Cloud ML Engine

Google Cloud Platform (GCP)

2017.08

박찬성

before getting started..

1. do you have a GCP account? if no, just create one
    1. if you have a gmail account, you are ok to go
2. just read THIS, it will give a basic idea how everything works. You don't need to follow every instructions. just reading is sufficient
3. do some of the pre-requisite steps introduced like..

Set up your GCP project

1. Select or create a GCP project.

   GO TO THE MANAGE RESOURCES PAGE

2. 프로젝트에 결제가 사용 설정되어 있는지 확인하세요.

   결제 사용 설정 방법 알아보기

3. Enable the Cloud Machine Learning Engine and Compute Engine APIs.

   ENABLE THE APIS

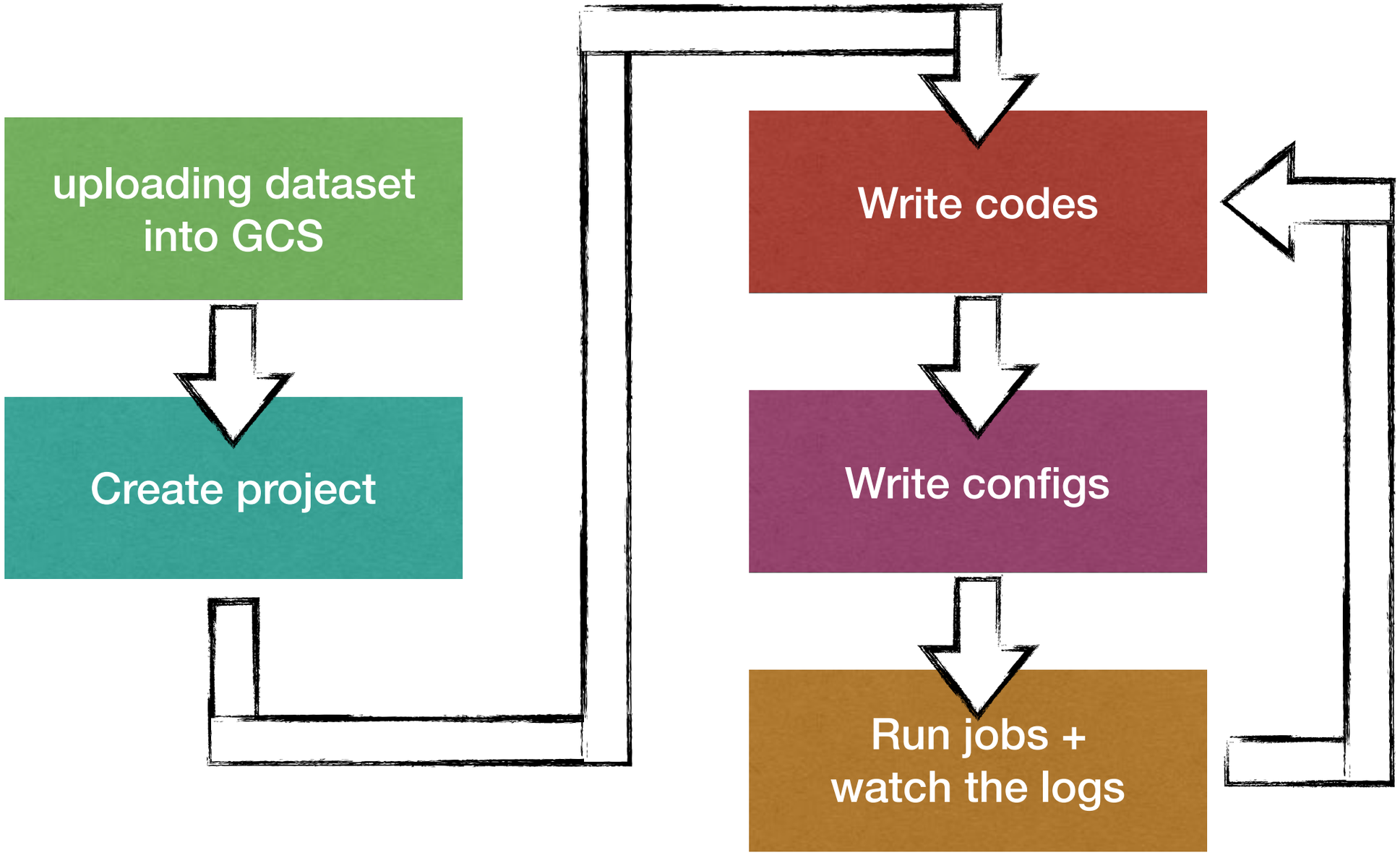
4. Set up authentication:
   a. Go to the **Create service account key** page in the GCP Console.

      GO TO THE CREATE SERVICE ACCOUