TrueăĹçđ'ž

```
paddle.enable_static()
configs = X.parse_train_configs()
fleet.init(is_collective=True)
```

ăĹăè;;ăĹăăđŊăŕĹăĹŤŕă■ő

çŤĹăĹăŕăžèéĂžèĹGX.applicationsæŮăŕčăĹăè;;ăĹŚăznéçĐăĚĹăőžăžĹ'ăè;çŽĐăĹăăđŊĹĹŊăç
SizeĹĹĹ130ĹĹĹ'ăĹèèçŽăăŊăő■çzČăĂĈ

ăyŮă■đ'ăŕŊăŮŮĹĹŊçŤĹăĹăŕăžèă;ŕçŤĹăĹŚăžŋçŽĐ'Downloader'æŮăŕčăyŊă;;éçĐăĚĹăĹă■ŮçŽĐ

```
model = X.applications.BertLarge()
downloader = X.utils.Downloader()
local_path = downloader.download_from_bos(
    fs_yaml='https://fleet.bj.bcebos.com/small_datasets/yaml_
    ↪example/wiki_cn.yaml',
    local_path='./data')
data_loader = model.get_train_data_loader(
    local_path,
    max_seq_len=512,
    batch_size=130,
)
```

ăőžăžĹ'Recompute Strategy ăŕĹ Optimizer

æŮăyŊăĹăăĹŚăznăŕăŕăžèăőžăžĹ'ăĹĒăyČăĹĹŕăő■çzČăy■ăĹĀăžŤçŤĹăĹŕçŽĐç■ŮçŤăăçĒăĂĈăyŊăĹă
recomputeèőç;ŕăyžTrue äžŮèőç;ŕăĹŚăznăžŊăĚĹăőžăžĹ'ăè;çŽĐcheckpointsăĂĈ

æŮăyŊăĹăèçŤĹăĹăĹĪĂèçĀăőžăžĹ'èő■çzČăy■ăĹŤ æŮŕăĹăăđŊăĹĀçŤĹăĹŕçŽĐăĹŮăŊăžĹĹŊăžŮă;ŕç
distributed_optimizeræŮăŕčăŕĒăĹŮăŊăžĹ;Ŋă■çăyžăĹĒăyČăĹĹŕăĹăĹĹŕăĂĈ

```
optimizer.minimize(model.loss)
```

```
dist_strategy = fleet.DistributedStrategy()
# ;fçTlRecomputeijñzûèøç;öcheckpoints
dist_strategy.recompute = True
dist_strategy.recompute_configs = {"checkpoints": model.checkpoints}

optimizer = fluid.optimizer.Adam(learning_rate=configs.lr)
optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)
```

ÄgNèøçZ

FleetX äyñijNæLSäznäyžçTlæLüæRŘä;ZäžEX.MultiGPUSigner æŌæRcñijNçTlæžŌGPUaLæyÇaijRèøçZÇäÄÇäEüäy■model aRL data_loader aLæLnažyžçññäzNæ■äy■aLæè;çZDælaadNæRLæTæ■öaÄÇstart_step èa|çd'žaijÄgNæLŠ■rèøçZÇlogçZDæ■æTñijNèNèçTlæLüæÇsad'■çŌæLSäznçZDælaadNèøçZÇæŠäžç

```
trainer = X.MultiGPUSigner()
trainer.fit(model, data_loader, epoch=10)
```

èŒRèaÑèøçZÇèDŽæIJñ

aoNæLRèDŽæIJñçZDçijÜaEŽaRŌæLSäznärsäRfäzëa;ççTlæzëäyNäS;äzd'èøçZÇaLæyÇaijRælaadNñij

```
fleetrun --gpus 0,1,2,3,4,5,6,7 bert_recompute.py
```

æTlædIJætNërT

æLSäznäIJIBERTælaadNäyLärzrecomputeçZDæTlædIJèŒZëaÑäžæætNërTñijNä;ççTlæRecomputeaRŌB sizeaRfäzëæL'l'ad'g9aÄ■ad'ZäÄÇäyŌæüüaRLçs;äžæyÄætüa;ççTlæÜñijNBatch_sizeaRfäzëèçZäyÄæ■æL

Model	Baseline	Recompute	Recompute + mixed precision
Batch size	14	130	145
speed	69.92 sents/s	45.76 sents/s	75.84 sents/s

12.3.2 Gradient Merge

äyNéicñijNæLSäznäzNçz■æÇä;Tä;ççTl Gradient Merge
ælæL'l'ad'gBERTælaadNäLæyÇaijRèøçZÇäy■çZD Batch
SizeñijLäAĞèøçèDŽæIJñäR■çğräyžbert_gradient_merge.pyñijZ

äyŌ Forward Recompute Backpropagation çZyâRñijNæLSäznèçÜaĒLèçAæüzaLäa;IètÜñijNäöZäzLäl

æúẏāāçlèŧŮ

```
# -*- coding: UTF-8 -*-
import paddle
import fleetx as X
import paddle.fluid
import paddle.distributed.fleet as fleet
```

ǎŏŽāzL'ǎLĚāyĈāijŔāelāāijŔāzúǎLǎgNāNŮ

ǎLǎèjǎelāāđNāŔLæŧŕæŋŏ

```
model = X.applications.Bert_large()
downloader = X.utils.Downloader()
local_path = downloader.download_from_bos(
    fs_yaml='https://fleet.bj.bcebos.com/small_datasets/yaml_
    ↪example/wiki_cn.yaml',
    local_path='./data')
data_loader = model.(
    local_path,
    max_seq_len=512,
    batch_size=13,
)
```

ǎŏŽāzL'Gradient Merge Strategy ǎŔL Optimizer

ǎIJlāyLéÍçŽDāzççǎAāyŋŋijNǎLŠāznǎŏŽāzL'āzĚBatch Sizeāyž13ŋijNǎIJlēŹāyǎæŋēāyŋŋijNéǎŽēŹGēŏ, SizeǎĚāelāēNšāyǎāyĥđ'gBatchçŽDēŏŋçžĈijNāzŎēǎNēççǎLŕāzĚBatch sizeāyž52çŽDēŏŋçžĈæŧLæđIJǎǎĈ

ǎIJgradient_merge_configsāyŋŋijNavgéǎLēāzçŧlāžŎæŎgǎLūāçŕāžçŧŕēŏaçŽDǎçāijŔijŽǎçŧŕæŋŋijNāijŽāŕzæŕŔāŋāçŽDæçŕāžçæŕŔāŋāzúǎAžāzšǎIŋijZǎŔŋāzNǎŕĚçŧŕæŎēārzæçŕāžçæŕŔāŋāzŋijN

```
dist_strategy = fleet.DistributedStrategy()
# āççŧŧlGradient mergeçŋŧçŧēāzŧŧēŏççŧçŽyǎĚšǎŔçæŧŕ
dist_strategy.gradient_merge = True
dist_strategy.gradient_merge_configs = {"k_steps": 4, "avg": True}
optimizer = fluid.optimizer.Adam(learning_rate=configs.lr)
optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)
```

ǎijǎāgNēŏŋçžĈ

Gradient Merge çŽDēŏŋçžĈāzççǎAāyŎ Recompute çŋŧçŧçŽyǎŔŋijNçŧŧlæLūāççŧŧlāyđ'ēāNāzççǎAā

```
trainer = X.MultiGPUPTrainer()  
trainer.fit(model, data_loader, start_step=10)
```

è£ŘèąÑèő■çżĈèĎŽæIĴň

```
fleetrun --gpus 0,1,2,3,4,5,6,7 bert_gradient_merge.py
```

CHAPTER 13

ä;ŁçŦÍLARS / LAMB äijŸăŇŮăĽĚăŸĈäijRèúĚăd'ğbatch èő■çžĈ

13.1 çőĂăžŇ

ăĬĬăŦŕă■őăžűëăŇăĽĚăŸĈäijRèő■çžĈăĬĬăŦŕăŸă■, äŸŸă;ŁçŦÍăĈďăĽăGPUăŦŕéĜŖçŽĐăŮžăijRăĬăĽăă
äŸžăžĚăĬăĬăŦŕăGPUçŽĐçőŮăĽăŽă;ŮăĽŕăĚĚăĽĚăĬŦŕŦÍ, æŕŦăijăGPUă■ăŸăŁçŽĐbatch
sizeéĈ;éĬĬăĚăĚăŸăď'ğ. äŽăă■ď'ăĬĬăĈďăĽăGPU æŦŕéĜŖăŦŕăŮăŮă, èő■çžĈçŽĐăĬăŸăĬ-
batch size äžšăijŽăŦŸăď'ğ.

ă;ĚăŮăĽăď'ğçŽĐăĬăŸăĬăbatch size äijŽăŸăæĬăèő■çžĈçŽĐăŦŮăŦŦŽéŮăćŸ[1] [2]:

- æĬăďŇăĬăĬăçžŁçš;ăžăæ■šăď'ś
- æŦŮăŦŦŽéĂšăžăăŦŸăĚć, éĬĬăĚăĚăŽŦ'ăď'ŽçŽĐepoch æĽ■èĈ;ăŦŮăŦŦŽ

LARS[3] äŇŇ LAMB[4] äŸď'äŸăijŸăŇŮç■ŮçŦăŸŸçŦĬăĬăğĈăĚăŸăŸăĽăŦŕăŮăĬăğbatch
èő■çžĈăŸçŽĐăŦŮăŦŦŽéŮăćŸ.

Paddle äőďçŦŕăžĚăŁăžăď'çğ■ăijŸăŇŮç■ŮçŦăijŇpaddle.distributed.fleet
ă;ĬăŸžPaddleéĂžçŦÍçŽĐăĽĚăŸĈäijRèő■çžĈăĬăŦŕăŸă;ŽăžĚçőĂă■ŦăŸŸçŦÍçŽĐăŮăăŦŕăĈ,
çŦĬăĽăŮăŦŕăĬăĬăĚăĚăŮăăĜăăăŇăžčăăĤăŦŕăŦŕăŦŦç■ŮçŦăăĽăăĚăăĽăŦŦŮăăĬĬçŽĐăő■çžĈăŸ■ăĂĈ
éĂžăŁăĜăŁăžăď'äŸăijŸăŇŮç■ŮçŦă,
ăĬĬăŦŕăŸă■ăőďçŦŕăžĚăŦŦ'ăŦŦçŽĐăŦŮăŦŦŽéĂšăžăăŦŮăŮă■šçŽĐçš;ăžă, çžšăŦŦFleet
äŸăăĚăăžăŮçŽĐç■ŮçŦă(e.g. [AMP](#)) äŦŕăžăăďăăď'ğçijŦçš■çŽĐăő■çžĈăŦŦă;šçŽĐtime2train.

13.1.1 13.1.1

resnet50 imagenet	Global batch size	epoch	top1
[Goyal et al]	8k	90	76.3%
LARS Paper	32k	90	72.3%
[fleet: lars + amp]	16k	60	76.2%
[fleet: lars + amp]	32k	62	75.9%

13.2 LARS

æŁŚāznāzēāIJlāTæIJžād'ŽāāāyŁResnet50 èőčzČāyžčōĀāTāĴNāRāzNčzfleet äyŁARSčŽDčTlæšTāĀĆ

```
import os
import fleetx as X
import paddle
paddle.enable_static()
import paddle.fluid as fluid
import paddle.distributed.fleet.base.role_maker as role_maker
import time
import paddle.distributed.fleet as fleet
```

éĀŽēfGX.parse_train_configs() æŌěāRčijNčTlæLŭāRrāzēāōŽāzL'èőčzČZyāĒščŽDāRČæT
init() æŌěāRčāōŽāzL'āžĒāLĒāyČāijRāēlāādNijNāyNéĪcāžččāĀāyčŽDis_collective=TrueēāĪčd'ž

```
paddle.enable_static()
configs = X.parse_train_configs()
fleet.init(is_collective=True)
```

čTlæLŭāRrāzēēĀŽēfGX.applicationsæŌěāRčāLāē;æŁŚāznēcDāĒLāōŽāzL'āē;čŽDāēlāādNijNāē

```
model = X.applications.Resnet50()
downloader = X.utils.Downloader()
local_path = downloader.download_from_bos(
    fs_yaml='https://fleet.bj.bcebos.com/test/loader/small_imagenet.
    ↪yaml',
    local_path='./data')
batch_size = 32
loader = model.get_train_dataloader(local_path, batch_size=batch_
    ↪size)
```

LARS āijYāNŮčōŮæšTčŽDāĒñāijRāēCāyN:

$$local_learning_rate = learning_rate * lars_coef * \frac{||param||}{||gradient|| + lars_weight_decay * ||param||}$$

$$velocity = mu * velocity + local_learning_rate * (gradient + lars_weight_decay * param + epsilon)$$

ĀRāzēçIJNāLřLARS āĚŭāōđæŸřāIJĭ āyēweight decay çŽĐmomentum
 āijYāNŪāZĭçŽĐāšžçāĀyŁāLāāĚēāžĚllocal learning rate çŽĐēĀžēçŠ,
 āřzærŘāyĀāsČçŽĐlearning rate ēfZēāNāžĚāTçijl'. fleet āřĚ LARSāōđçŌřāyžāyĀāyĭ
 fleet meta optimizer, āIJlā;ŁçTĭlāŪŭēIJĀēēĀēō;ç;ōāyĀāyNāĠāçČz:

- ```
dist_strategy = fleet.DistributedStrategy()

dist_strategy.lars = True
dist_strategy.lars_configs = {
 "lars_coeff": 0.001,
 "lars_weight_decay": 0.0005,
 "exclude_from_weight_decay": ['batch_norm', '.b_
↪0']
}

optimizer = fluid.optimizer.Momentum(learning_rate=0.01, momentum=0.
↪9)
optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)
```

```
place = fluid.CUDAPlace(int(os.environ.get('FLAGS_selected_gpus', 0)))
exe = fluid.Executor(place)
exe.run(fluid.default_startup_program())

for i, data in enumerate(loader()):
 start_time = time.time()
 cost_val = exe.run(model.main_prog,
 feed=data,
 fetch_list=[model.loss.name])

 end_time = time.time()
 print(
```

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(c) 2019

```
"worker_index: %d, step%d cost = %f, speed: %f"
% (fleet.worker_index(), i, cost_val[0], batch_size / (end_
↪time - start_time)))
```

### 13.2.1 13.2.1

13.2.1 13.2.1

```
fleetrun --gpus 0,1,2,3,4,5,6,7 --log_dir log example_lars.py
```

## 13.3 LAMB

13.3 LAMB 13.3 LAMB

```
import os
import fleetx as X
import paddle
paddle.enable_static()
import paddle.fluid as fluid
import paddle.distributed.fleet.base.role_maker as role_maker
import time
import paddle.distributed.fleet as fleet
```

13.3 LAMB 13.3 LAMB

```
paddle.enable_static()
configs = X.parse_train_configs()
fleet.init(is_collective=True)
```

13.3 LAMB 13.3 LAMB

```
model = X.applications.Resnet50()
downloader = X.utils.Downloader()
local_path = downloader.download_from_bos(
 fs_yaml='https://fleet.bj.bcebos.com/test/loader/small_imagenet.
↪yaml',
 local_path='./data')
batch_size = 32
loader = model.get_train_dataloader(local_path, batch_size=batch_
↪size)
```

13.3 LAMB 13.3 LAMB

$$m_t = \beta_1 m_{t-1} + (1 - \beta_1) g_t$$



(çzäyLéat)

```

end_time = time.time()
print(
 "worker_index: %d, step%d cost = %f, speed: %f"
 % (fleet.worker_index(), i, cost_val[0], batch_size / (end_
→time - start_time)))

```

### 13.3.1 èŒŘëąÑèóçzĈèĎŽæIĴň

áoŒæĹŔäyŁèřřèĎŽæIĴňçŽĎçijŮâĚZâŔŔijŒæĹŚäznâřsâŔřäzëä;řçŤlázëäyŒâŚ;äzd'äyÄëąŒâŔřâĹlâ■T

```
fleetrn --gpus 0,1,2,3,4,5,6,7 --log_dir log resnet50_lamb.py
```

## CHAPTER 14

---

ä;£çTÍFleetè£ZèaÑaijCædDåRCæT̃ræIJ■åŁaāZíèõ■çzČ

---

åijCædDåRCæT̃ræIJ■åŁaāZíçZõaL'■äzĚæT̃ræÑAaIJléIZæĀAāZ;äyÑè£RèaÑ

### 14.1 äzĀäZŁæYřaijCædDåRCæT̃ræIJ■åŁaāZíiijš

åIJlāijĀāgÑā;£çTÍāi jCædDåRCæT̃ræIJ■åŁaāZíāL'■iijNæCíéIJĀèçAāĚLäzĚègčāRCæT̃ræIJ■åŁaāZí

#### 14.1.1 åRCæT̃ræIJ■åŁaāZíçZDāžTçTíécĒaššäzèaRŁègčāĒççZDēUóécY

åRCæT̃ræIJ■åŁaāZíéZĒäy■āžTçTíāIJlNLPāĀAæŌíè■Ř äžčāRŁ  
æŘIJçt' çécĒaššiiijNāĚüäyžèçĀéŠLāržäzèäyNāyd' äyĽeUóécYiijZ

##### 1. åd'gæT̃ræ■ōiijZ

åŌšāgNæT̃ræ■óéZĒäždāð' giiijNāLíè;DāGāçZ;GçZDæT̃ræ■óéGRiijNā■TæIJžèõ■çzČéĀšāžçéZ;äžæL

##### 2. åd'gāRCæT̃riijZ

åIJläyŁè£řāIJžæZřäy■iijNæT̃ræ■óéGRād' gçZDāRÑæUüiijNāijt' éZŘçĪĀçL'žā;AçZDçĪĀçÚRæĀgiiijNā

åIJlāuēäyZāIJžæZřäy■iijNæřēEmbeddingāRCæT̃rçZDçzt' āžçā;Āā;ĀæYřäz£çžgiiijNā■āçTíçZDā■YāCí

#### 14.1.2 äijăçzšāRCæT̃ræIJ■åŁaāZíçZDāsĀéZŘ

ā;šāL'■åRCæT̃ræIJ■åŁaāZíçZDWorkerèŁCçCziiijNéĀžāyÿä;£çTÍCPUæĹŪGPUæIJžāZíāōNæLRæĪāā

ā;šWorkerä;£çTíçZDèõ;ād' GçāōāōZiijNāĚüçāñāzūçōŪāŁZçZDēĒ■ærTāzšéZŘāzNāZžāōZāĀCāZžāō

### 1. GPUæIJžāZíāL'çTíçÓĖè;Čä;Ö

ēNēō■czČžDāīāđNāy■ād'■āiCīijNāçCāŌīē■RēcEāššāyçTíçžžDeepFMāĀLRīijNāīāđNēōāç

### 2. CPUæIJžāZíōŪāLZæIJL'çŠúéĹ

CPUæIJžāZíēĀžāyāyāfČæTřè;Čād'ŽīijNāžūāyTæIJžāZíāzūāījāžšæŽt'ā;ŁāōIJīijNāRřāzēāĒĒāŁEā

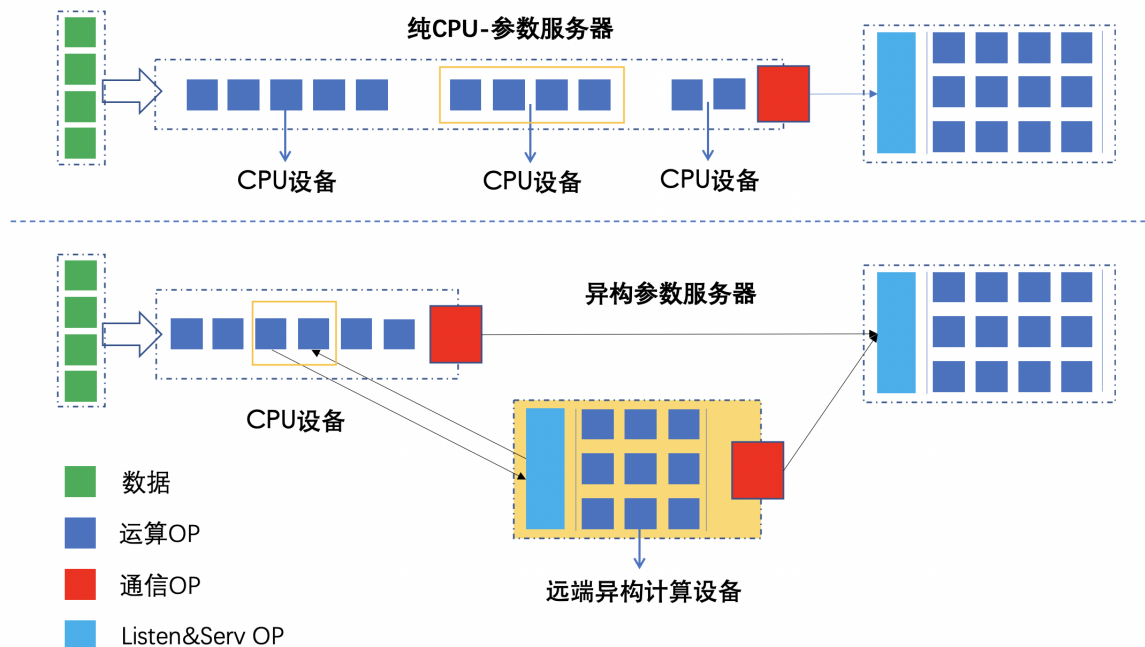
### 3. æŪřāđNçōŪāLZæŌēāĒēāĹRæIJñè;Čād'g

ēŽRçīĀAiēŁrçL'ĠāRŠāsTæŪēæŪřæIJĹāijCīijNāRĐçg■ēnYçōŪāLZā;ŌāĹRæIJñçžžDēŁrçL'ĠāũšēŁZā

## 14.1.3 āijCæđDāRCæTřæIJ■āŁāāZíāzNçz■

éCčāzĹīijNāRřāy■āRřāzēāĹīēĀĀēřČæTř'æIJžāZíēĒ■æřTīijšāRŇæŪūēgčāEšIOçŠúéĹāzēāRŁçōŪāŁZē

PaddlePaddleāšžāžŌāũēāyŽāōđēūīijNāĹZæŪřæĀgçžžDæRŘāĠžāžEāijCæđDāRCæTřæIJ■āŁāāZíīijNæ'



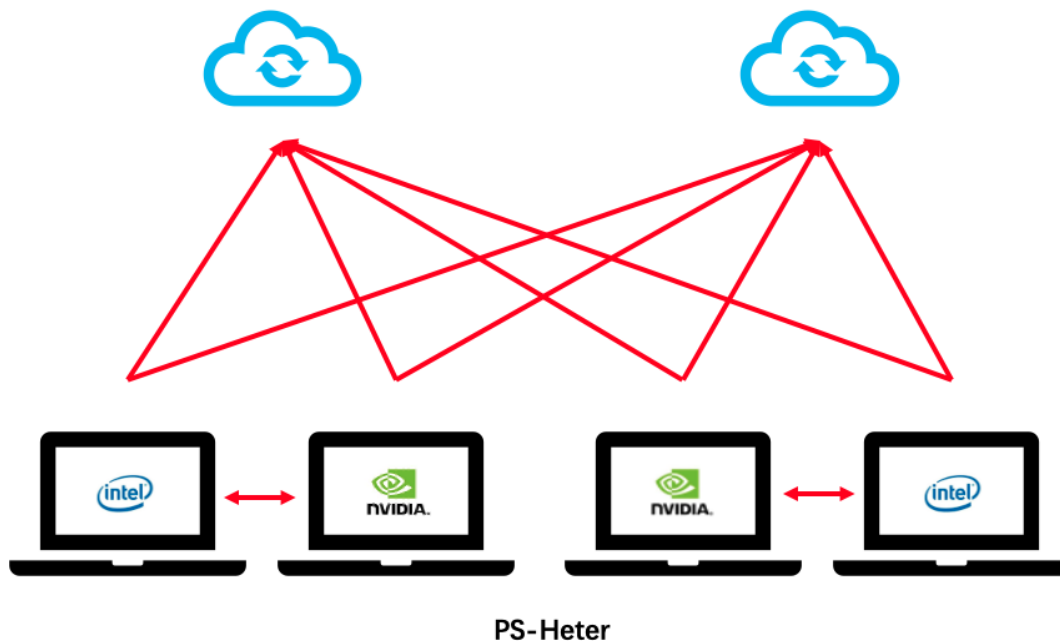
## āijCæđDāRCæTřæIJ■āŁāāZíāšžæIJñāŌšçŘĒ

āyĀāyĹēūšāžēā■ēāžāēīāđNçžžDēō■czČēŁĠīNāRřāzēāŇēāĹēāyžāyL'æ■ēīijŽīāĀāāL'■āRŠēōāçōŪŁō

āRCæTřæIJ■āŁāāZíēāījRāyNīijNāL'■āRŠāRĹāR■āRŠæ■ēēĹd'āIJĹworkerçnr(āžšçgřāyžTrainer)āō

āijCæđDāRCæTřæIJ■āŁāāZíēāījRāy■īijNāĹSāžñēŁZāyĀæ■ēāŇēāĹēāL'■āRŠāRĹāR■āRŠīijNāRřāzē

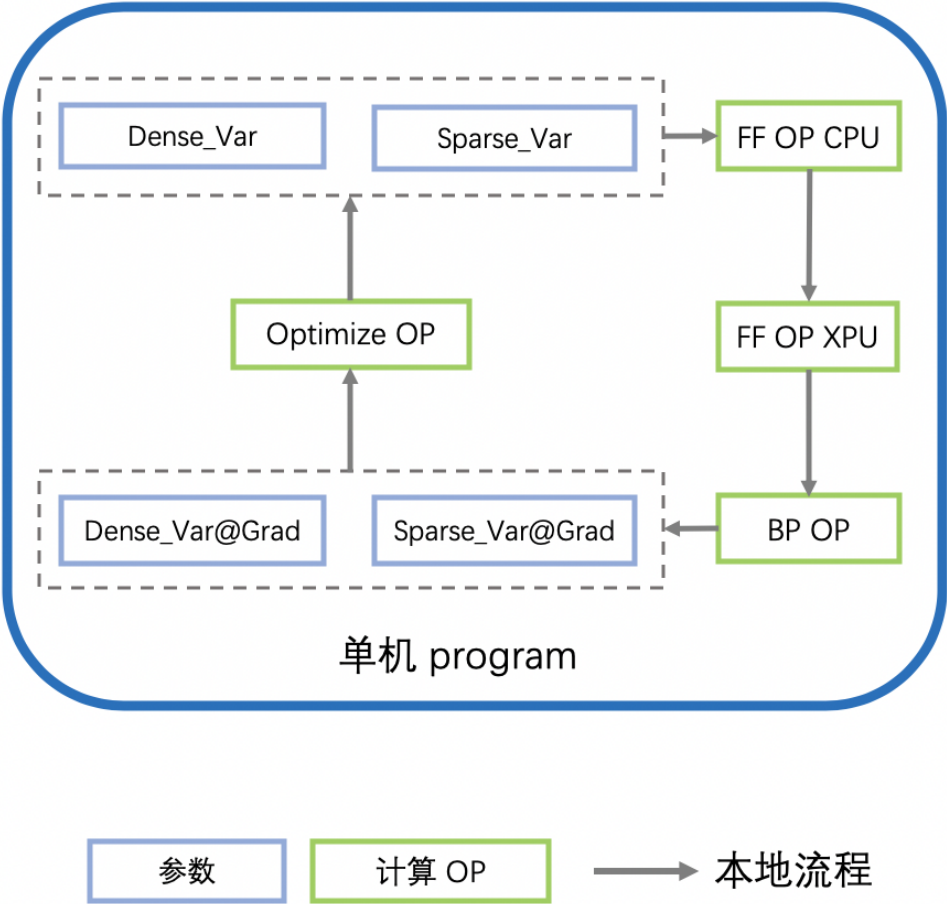
CPU-TrainerāŠŇHeter-TrainerāžNēŪt'āijžēŁZēāŇēĀžāēāīijNāžd'æ■ç;ŠçzIJēŁRēāNāL'ĀēIJ



## 14.1. FleetX 0.1.0.beta

- FleetX 0.1.0.beta

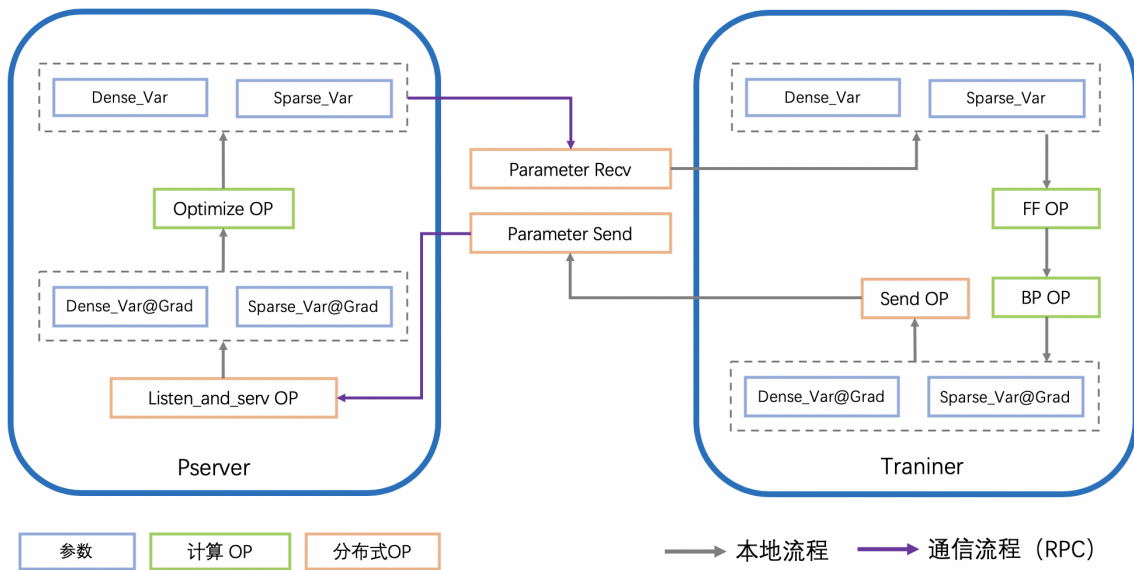
Paddle单机运行流程



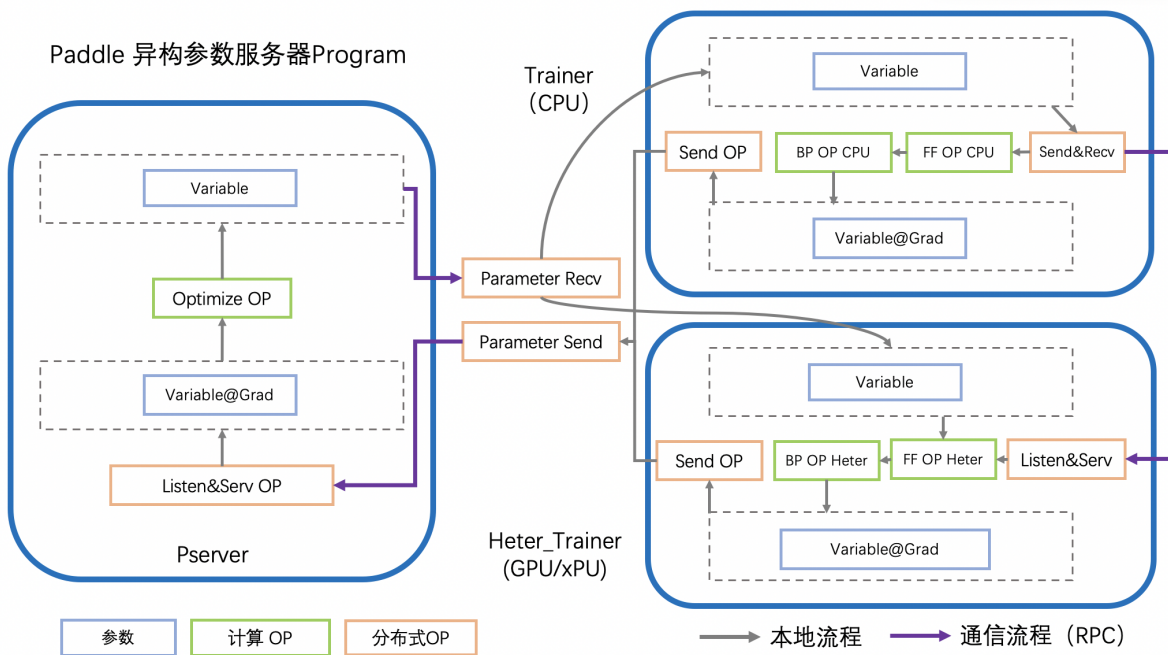
- äijăczşăŖĆæŦŕæIJ■ăŁăăŹÍçŽĐëŦŖëăŦăŦşçŖĖăŹ;



Paddle 纯CPU参数服务器Program



### • Paddle 异构参数服务器Program



### • Paddle 异构参数服务器Program

ayNéIcäzNçz■aijCædDâRÇæTŗæIJ■āLāāZīcZDā;ŁçTīæŪzæşTīijNæŌlè■RāĒLāIĴā■čäyŷāRÇæTŗæIJ■  
 äzëäyNçd'zä;NçZDāōNæTŗ'äzççāAä;■āZŌFleetX/example/  
 heter\_parameter\_server/demo.py

### • 1āĀAēō;ç;ōēŁRēāNāIĴāijCædDēō;ād'GäyŁçZDçzDç;Ś

æūsāžēā■ēäzçzDç;ŚīijNéĀZäyŷāRŗäzēæNĒēğçäyžäyð'ēČlāĒEīijZ1āĀAĴōārEēZEādNçzDç;ŚīijZ2āĀA

```
with fluid.device_guard("cpu"):
 input_data = paddle.data(name="sparse_input", shape=[None, 1],
↪dtype="int64")
 input_label = paddle.data(name="label", shape=[None, 1], dtype=
↪"int64")
 label = paddle.cast(input_label, dtype="float32")
 embedding = paddle.static.nn.embedding(input_data, is_
↪sparse=True, size=[1000, 128])

with fluid.device_guard("gpu"):
 fc1 = paddle.static.nn.fc(embedding, size=1024, act="relu")
 fc2 = paddle.static.nn.fc(fc1, size=512, act="relu")
 fc3 = paddle.static.nn.fc(fc2, size=256, act="relu")
 predict = paddle.static.nn.fc(fc3, size=2, act="softmax")
 label = paddle.cast(label, dtype="int64")
 cost = paddle.nn.functional.cross_entropy(input=predict,
↪label=label)
```

1. IOârEęŻEąđNçŽĐOPéÁĆăŘĹĹaIJÍCPUeőĹăđ' ĞăyĹëfŘëąNġijNăĵăŧăŦŕă■őëĴŞăĔëëĴŞăĢžăy■ăE■ăĹŔă
2. èőąçőŰârEęŻEąđNOPæŦĴăIJÍGPUç■Ĺ'AlëĹŕçĹ'ĢëőĴăđ' ĞăyĹiijNăŔŕăžăăĔĔăĹĹăĹĹ'çŦĴçőŰăĹŽiijNăĹ

ä;£çŦİfleet apiäŦŦäLäiäŦCädŦDäŦŦCæŦŦŦäJ■äLääŦŦliiŦŦŦäJŦÄèçÄèĖ■ç;őDistributedStrategyiiŦŦäŦä;



## ä;ŁçŤífleetrunāŘřāŁíāijČædĎāŘČæŤřæIJ■āŁāāŽíēő■çzČ

fleetrun æŸř paddle2.0rcçŁ'ŁæIJñäzēāŘŌæŮřāŁāāĚēçŽĎāŁēāyČāijŘēő■çzČāŘřāŁíāüēāĚüüijŇ  
 ā;Şēő■çzČäzčçāAreadyäzēāŘŌüijŇāĀĜāēČēő■çzČāŘřāŁíāĚēāŘČæŸřtrain.  
 pyüijŇāŁŽāŘřæŇŁ'çĚğäzēäyŇçŽĎæŮzāijŘāŘřāŁíāijČædĎāŘČæŤřæIJ■āŁāāŽíēő■çzČüijŽ  
 æŮzæşŤäyÄüijŇēŚĹāřzā■ŤæIJžæíāæŇşāŁēāyČāijŘēő■çzČüijŇā;ŁçŤíēĜíāŁíāŁēĚ■çzČĎipāŞŇport

```
fleetrun --server_num=2 --worker_num=2 --heter_worker_num=2 train.py
```

æŮzæşŤäzŇüijŇēŚĹāřzā■ŤæIJžüijŇæŁŮēĜíāőŽāzŁ'çŽĎāđ'ŽæIJžēő■çzČüijŇā;ŁçŤíæŇĜāőŽçŽĎipāŘŁç

```
fleetrun --servers=ip:port,ip:port --workers=ip:port,ip:port --

 ↳heter_workers=ip:port,ip:port train.py
```

æŮzæşŤäyŁ'üijŇēŚĹāřzPaddleCloudāzşāŘřçŽĎcustom-frameworkæíāāijŘüijŇæŇĜāőŽāzāŁççŽĎāŘřāŁ  
 PaddleCloudæŸřçŽ;āžçāĚēĚČíçŽĎæūsāžçā■ēäzāāzāŁāāzşāŘřüijŇæŘŘā;ŽāžĚā;Łæ■űçŽĎæŘŘāžđ'æt;

```
heter_workeræŤřéĜŘāijžæāžæ■őéĚ■ç;őçŽĎGPUēőç;ād'ĜæŤřéĜŘēĜíāŁíēřČæŤt'

 ↳æűzāŁāēřēēĚ■ç;őæŸřäyžāžĚæŇĜāőŽfleetrunēĹŘēāŇāIJíāijČædĎāŘČæŤřæIJ■āŁāāŽíēő■çzČüijŇā

 fleetrun --heter_worker_num=2 train.py
```

## āijČædĎāŘČæŤřæIJ■āŁāāŽíā;ŁçŤíçđ'zä;Ň

çđ'zä;ŇäzčçāĀä;■āžŌFleetX/example/heter\_parameter\_server/

- æŤřæ■őäyŇē;;

```
bash sh download_data.sh
```

æŁ'ğēāŇērēēĎŽæIJñüijŇāijŽāzŌāŽ;āĚēæžŘçŽĎæIJ■āŁāāŽíāyŁäyŇē;CriteoæŤřæ■őéŽĚüijŇāzűēğçāŌ  
 /train\_data\_full/üijŇāĚíēĜŘætŇērŤæŤřæ■őæŤ;ç;őāžŌ./test\_data\_full/  
 üijŇçŤíāžŌāŁnéĀşēíŇērĀçŽĎēő■çzČæŤřæ■őäyŌætŇērŤæŤřæ■őæŤ;ç;őāžŌ./  
 train\_data/äyŌ./test\_data/āĀČ

ēĜşæ■đ'üijŇæŁŚāzñāūsāőŇæĹŘæŤřæ■őāĜĚād'ĜçŽĎāĚíēČíāüēä;IJāĀČ

- āŘřāŁíēő■çzČ

```
ps-cpu

fleetrun --server_num=2 --worker_num=2 heter_train.py

ps-heter

fleetrun --server_num=2 --worker_num=2 --heter_worker_num=2 heter_

 ↳train.py
```

1. éčđæąłǻżȚǻśĆǻŁĘǻŸČǻijŔAPIçŽĎǻj£çŦíæąŁǻčŦ

## 15.1 1.1 çõÄäzÑ

95%çŽĎæĭađNēōaçoŮéĠRiĭžāĒĲēđæŌēāsCāŊĒāŔnčē95%çŽĎæĭađNāŔĲæŦŕéĠŔāšŊčē5-  
 10%çŽĎæĭađNēōaçoŮéĠŔāĀĆéĀŽāyŷāĲēōōiĭŊāŋčŕāsĆéĀĈŔĲéĠĠæŦŕæāžūēāNĭĭŊāŽāyžāŋ  
 æĲŋæŮĠæāçāžēAlexNetçŦçŽĲāyžāŋāžŊčŋāēĆāŦāŦçŦĲēđæāĲçŽĎāžŦāsĆéŽĒŔĲéĀžāŦāĲĲāōđ

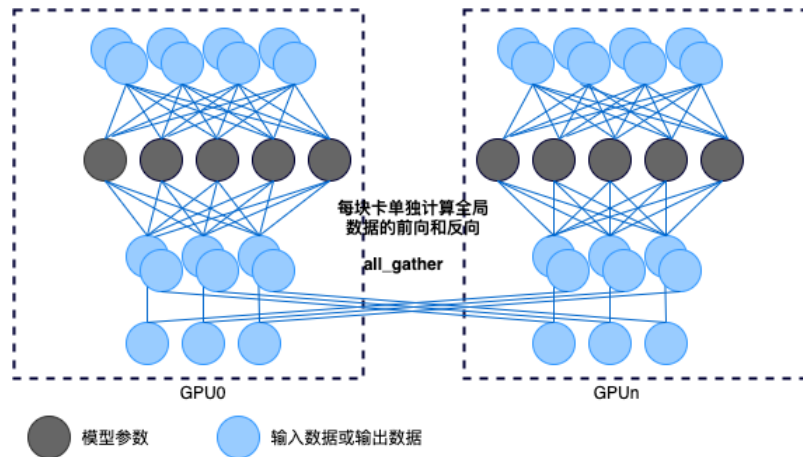
## 15.2 1.2 æl̥aɖd̥N̥aʒũèaŋ̊N̥aŌʂçR̥ɛ̊aʂN̥aŋ̊d̥çŌ̊r

### 15.2.1 1.2.1 çL'ŁæIĴñèçAæśĆ

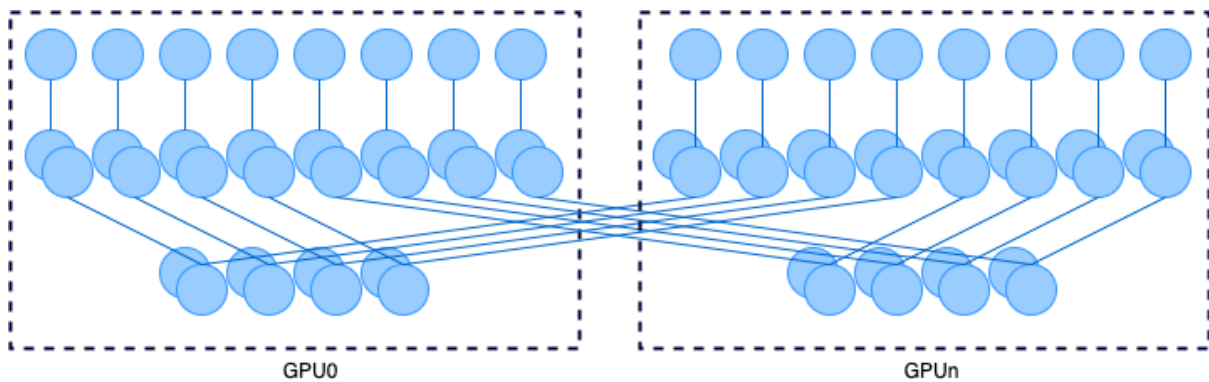
- paddlepaddle 2.0-rc-gpuçL'ŁæIJňăRŁäzěäyŁ

[illegible]

æIjñĚĆiijNæĹSäznäzNçz■æĆä;TǎođĊŎřǎĚĹēđæŎěāsĆçŽDæĹqāđNázűèqNǎǺĆ  
ééŬǎĚĹiijNæsĠèAŻǎRĎǎiŬGPUǎ■ǎǎĚĹēđæŎěāsĆçŽDè;SǎĚěæTřæ■őiiJNǎ;ŬǎĹǎǎĚĹāsǺē;SǎĚěæTřæ



æŒçĭÄiijŊæsĠëAžĀRĎăiŮGPUă■ăăĒĲēđæŒëăĲçŽĎëĴŖăĠžæŤræ■ŕiijŊăžŭæĴĵăŖŮæIJăĲăIŮGPUçŽ



### 15.2.3 1.2.3 æĲăĀăŽĴăŏđçŎř

äŷĴēĲřēĴĠçĲĲăŖŖēĲřăIJřăŏŊæŤr'ăĴ■ăŖŖēŏăçŏŮēĴĠçĲĲăŏđçŎřăžççăAăĲçăŷŊiijŽ

```
-*- coding: UTF-8 -*-
import paddle
import paddle.nn as nn

æŒžăžĴĴ'æĲăăđŊăžŭëăŊçŽĎăĒĲēđæŒëăĲçŽĎëĴŖăĠžæŤræ■ŕiijŊăžŭæĴĵăŖŮæIJăĲăIŮGPUçŽ
↪ Layer
class ModelParallelLinear(nn.Layer):
 def __init__(self,
 in_dim,
 rank_num,
 rank_id,
 class_num):
 super(ModelParallelLinear, self).__init__()
 if class_num % rank_num:
 raise ValueError("Number of classes must be divisible "
 "the number of ranks.")
 shard_dims = class_num // rank_num
 self.linear = nn.Linear(in_dim, shard_dims)
```

(äŷŊéăĲçŽçç■)

(çzäyLéat)

```

self.rank_num = rank_num
self.rank_id = rank_id
for parameter in self.linear.parameters():
 parameter.is_distributed = True

def forward(self, x):
 global_x_list = []
 paddle.distributed.all_gather(global_x_list, x)
 global_x = paddle.concat(global_x_list, axis=0)
 out = self.linear(global_x)
 global_out_list = []
 paddle.distributed.all_gather(global_out_list, out)
 all_outs = paddle.concat(global_out_list, axis=1)
 out = paddle.split(all_outs, self.rank_num)[self.rank_id]
 return out

```

ad'GæşliijŽāZāäyžæfRāiŪGPUā■āfiā■YéČlāLēāĖlēfđæŌēāsCāRCæTŗijNāyLēlćçŽDäçNā■Rāy■ēōç  
 āōNæTŗ'āIŗēō■čZāzčçāAāōđçŌŗæCāyNīijŽ

```

-*- coding: UTF-8 -*-
import paddle
import paddle.nn as nn
import paddle.nn.functional as F
from paddle.fluid.dygraph import Conv2D
#āĹēāyČāijŖstep 1: āŗijāĖĖpaddle.distributed.fleetāNĖ
from paddle.distributed import fleet
from model_parallel_linear import ModelParallelLinear

āōŽāzL'āĖlēfđæŌēç;ŚçzIJiijNéIJĀçzğæL'ēēĞlñn.Layer
class SimpleModelParallelClassifierNet(nn.Layer):
 def __init__(self,
 class_num,
 rank_num,
 rank_id):
 super(SimpleModelParallelClassifierNet, self).__init__()
 self.conv1 = nn.Conv2d(3, 64, kernel_size=11, stride=4,
→padding=2)
 self.max_pool1 = nn.MaxPool2d(kernel_size=3, stride=2)
 self.conv2 = nn.Conv2d(64, 192, kernel_size=5, padding=2)
 self.max_pool2 = nn.MaxPool2d(kernel_size=3, stride=2)
 self.conv3 = nn.Conv2d(192, 384, kernel_size=3)
 self.conv4 = nn.Conv2d(384, 256, kernel_size=3)
 self.conv5 = nn.Conv2d(256, 256, kernel_size=3)
 self.max_pool5 = nn.MaxPool2d(kernel_size=3, stride=2)
 self.model_parallel_linear1 = ModelParallelLinear(2304,
 rank_num,
 rank_id,
 4096)

```

(äyNéatçzğçz■)



```

self.model_parallel_linear2 = ModelParallelLinear(4096,
 rank_num,
 rank_id,
 4096)

self.model_parallel_linear3 = ModelParallelLinear(4096,
 rank_num,
 rank_id,
 class_num)

self.droupout = nn.Dropout(0.5)
self.relu = nn.ReLU()

def forward(self, x):
 x = self.conv1(x)
 x = self.relu(x)
 x = self.max_pool1(x)
 x = self.conv2(x)
 x = self.relu(x)
 x = self.max_pool2(x)
 x = self.conv3(x)
 x = self.relu(x)
 x = self.conv4(x)
 x = self.relu(x)
 x = self.conv5(x)
 x = self.relu(x)
 x = self.max_pool5(x)
 x = F.dropout(x, 0.5)
 x = paddle.reshape(x, [x.shape[0], -1])
 x = self.model_parallel_linear1(x)
 x = F.dropout(x, 0.5)
 x = self.model_parallel_linear2(x)
 out = self.model_parallel_linear3(x)
 return out

```



(çz■äyŁéął)

```

for step in range(20):
 # 2. æL'ğèàŃâL'■âŔŚç;ŚçżIJ
 image = paddle.randn([1, 3, 224, 224], 'float32')
 label = paddle.randint(low=0, high=10, shape=[1,1])
 output = dp_layer(image)
 loss = F.softmax_with_cross_entropy(output, label)
 loss = paddle.mean(loss)

 print("step:{}\tloss:{}".format(step, loss.numpy()))

 # 3. æL'ğèàŃâŔ■âŔŚèóaçŎŮâŠŃâŔĆæŦřæŽt'æŮř
 # âĹĒâÿČâijŔstep 4:␣
 ↪âIJlæL'ğèàŃâŔ■âŔŚiijŁbackwardâĜ;æŦřiijL'âL'■âŔŎèŁŽèàŃæ■Śâd'śçijl'æŦçâŠŃâŔ■âŔŚæ
 loss.backward()

 adam.step()
 adam.clear_grad()

```

ârĒäyŁèçřäzčçăAăfĬâ■Ÿäyžtrain.pyijŃâAĜèŎçèçAèçŔèàŃ2â■ăăzzaŁajijŃéČčäzŁâŔĬéIJĀèçAăIJlâŚjăz

```
fleetrn --gpus=0,1 tain.py
```

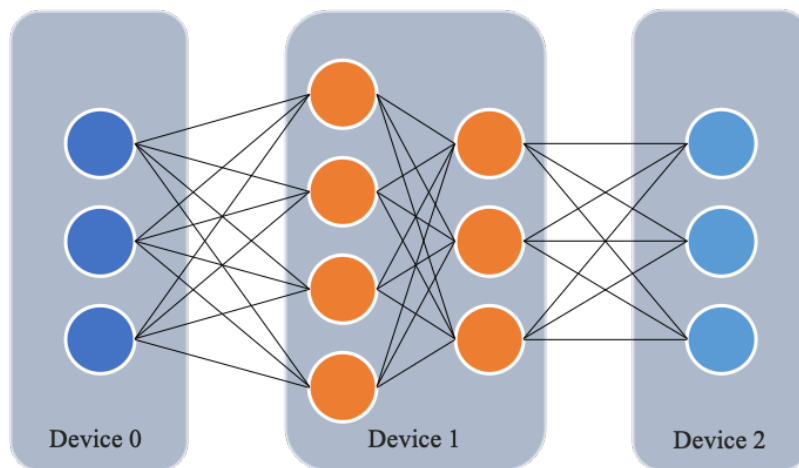
## ä;£çŦíæŧAæřťçžŁąúèąŃèŁŻèąŃèő■çžČ

### 16.1 çőÄäžŇ

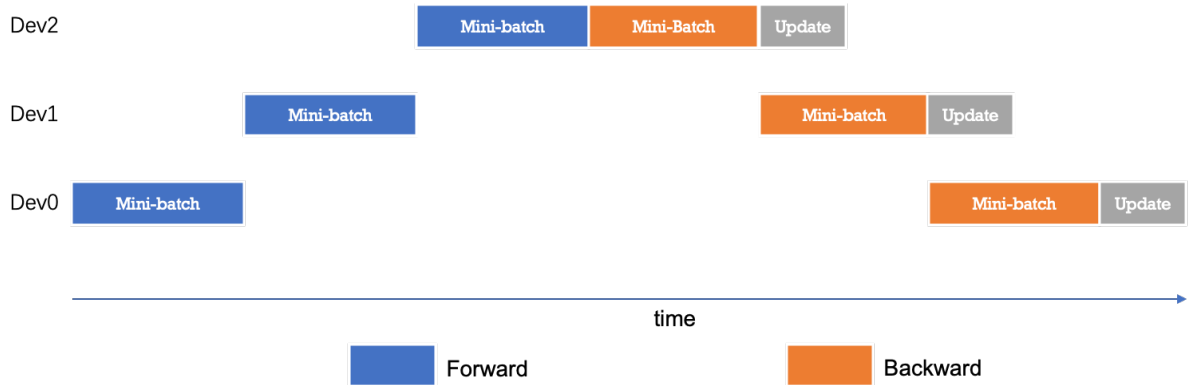
éŽŔçĹĀād'Žçğ■çěđçžŔç;ŚçžIJāŁăĀšèő;ăđ'ĠăŠŇăŷŚçŦíçěđçžŔç;ŚçžIJèőąçőŮèŁŕçŁ'ĠçŽĐăĠžçŎřijŇ

### 16.2 ăŎŞçŘĚ

æŧAæřť çžŁąúèąŇăĹĚăŷČăijŔăĹăæIJŕăŷŎæŦŕăë■őăžúèąŇăŷ■ăŔŇijŇéĂŽèŁĠăŕĚăĹăăđŇăĹĠăĹĚăĹăŕ



ăĚŮă;ŞăIJŕèőřijŇăĹ■ăŔŚèőąçőŮèŁĠŕçĹŇăŷ■ijŇè;ŞăĚăæŦŕăë■őéęŮăĚăĹĹèő;ăđ'Ġŏăŷ■éĂŽèŁĠçňŇăĹă  
ăęČăŷŇăŽ;ijŇăŷžæŧAæřť çžŁąúèąŇăŷ■çŽĐăŮŮăžŔăŽ;ăĂČçőĂă■ŦçŽĐæŧAæřť çžŁąúèąŇăŷŮăŷăijŔăŷ



äyžäZĖäijŸāŃŪætAært' çžfázúëāŃçŽDæĀgèĈ;ijjNæĹSäznāRřäzēārĖmini-batchāĹĜāĹĖæĹRèÑeāzšæŽt' āRçšŠāžççŽDmicro-batchijjNæRŘā■ĜætAært' çžfázúëāŃçŽDāzūāRŠāžçijjNæbatchècñāĹĜāĹĖäyž4äyĭmicro-batchijjZāL■āRŠéŸūæōtjijNæfRäyĭèōĹād' ĜäĹĭæñæèōaçōŪā■Täyĭmicro-batchçŽDçzŠædIJijjZèĹŽçg■āĜRārRmini-batchçŽDæŪzāijRāĜRārSäžĖæfRäyĭèōĹād' ĜāōNæĹRäyĭĀæñæèōā



äyŃéĭcæĹSäznāRřĖĀŽèĹĜäĹNā■RäyžæĈĭèōšègçæĈä;Tä;ĹçTĭpipelineç■ŪçTēāIJläyð' āijäGPUäyĭĹèō■ç

## 16.3 äĭĹçTĭæäŭäĹŃ

### 16.3.1 ārijāĖĕäĹĭètŪ

```
-*- coding: UTF-8 -*-
import os
import argparse
import paddle
import time
import math
import numpy as np

import paddle.distributed.fleet as fleet
import paddle.static.nn as nn
paddle.enable_static()
```

### 16.3.2 16.3.2

16.3.2 16.3.2

```
16.3.2 16.3.2
def build_network():
 # Step1: 16.3.2 16.3.2
 with paddle.fluid.device_guard("cpu"):
 data = paddle.data(name='sequence', shape=[1], dtype='int64')
 data_loader = paddle.io.DataLoader.from_generator(
 feed_list=[data],
 capacity=64,
 use_double_buffer=True,
 iterable=False)
 emb = nn.embedding(input=data, size=[128, 64])
 with paddle.fluid.device_guard("gpu:0"):
 fc = nn.fc(emb, size=10)
 loss = paddle.mean(fc)
 return data_loader, loss
```

### 16.3.3 16.3.3

16.3.3 16.3.3

```
fleet.init(is_collective=True)

data_loader, loss = build_network()

dist_strategy = paddle.distributed.fleet.DistributedStrategy()
dist_strategy.pipeline = True
optimizer = paddle.fluid.optimizer.SGDOptimizer(learning_rate=0.1)
optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(loss)

def train_reader():
 for _ in range(100):
 data = np.random.random(size=[32, 1]).astype("int64")
 yield data
```

### 16.3.4

```
place = paddle.CPUPlace()
exe = paddle.static.Executor(place)

data_loader.set_sample_generator(train_reader, batch_size=2)

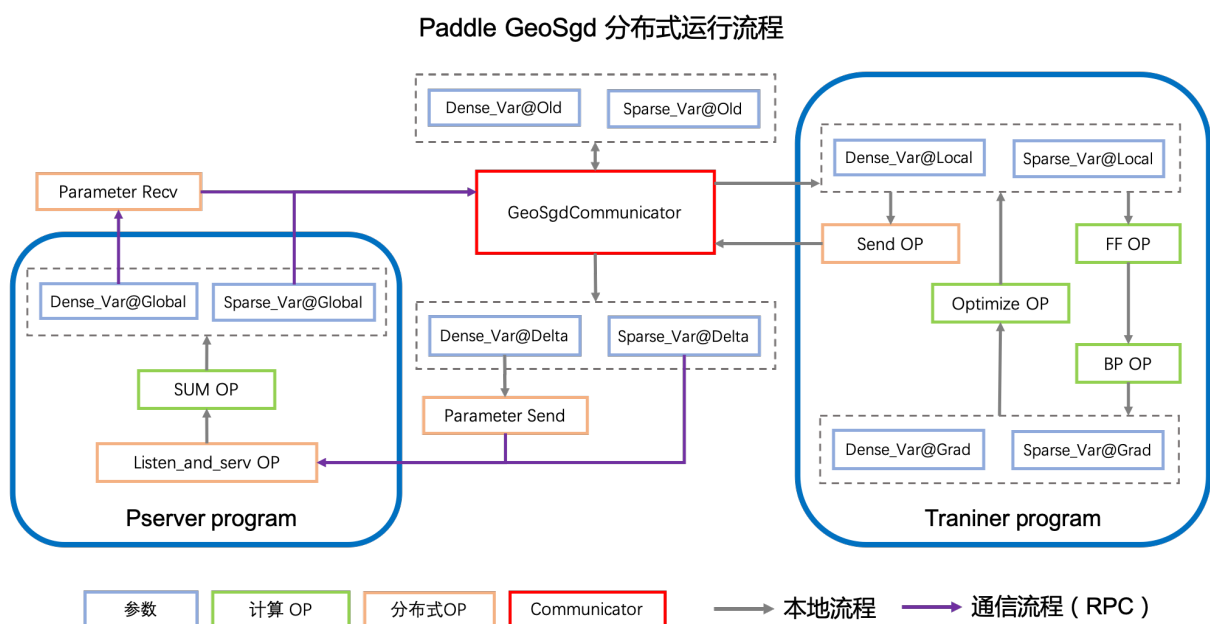
exe.run(paddle.static.default_startup_program())

data_loader.start()
exe.train_from_dataset(paddle.static.default_main_program())
```

## 17.1 分布式训练

分布式训练是指将训练任务分解成多个子任务，每个子任务由一个或多个计算节点（Worker）并行执行。这些子任务通过一个中央协调器（Coordinator）进行通信和同步。在Paddle GeoSgd中，训练流程分为Pserver程序和Trainer程序。

## 17.2 分布式训练流程



在Paddle GeoSgd中，训练流程分为Pserver程序和Trainer程序。Pserver程序负责全局变量的管理和聚合，而Trainer程序负责本地计算和梯度更新。两者通过GeoSgdCommunicator进行通信和同步。

- äyŌæZőéĀŽçŽDāRCæTṛæIJ■āLāāZíāy■āRŇiijŇāIJÍGEOç■ŪçTěäy■iijŇæfRäyŋTrainerèt' šèt' čāIJāeIJ
- GEOæZt' æŪřç■ŪçTěäijŽāIJlěō■çzČèŁGčlŇāy■āRřāLíād' ŽāyŋèŁŽčlŇiijŇèt' šèt' čāRCæTṛæZt' æŪřāRL  
GEOç■ŪçTěéĀŽèŁGāladŇēō■çzČäyŌèŁČçČzéĀŽāāāRŇāēēŁŽēāŇçŽDæŪzāijRiijŇāIJlāēlērAēlā  
æŌěäyŇælēæLSāznārĒéĀŽèŁGā; Ňā■RäyžæČlěōšēğčGEOāIJÍFleetäy■æŸřāČä; TāžTçTlçŽDāĀČ  
āIJlāijĀāğŇāzŇāL■æLSāznēēŪāĒLēIJĀēēAäyŇē; jēō■çzČäy■æL' ĀēIJĀēēAçŽDæTṛæ■ōiijŽ

```
äyŇē; jāžŭēğčāŌŇæTṛæ■ōiijŇēō■çzČæTṛæ■ōēōšāfĪā■ŸēĞšāR■äyž raw_data_
→çŽDæŪĞāžŭād' ž
wget --no-check-certificate https://fleet.bj.bcebos.com/ctr_data.
→tar.gz
tar -zxvf ctr_data.tar.gz
```

## 17.3 æŞ■ä|Jáóðèùt

### 17.3.1 æûzāŁäāçlètŪ

éçŪāĒLæLSāznēIJĀēēAæûzāŁäēō■çzČäy■æL' ĀçTlāLřçŽDpythonāēlāiŪiijŇfleetx  
āRřāžčçTlāžŌāŁäē; jēLSāznāyžçTlāLŭārAēčĒçŽDæŌēāRčāçCiijŽāŁäē; jēlāādŇāRLæTṛæ■ōiijŇælāādŇēō  
distributed.fleet äy■āōŽāzL'āžĒāyřārŇçŽDāLĒāyČāijRç■ŪçTěä; ŽçTlāLŭā; ŁçTlāĀČ

```
import paddle
import fleetx as X
import paddle.fluid as fluid
import paddle.distributed.fleet as fleet
import paddle.distributed.fleet.base.role_maker as role_maker
```

### 17.3.2 āōŽāzL'āLĒāyČāijRāēlāaijRāzŭāLlāğŇāŇŪ

éĀŽèŁGX.parse\_train\_configs() æŌēāRčiiijŇçTlāLŭārRāžēāōŽāzL'ēō■çzČçŽyāĒşçŽDāRCæT  
init() æŌēāRčāōŽāzL'āžĒāLĒāyČāijRāēlāaijRiijŇāōŽāzL'GEOç■ŪçTěä; ŁçTlçŽDāLlāğŇāŇŪæŌēāRčāyČ

```
paddle.enable_static()
configs = X.parse_train_configs()
role = role_maker.PaddleCloudRoleMaker()
fleet.init(role)
```

### 17.3.3 āŁäē; jēlāādŇāRLæTṛæ■ō

āIJlēŁŽāyŋā; Ňā■Räy■æLSāznā; ŁçTlāžĒāyŌāRŇāēē/āijČæēāRCæTṛæIJ■āLāāZíçŽyāRŇçŽDCTR-  
DNNælāādŇāĀČçTlçX.applicationsæŌēāRčāŁäē; jēlāādŇiijŇāzŭāŁäē; jēāōŽāLŭāŇŪçŽDæTṛæ■ōāĀČ

```
model = X.applications.MultiSlotCTR()
loader = model.load_multislot_from_file('./train_data')
```

### 17.3.4 17.3.4 Strategy and Optimizer

The Fleet API provides a `fleet.DistributedStrategy()` and `fleet.DistributedOptimizer()` to help you configure the distributed training process. The `fleet.DistributedStrategy()` class is used to configure the distributed training process, and the `fleet.DistributedOptimizer()` class is used to configure the optimizer.

The `fleet.DistributedStrategy()` class is used to configure the distributed training process. The `fleet.DistributedOptimizer()` class is used to configure the optimizer.

```
dist_strategy = fleet.DistributedStrategy()
dist_strategy.a_sync = True
dist_strategy.a_sync_configs = {"k_steps": 10000}

optimizer = fluid.optimizer.SGD(learning_rate=0.0001)

optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)
```

### 17.3.5 17.3.5 Server and Worker

The Fleet API provides a `fleet.init_server()` and `fleet.init_worker()` to help you configure the distributed training process. The `fleet.init_server()` function is used to initialize the server, and the `fleet.init_worker()` function is used to initialize the worker.

```
if fleet.is_server():
 fleet.init_server()
 fleet.run_server()
else:
 fleet.init_worker()
 trainer = X.Trainer(fluid.CPUPlace())
 trainer.fit(model, loader, epoch=10)
```

### 17.3.6 17.3.6 Running FleetX

The FleetX can be run on a single machine or a distributed machine. The `fleetrun` command is used to run FleetX. The `fleetrun` command is used to run FleetX.

```
fleetrun --server_num=1 --worker_num=2 ctr_app.py
```



## Knowledge Distillation

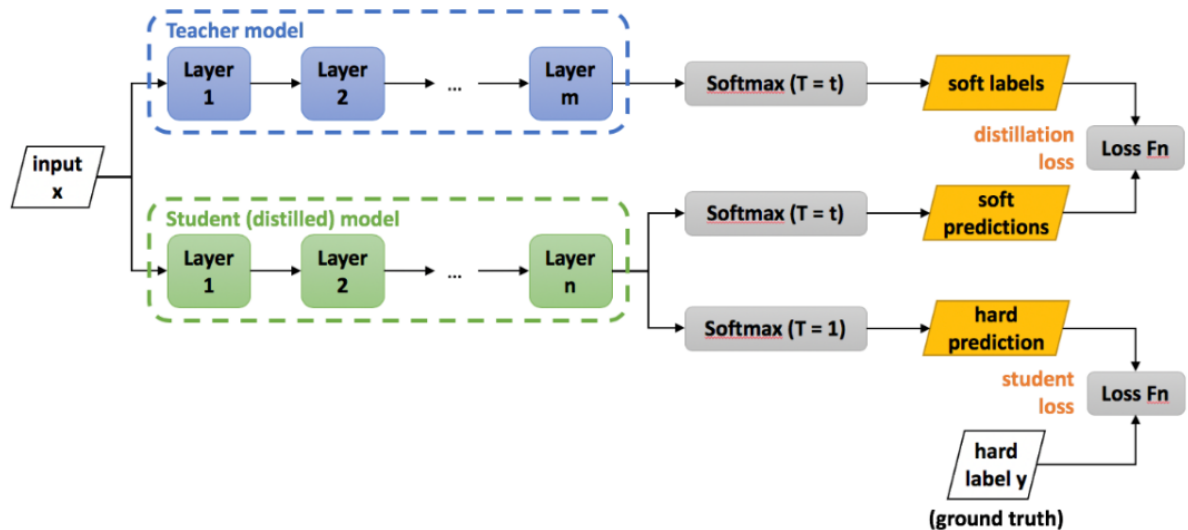
### 18.1 Knowledge

#### 18.1.1 Knowledge Distillation

Knowledge distillation is a technique for transferring knowledge from a large, pre-trained model (the teacher) to a smaller, less complex model (the student). The goal is to make the student model perform as well as the teacher model, but with a smaller footprint.

The teacher model is trained on a large dataset and produces soft labels (probability distributions over classes). The student model is trained on the same dataset and produces hard labels (class predictions). The distillation loss is calculated between the soft labels of the teacher and the soft predictions of the student.

- Knowledge distillation is a technique for transferring knowledge from a large, pre-trained model (the teacher) to a smaller, less complex model (the student).
- The teacher model is trained on a large dataset and produces soft labels (probability distributions over classes). The student model is trained on the same dataset and produces hard labels (class predictions). The distillation loss is calculated between the soft labels of the teacher and the soft predictions of the student.



## 18.1.2 æIJ■āŁaąđNèŠyéeRèó■czČ

æIJ■āŁaąđNèŠyāŠNāEūāzŪāyÿèġAèŠyéeRæŪzāijRčŽDārzaēfT:

- **czczčēēŠyéeRèó■czČ:** āĒLā;ŁčTīTeacherāŁaąđNāAŽæŌlčRĒāzūārEčzŠæđIJāŁā■ŸāIJlččAčZŸāy■iij
- **āyÿèġDèŠyéeRèó■czČ:** āyÿèġDèŠyéeRèó■czČæŸræNĠārE Teacher  
 āŁaąđNāŠN Student āŁaąđNæT;āĒēāRŊāyĀč;ŠčzIJāy■iijNāZžāōŽ Teacher  
 āŁaąđNāRČæTŕāRġāAŽāL■āRŠiijNStudent āŁaąđNāLŽæ■čāyÿāAŽāR■āRŠaijāæŠēō■czČāĀČēŁZāzš  
 ā■TēŁŽčġ■æŪzāijRāyN Student āŁaąđNčŽDèó■czČāōNāĒlā;ĪēŮ Teacher  
 āŁaąđNīijNStudent āŁaąđNēēAč■L Teacher āŁaąđNē;ŠāGžāyĀāyġ batch  
 čŽDæŌlčRĒčzŠæđIJæL■āRfāzēēō■czČīijNēĀN teacher āŁaąđNāzšēēAč■L  
 Student ēō■czČāōNāyĀāyġ batchiijNæL■ēČ;āijĀāġNāyNāyĀāyġ batch  
 čŽDæŌlčRĒiijNārzaēTŕā;ŠčŽDèó■czČēĀšāžæIJLāyĀāōZčŽDā;śāŠ■āĀČ
- **æIJ■āŁaąđNèŠyéeRèó■czČ:** æŸrāšžāžŌ Elastic Deep Learning  
 æRŔāGžčŽDāyĀčġēō■czČæŪzæāŁāĀČāōČārETeacherāŁaąđNāŠNStudent-  
 tēāŁaąđNēġcēĀēiijNTeacherāŁaąđNēcēČlč;šāyžčžāyŁæŌlčRĒæIJ■āŁāiijNStudentāŁaąđNāLŽāzēāōc

## 18.1.3 æIJ■āŁaąđNèŠyéeRèó■czČæTūčZŁ

- **ēŁČčžæŸ;ā■ŸēŮDæžRīijŽ** čTšāžŌStudentāŁaąđNāŠNTeacherāŁaąđNčŽDēġcēĀēiijNæLĀāzēæIJ■āŁā
- **æRŔā■Ġēō■czČēĀšāžēiijŽčTšāžŌēŁČčžæžEæŸ;ā■ŸēŮDæžRīijNēŁZæāūāršāRfāzēā;Ł**StudentāŁaąđN  
 sizeiijZāRŊæŪūčTšāžŌStudentāŁaąđNāŠNTeacherāŁaąđNæŸrāijČæđDætAæt'čžēiijNStudentāŁaąđN
- **æRŔēnŸēō■czČēŮDæžRāLŮčTlčŌĠiijŽ**æŁšāzŋāRfāzēārETeacherāŁaąđNēČlč;šāLŕčžāyŁčŽDāijzæ.
- **æRŔā■Ġēō■czČæTlčŌĠiijŽčTlāLūāRfāzēæāžæ■ō**TeacherāŠNStudentčŽDāRđāRŔæĀġēČ;čAæt'zē

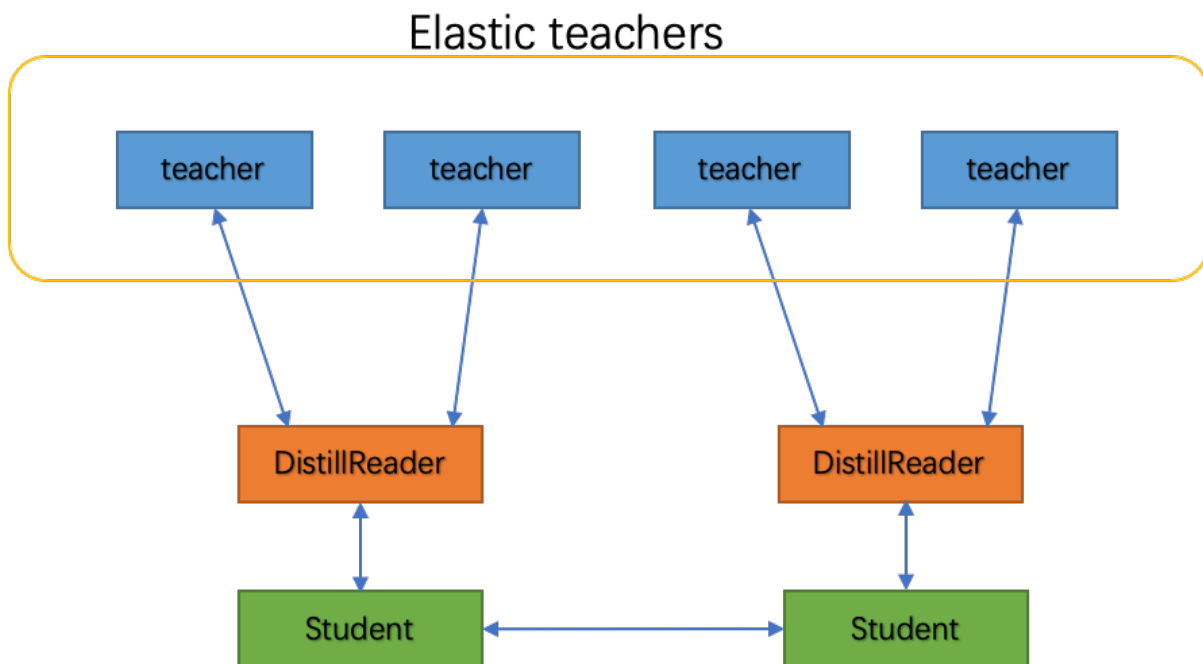
## 18.2 EDL æIJ■āŁaąđNāijzæĀġèŠyéeRæTlæđIJ

ResNet50\_vdāŁaąđN, ImageNet æTŕæ■ōēZE

## 18.3 æIJ■āŁaąđNāijzæĀġèŠyéeR

### 18.3.1 DistillReader

æIJ■āŁaąđNāijzæĀġèŠyéeRčŽDæāyāŁČæŸrārETeacherāŁaąđNēČlč;šāLŔāžEæIJ■āŁāčīriijNēĀNStudent  
 ārETeacherāŁaąđNēcēČlč;šāyžāIJlčžēāRŔāōžēTŽāijzæĀġæIJ■āŁā, āIJlS-  
 tudēāŁaąđNāyĀā;ġāLŽēĀŽēŁĠ DistillReader æĪēārAēēĒStudent-  
 tēāŁaąđNāyŌTeacherāŁaąđNāzNēŮŕčŽDēĀŽāfāiijNēōŁēŪōTeacheræIJ■āŁāĀČ





```
cd example/distill/resnet

wget --no-check-certificate https://paddle-edl.bj.bcebos.com/
→distill_teacher_model/ResNeXt101_32x16d_wsl_model.tar.gz
tar -zxvf ResNeXt101_32x16d_wsl_model.tar.gz

python -m paddle_serving_server_gpu.serve \
 --model ResNeXt101_32x16d_wsl_model \
 --mem_optim True \
 --port 9898 \
 --gpu_ids 1
```

### 18.4.3 18.4.3 Student 18.4.3 Student

æCäyNäS;äzd'âIJl0âRûGPUââRfâLÍStudentæIqadNëöçzÇ  
 äËüäy■train\_with\_fleet.pyæYrcTlâžŌâRfâLleöçzÇZDeDŽæIJnñijNçTlæLûéIJÄèçAâIJlâËüäy■æûzâLæŠy

```
python -m paddle.distributed.launch --gpus 0 \
 ./train_with_fleet.py \
 --model=ResNet50_vd \
 --data_dir=./ImageNet \
 --use_distill_service=True \
 --distill_teachers=127.0.0.1:9898
```

## 18.5 æŌlé■RéYĚèrz:

EDLæIJ■âLqadNâijzæÄgèŠyèçRGithub éçðæaIçšèèrEèŠyèçR

## CHAPTER 19

# Deep Gradient Compression

Deep Gradient Compression (DGC) is a technique for reducing the communication overhead of distributed training. It achieves this by compressing the gradients using a combination of quantization and sparsification. The quantization step involves mapping the continuous gradient values to a discrete set of values, while the sparsification step involves zeroing out a portion of the gradient elements. This process is repeated for each iteration of the training loop.

## 19.1 DGC

### 19.1.1 DGC

The DGC algorithm consists of several key steps: 1. Gradient Calculation: Compute the gradients for each parameter. 2. Quantization: Map the gradients to a discrete set of values. 3. Sparsification: Zero out a portion of the gradient elements. 4. Communication: Send the compressed gradients to the master node. 5. Aggregation: Receive the gradients from the workers and aggregate them. 6. Update: Update the parameters based on the aggregated gradients. The quantization step is performed using a uniform quantization scheme, where the gradient values are mapped to a fixed number of discrete levels. The sparsification step is performed using a random sampling process, where a subset of the gradient elements are zeroed out. This process is repeated for each iteration of the training loop.

### 19.1.1.1 DGC

The DGC algorithm is implemented in the following code snippet:

### 19.1.2 érŢélŊæŢĽædĬJ

- æĭađŊiijŽFasterRCNN
- ɕaŋäzŭiijŽ P40äyd' æĬJžāĽĒäŷĈäijŖiijŊæŕŔāŖæĬJžāŽĭäŷĀā■äiijŊTCPɕ;ŚɕzĬJætŊërŢāĀĆ
- áŔŰ300-700æ■ēēĀŬæŬŭ/400stepāĀĆ
- ɕſ;āžæŬäæ■ſāĀĆ

### 19.1.3 DGC áŎſɕŖĒɕóĀäzŊ

èĤŽéĠŊāŕĒɕóĀā■ŢāzŊɕz■äzŊɕz■Fleet DGC äŷ■ɕŽDäŷĀäzŽāŎſɕŖĒæſŊāŕzāžŢāŖĆæŢŕāzŢèŕæĈä;Ţ

#### æĉŕāžɕĭĀɕŬŖ

DGCŽŽDāſžæĬJŋæĀĭèŭŕæŶŕéĀŽèĤĠāŖĭäijæĀĀéĠ■ēɕAæĉŕāžɕiijŊā■ſāŖĭāŖŚéĀĀāđ' gāžŎɕzŽāōŽéŶĬ  
æ■cāŷĭèġſāžɕiijŊāzŎɕŖĒèōžā;ĭæ■ōäŷĽæĭɕĬJŊiijŊāſĀéĈĭæĉŕāžɕŦŦŕāĽāɕ■ĽāŖŊāzŎéŽŖæŬŭéŬŦ' æŎĭɕgžā  
sizeiijŊiijĽDGCɕŽŷā;ſāžŎæŕŖäŷĀäŷĭæĉŕāžæĬJĽ'èĠāŭſɕŽĎbatch sizeiijĽ'āĀĆ

āĀĠèō; NæŶŕèō■ɕzĈèĽĈɕĈzäŷĭæŢŕ, bāŷžā■Ţā■ābatch  
sizeiijŊāſĀéĈĭæĉŕāžɕŦŦŕāĽāŖŕāzèècŋèōđ' äŷžbatch sizeäzŎNbācđāđ' gāŷžNbŦiijŊāĒŭäŷ■ŢæŶŕäŷd' æŋæŽ  
[1] éĉĎĈ■ēŕĈāŖĆ ^^^^^^^

āržāžŎæ■cāŷŷɕŽĎèō■ɕzĈiijŊā;ĤɕŦĬDGCäŷĀèĽŋéĬJĀèĤŽæŋéĉĎĈ■èō■ɕzĈiijŊāŖæĀĽāŖŕèĈ;äijŽæĬ

```
1.
→ äžē1252äŷĭstepäŷžäŷĀäŷĭepochiijŊāĽ'■2äŷĭepochsä;ĤɕŦĭā■cāŷŷdenseéĀŽäſäiijŊāŖŎ3ä
→ 9%
strategy.dgc_configs = {
 "rampup_begin_step": 1252*2,
 "rampup_step": 1252*3,
 "sparsity": [0.984375, 0.996, 0.999]
}
2. āĽ'■éĭc4äŷĭepochséĈ;ä;ĤɕŦĭĬdenseéĀŽäſäiijŊāzŊāŖŎéžŶèōđ' 0.
→ 999ɕĭĀĈŬŖāžæĤŖèāŊ
strategy.dgc_configs = {
 "rampup_begin_step": 1252*4,
 "rampup_step": 1,
 "sparsity": [0.999]
}
```

āržāžŎFine-tuningèō■ɕzĈiijŊāŖŕæŬäéĬJĀéĉĎĈ■èō■ɕzĈiijŊāzŎɕŋŋ0äŷĭepochɕŽŦ' æŎä;ĤɕŦĬDGCā■ſā

```
äžŎɕŋŋ0æ■ēäiijĀāġŊDGCĭĀĈŬŖéĀŽäſä
strategy.dgc_configs = {
 "rampup_begin_step": 0,
 "rampup_step": 1,
 "sparsity": [0.999]
}
```







## DGC 配置

配置DGC的步骤如下：

- `rampup_begin_step` (int) 设置DGC开始训练的步数
- `rampup_step` (int) 设置DGC训练的步数。默认值为100。在0~19 steps 时 `sparsity=0.75`；在20~39 steps 时 `sparsity=0.9375`；在40~59 steps 时 `sparsity=0.996`；在60~79 steps 时 `sparsity=0.999`。
- `sparsity` (list[float]) 设置DGC训练的步数，(1 - current sparsity) % 设置DGC训练的步数。

```
dist_strategy = fleet.DistributedStrategy()

dist_strategy.dgc = True
dist_strategy.dgc_configs = {
 "rampup_begin_step": 1252*2,
 "rampup_step": 1252*3,
 "sparsity": [0.984375, 0.996, 0.999]
}

optimizer = fluid.optimizer.Momentum(learning_rate=0.01, momentum=0.9)
optimizer = fleet.distributed_optimizer(optimizer, dist_strategy)
optimizer.minimize(model.loss)
```

## 配置GPU

配置GPU的步骤如下：

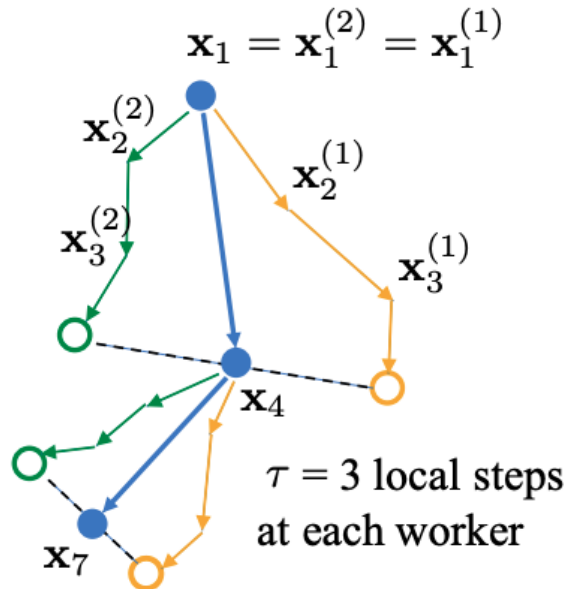
```
place = fluid.CUDAPlace(int(os.environ.get('FLAGS_selected_gpus', 0)))
exe = fluid.Executor(place)
exe.run(fluid.default_startup_program())

for i, data in enumerate(loader()):
 start_time = time.time()
 cost_val = exe.run(model.main_prog,
 feed=data,
 fetch_list=[model.loss.name])

 end_time = time.time()
 print(
 "worker_index: %d, step%d cost = %f, speed: %f"
 % (fleet.worker_index(), i, cost_val[0], batch_size / (end_time - start_time)))
```



čDúāŘŌéZEç; d'äy■čŽDæL' ÅæIJL' trainer äijŽèfZèaÑéĀŽāfaiijNāRÑæ■ëijLaveragingijL'æL' ÅæIJL' trainers äyŁçŽDāRCæTřāĀCāyĀäyĹāRÑ trainersijNāRÑæ■éŮt' éŽŽäyž3 æ■ééTfijLiterationsijL' çŽDLocal SGDèfGčĹNāeČāyNāZĹæL' Āčd' žāĀCézDčžfäyd' æĹæùrāĹDèāĹčd' žäyd' ā trainers āRĐèGĹčŽD Local SGD æŽt' æŮřèfGčĹNijNäy■éŮt' çŽDèŠĹèL' šèùrāĹDèāĹčd' žāRÑæ■ēāŘŌçŽDæĹāā



Local SGDäy■čŽDäyĀäyĹāĒšéTŏéŮŏéCŸæŸrāeČä;TřāŏāŏŽāRCæTřāRÑæ■ēçŽDèŮt' éŽT(éCŠçŌĞ)ijN

- āčđād' gāRCæTřāRÑæ■ēçŽDèŮt' éŽTāRfāzēāGRārš trainers éŮt' éĀŽāfāzūēfšçŽDā;śāš■æRŘénŸèŏ■çzČāRđāRŘ.
- äĹāčđād' gāRÑæ■ééŮt' éŽTāRřèČ;äijŽéĀāēĹRæIJĀçzĹèŏ■çzČçšĹāžèçŽDæ■šād' śāĀC [1]

äžēäyNäyd' äyĹç■ŮçTēäzŌäy■āRÑègŠāžèērTāZĹèĹ;āĹræŽt' äē;çŽDāzšēāaiijŽ

- **post Local SGD** āřēŏ■çzČèfGčĹNāĹæĹRäyd' äyĹéŸūæŏtĥijŽçññäyĀéŸūæŏt wokers éŮt' āRÑæ■ēçŽDèŮt' éŽTäyž 1 äyĹæ■ēéTfijNāšāRÑæ■ēSGDijNāĹēāĹērAæIJĀçzĹèŏ■çzČçšĹāžèijN HijNāĹēāRŘā■Gèŏ■çzČāRđāRŘāĀC
- **Adaptive Communication Local SGD** éĀŽèfGāĹĹāēĀAçŽDèřCæTřt' āRCæTřāRÑæ■ēçŽDèŮt' éŽTāĹēār

Fleet äy■āŏđçŌřāžE post Local SGD āŠÑ Adaptive Communication Local SGD äyd' çg■ç■ŮçTēāĀC äy■äyNāēŮGārEçzŽāGž Fleetäy■ Local SGD çŽDāŏđēūāTĹædIJijNāzūēĀŽèfGāyĀäyĹçŏĀā■TäĹNā■RāzNçz■āeČä;TřāIJĹFleet äy■ā;ĹçTĹ Local SGDāĀC

## 19.2.2 èřTéĹNæTĹædIJ

èřTéĹNèŏĹç;ŏ

| model    | dataset  | local size | batch | cluster      | dtype | warming up | learning rate decay |
|----------|----------|------------|-------|--------------|-------|------------|---------------------|
| resnet50 | ImageNet | 128        |       | 4 x 8 x V100 | FP32  | 30         | polynomial          |

Table 1: Training configurations for the ResNet50 model on the ImageNet dataset.

| local step | qps     | acc1   | acc5   |
|------------|---------|--------|--------|
| 1          | 8270.91 | 0.7579 | 0.9266 |
| 2          | 8715.67 | 0.7533 | 0.9265 |
| 4          | 8762.66 | 0.7551 | 0.9260 |
| 8          | 9184.62 | 0.7511 | 0.9239 |
| 16         | 9431.46 | 0.7429 | 0.9206 |
| ADACOMM    | 8945.74 | 0.7555 | 0.9270 |

Table 2: Performance comparison of Local SGD and ADAPTIVE COMMUNICATION (ADACOMM) on the ResNet50 model. The table shows the number of queries per second (qps), accuracy at 1% (acc1), and accuracy at 5% (acc5) for different local batch sizes (1, 2, 4, 8, 16) and for the ADACOMM configuration.

### 19.2.3 Local SGD

The Local SGD configuration is used for training the ResNet50 model. The configuration is defined in the `fleetx` module. The `Local SGD` configuration is used for training the ResNet50 model. The configuration is defined in the `fleetx` module. The `Local SGD` configuration is used for training the ResNet50 model. The configuration is defined in the `fleetx` module.

Figure 1: Local SGD configuration.

```
import os
import fleetx as X
import paddle
import paddle.fluid as fluid
import paddle.distributed.fleet.base.role_maker as role_maker
import time
import paddle.distributed.fleet as fleet
```

Figure 2: Local SGD configuration.

The `Local SGD` configuration is used for training the ResNet50 model. The configuration is defined in the `fleetx` module. The `Local SGD` configuration is used for training the ResNet50 model. The configuration is defined in the `fleetx` module.

```
paddle.enable_static()
configs = X.parse_train_configs()
fleet.init(is_collective=True)
```





## CHAPTER 20

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éčđæąĺáŁĘąŸČǎıjŘèő■çžČǎšžçžŁæŁěǎŚŁ

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# CHAPTER 21

## Resnet50æĀğèĈjåšžåĜĖ

Resnet50æŸřå;ŠåL'■èğĖèğL'écĖåššæřTè;ĈéĀŽçTlçZĎécĎèð■çzĈæłåđNåRŌçñřijNåRÑæUũázšä;IJå

### 21.1 èjřçañäzúéĚ■çjőæĈĚåĖt

#### 21.1.1 åšžæIJñçL'ŁæIJñäĖæAř

|                          |                                            |
|--------------------------|--------------------------------------------|
| èjřçañäzúéĚ■çjőæĈĚåĖt    | åĚũä;ŠéĚ■çjő                               |
| åđđä;ŇçszådŇ             | çŽ;åžęX-Man 2.0                            |
| å■Tåđđä;ŇGPU             | 8x NVIDIAĀđ TeslaĀđ V100                   |
| æŠ■ä;IJçszçzš            | Ubuntu 16.04 LTS with tests run via Docker |
| CPU                      | Intel(R) Xeon(R) Gold 6271C CPU @ 2.60GHz  |
| åĖĚå■Ÿ                   | 512G                                       |
| CUDA / CUDNNçL'ŁæIJñ     | 10.1 / 7.6.5                               |
| NCCL / DALI çL'ŁæIJñ     | 2.4.7 / 0.24.0                             |
| åd'ŽGPUåđđä;ŇäzšĚåTăĖæAř | InfiniBand 100 Gb/sec                      |
| Paddle Github Commit     |                                            |
| FleetX Github Commit     |                                            |
| çañçŽŸçszådŇ             | æIJñåIJřSSDçañçŽŸ                          |
| æTřæ■őéŽĖ                | ImageNet                                   |
| èřĎäijřæłåđŇ             | Resnet50                                   |
| åd'■çŌřäzççåAåIJřåĬ      | <a href="#">Resnet50-Benchmark</a>         |
| PythonçL'ŁæIJñ           | 3.7                                        |

## 21.1.2 çañäzúæNšæL'S

```
nvidia-smi topo -m
```

| GPU0    | GPU1         | GPU2 | GPU3 | GPU4 | GPU5 | GPU6 | GPU7 |     |
|---------|--------------|------|------|------|------|------|------|-----|
| →mlx5_0 | CPU Affinity |      |      |      |      |      |      |     |
| GPU0    | X            | NV2  | NV2  | NV1  | NV1  | NODE | NODE |     |
| →NODE   | NODE         | 0-23 |      |      |      |      |      |     |
| GPU1    | NV2          | X    | NV1  | NV1  | NODE | NV2  | NODE |     |
| →NODE   | NODE         | 0-23 |      |      |      |      |      |     |
| GPU2    | NV2          | NV1  | X    | NV2  | NODE | NODE | NV1  |     |
| →NODE   | NODE         | 0-23 |      |      |      |      |      |     |
| GPU3    | NV1          | NV1  | NV2  | X    | NODE | NODE | NODE | NV2 |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| GPU4    | NV1          | NODE | NODE | NODE | X    | NV2  | NV2  | NV1 |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| GPU5    | NODE         | NV2  | NODE | NODE | NV2  | X    | NV1  | NV1 |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| GPU6    | NODE         | NODE | NV1  | NODE | NV2  | NV1  | X    | NV2 |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| GPU7    | NODE         | NODE | NODE | NV2  | NV1  | NV1  | NV2  | X   |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| mlx5_0  | NODE         | NODE | NODE | NODE | NODE | NODE | NODE |     |
| →NODE   | X            |      |      |      |      |      |      |     |

Legend:

X = Self

SYS = Connection traversing PCIe as well as the SMP interconnect, between NUMA nodes (e.g., QPI/UPI)

NODE = Connection traversing PCIe as well as the interconnect, between PCIe Host Bridges within a NUMA node

PHB = Connection traversing PCIe as well as a PCIe Host Bridge (typically the CPU)

PXB = Connection traversing multiple PCIe switches (without traversing the PCIe Host Bridge)

PIX = Connection traversing a single PCIe switch

NV# = Connection traversing a bonded set of # NVLinks

## 21.2 æĀğèČ;ætĴNèrĴæÚzæšĴ

- çañäzúætĴDæžŘ éĠĠčĴlād'ŽæIJžād'ŽææøçzČrijNäzēāōdā;NæĴř x  
āĴāōdā;NGPUāāæĴřā;IJäÿžerDäzūæāĠĠĠēijNèrDäzū 1 x 1, 1 x 8, 2 x  
8, 4 x 8æČĒæĴäÿNçŽDæĀğèČ;āšžāĠĠæĀČ
- èøçzČēūĒāRCæĴř æL'zéĠRād'ġārRijĴBatch SizeijL'āržèøçzČæĀğèČ;ā;šāšæIJĀād'ġrijNāZæād'
- æĴNèrĴæNĠæāĠèŌūāRŪæÚzæšĴ ā;šāL'äÿzæĴAçŽDæūsāžæāçäzāæĴæđūéĀŽäÿÿéĠĠčĴlāijČæææ



äyžāžÈŌũā; ŪāŽt' äē; çŽDæĀgēČ; iijNæĹSāžnéžYēōd' æL'ŠāijĀāžĒDALēfZēaŊæTṛa■ōIOiijNēfZēG.

| batch / node | 1 x 1  | 1 x 8   | 2 x 8    | 4 x 8    |
|--------------|--------|---------|----------|----------|
| 32           | 666.38 | 4467.82 | 8711.69  | 19107.42 |
| 64           | 761    | 6148.98 | 12076.77 | 24314.58 |
| 128          | 890.03 | 6793.73 | 13514.66 | 27277.36 |
| 256          | 938.57 | 7305.66 | 14599.55 | 29361.24 |

## CHAPTER 22

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BertæĺađÑèő■çžČæǺğèČĵå\$žçž£

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## CHAPTER 23

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TransformeræĺąåđŃæĂğèĈĵå\$žçž£

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## CHAPTER 24

### VGG16æġađNèő■çzČæĀğèČġāšžčžĚ

VGG16æŸřā;ŠāL■èğEèğL'écEāššæfTè;ČéĀŽčTlčZDécDèő■çzČæġađNāŘŌčńřġNāŘNæŮüázšā;IJāy

#### 24.1 èġřčāñāžúéĚ■çġőæČĚāEġ

##### 24.1.1 āšžæIJñçL'ŁæIJñăĚqæAř

|                           |                                            |
|---------------------------|--------------------------------------------|
| èġřčāñāžúéĚ■çġőæăĢ        | ăĚŮăġŠéĚ■çġő                               |
| ăőđăġNçszăđN              | çŽġăžęX-Man 2.0                            |
| ă■TăőđăġNGPU              | 8x NVIDIAĀő TeslaĀő V100                   |
| æŠ■ăġIJçszčžš             | Ubuntu 16.04 LTS with tests run via Docker |
| CPU                       | Intel(R) Xeon(R) Gold 6271C CPU @ 2.60GHz  |
| ăĚĚă■Ÿ                    | 512G                                       |
| CUDA / CUDNNçL'ŁæIJñ      | 10.1 / 7.6.5                               |
| NCCL / DALI çL'ŁæIJñ      | 2.4.7 / 0.24.0                             |
| ăđ'ŽGPUăőđăġNăžŠèAřăăqæAř | InfiniBand 100 Gb/sec                      |
| Paddle Github Commit      | b6711e24ea5e17fa24e9b1ba516fe03186dd4a2c   |
| FleetX Github Commit      | 296e0d71a33a8d250f6626733bc2535465f5f6e0   |
| čāñçŽŸçszăđN              | æIJñăIJřSSDçāñçŽŸ                          |
| æTřæ■őéŽĚ                 | ImageNet                                   |
| èřDăġjřăġađN              | VGG16                                      |
| ăđ'■çŌřăžčçăAăIJřăĬĂ      | <a href="#">VGG16 -Benchmark</a>           |
| PythonçL'ŁæIJñ            | 3.7                                        |

## 24.1.2 çāñāzúæNšæL'S

```
nvidia-smi topo -m
```

| GPU0    | GPU1         | GPU2 | GPU3 | GPU4 | GPU5 | GPU6 | GPU7 |     |
|---------|--------------|------|------|------|------|------|------|-----|
| →mlx5_0 | CPU Affinity |      |      |      |      |      |      |     |
| GPU0    | X            | NV2  | NV2  | NV1  | NV1  | NODE | NODE |     |
| →NODE   | NODE         | 0-23 |      |      |      |      |      |     |
| GPU1    | NV2          | X    | NV1  | NV1  | NODE | NV2  | NODE |     |
| →NODE   | NODE         | 0-23 |      |      |      |      |      |     |
| GPU2    | NV2          | NV1  | X    | NV2  | NODE | NODE | NV1  |     |
| →NODE   | NODE         | 0-23 |      |      |      |      |      |     |
| GPU3    | NV1          | NV1  | NV2  | X    | NODE | NODE | NODE | NV2 |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| GPU4    | NV1          | NODE | NODE | NODE | X    | NV2  | NV2  | NV1 |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| GPU5    | NODE         | NV2  | NODE | NODE | NV2  | X    | NV1  | NV1 |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| GPU6    | NODE         | NODE | NV1  | NODE | NV2  | NV1  | X    | NV2 |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| GPU7    | NODE         | NODE | NODE | NV2  | NV1  | NV1  | NV2  | X   |
| →       | NODE         | 0-23 |      |      |      |      |      |     |
| mlx5_0  | NODE         | NODE | NODE | NODE | NODE | NODE | NODE |     |
| →NODE   | X            |      |      |      |      |      |      |     |

Legend:

X = Self

SYS = Connection traversing PCIe as well as the SMP interconnect between NUMA nodes (e.g., QPI/UPI)

NODE = Connection traversing PCIe as well as the interconnect between PCIe Host Bridges within a NUMA node

PHB = Connection traversing PCIe as well as a PCIe Host Bridge (typically the CPU)

PXB = Connection traversing multiple PCIe switches (without traversing the PCIe Host Bridge)

PIX = Connection traversing a single PCIe switch

NV# = Connection traversing a bonded set of # NVLinks

## 24.2 æĀğèČ;ætNèrTæÚzæšT

- çāñāzúætDæžŘ éĜĜčTlād'ŽæIJžād'ŽāæèőčzČrijNāzēāōdā;NæTř x  
āTāōdā;NGPUāæTřā;IJäÿžerDäzūæāĜāĜEijNerDäzū 1 x 1, 1 x 8, 2 x  
8, 4 x 8æČĒāEtäÿNčŽDæĀğèČ;āšžāĜEāĀĆ
- èőčzČeūĒāRČæTř æL'zéĜRād'ġārRijLBatch SizeijL'āržeőčzČæĀğèČ;ā;šāšæIJĀād'grijNāZæād'  
ijNāšäÿæĀČèŽŠæTūæTžčŽDāržærTāĀĆ



- ## 24.3 åŹǻĜÆtÑèŦçzŞæđĲ

- ```
exec_strategy = fluid.ExecutionStrategy()
dist_strategy = fleet.DistributedStrategy()
exec_strategy.num_threads = 2
exec_strategy.num_iteration_per_drop_scope = 100
dist_strategy.execution_strategy = exec_strategy
build_strategy = fluid.BuildStrategy()
build_strategy.enable_inplace = False
build_strategy.fuse_elewise_add_act_ops = True
build_strategy.fuse_bn_act_ops = True
dist_strategy.build_strategy = build_strategy
dist_strategy.nccl_comm_num = 1
```

- `Images/sijNajLçTlèGlàLlæuũãRLçşçâzeAutomaticMixedPrecision(AMP)èfZèaÑèõçÇiijÑDistributedStrategyæCâyNiiž`

```
import paddle
import paddle.distributed.fleet as fleet

dist_strategy = fleet.DistributedStrategy()
exec_strategy = fluid.ExecutionStrategy()
exec_strategy.num_threads = 2
exec_strategy.num_iteration_per_drop_scope = 100
dist_strategy.execution_strategy = exec_strategy
build_strategy = fluid.BuildStrategy()
build_strategy.enable_inplace = False
build_strategy.fuse_elewise_add_act_ops = True
build_strategy.fuse_bn_act_ops = True
dist_strategy.build_strategy = build_strategy
dist_strategy.amp = True
dist_strategy.nccl_comm_num = 1
```

24.3. ǎšžǎĜEætNèrTçzSædIJ

- `Images/s, DistributedStrategy`

```
import paddle.distributed.fleet as fleet
dist_strategy = fleet.DistributedStrategy()
dist_strategy.auto = True
```

batch / node	1 x 1	1 x 8	2 x 8	4 x 8
32	409.68	3044.60	4840.74	7668.70
64	455.98	3395.67	6525.20	12237.04
128	472.81	3587.29	7019.13	13562.80
256	407.88	3154.15	6217.92	12147.46

CHAPTER 25

Word2vec

word2vec is a neural network architecture for learning word embeddings. It is based on the idea of predicting missing parts of words given the context. The most common implementation is the Skip-gram model, which takes a word and predicts the surrounding words. The other common implementation is the Continuous Bag-of-Words (CBOW) model, which takes the surrounding words and predicts the target word. Both models are trained on large corpora of text.

25.1 Introduction

25.1.1 Problem Statement

word2vec	word2vec
-billion	1-billion
Work2vec	Work2vec
Word 2vec-Benchmark	Word 2vec-Benchmark
Python	3.7

25.2 æǺǧèĈ;ætǔNèŕŤæŮzæſŤ

- çǎñǎžŭètǔDæžŔ éĜĜĉŤlǎd'ŽæIJžǎd'ŽèĚŽĉlNèő■çžĈiijNærŔǎŸǺǎŔŕæIJ■ǎŁǎǎŽlǎǎđǎ;NǎlĜǎŔŕǎŁlǎŸǺǎ
4, 8, 16, 32æĈĚǎĖŧǎŸNĉŽDæǺǧèĈ;ǎſžǎĜĖǎǺĈ
- èő■çžĈèŭĚǎŔĈæŤŕ æL'zéĜŔǎd'ğǎŕŔiijŁBatch SizeiijL'ǎŕzéő■çžĈæǺǧèĈ;ǎ;ſǎſ■æIJǺǎd'ğŕiijNǎŽǎæ■d'ǎ
- æŧNèŕŤæŤŔǎŤŔǎǧǎǧèŬǎŔŮæŮzæſŤ ǎ;ſǎL■ǎŸzæŧAçŽDæŭſǎžęǎ■ęǎžǎæǎĖǎđŭéǺŽǎŸŸéĜĜĉŤlǎijĈæ■ěæ

25.3 ǎſžǎĜĖætǔNèŕŤçžſǎedIJ

- ǎ■Ťǎ;■iijŽImages/siijNǎ;ĚĉŤlĉſ;ǎžęFP32iijNŤdistributedStrategyǎĖĈǎŸNŕiijŽ

```
import paddle
import paddle.distributed.fleet as fleet

dist_strategy = fleet.DistributedStrategy()
dist_strategy.a_sync=True
dist_strategy.a_sync_configs = {"k_steps": 100}
```

batch / node	4	8	16	32
100	55597.5	55082.37	53302.63	47280.91

CHAPTER 26

çŤíæŁŭFAQ

- TBA
-

FleetXä;£çŤíApache License 2.0âijĂæžŘâ■Rèőő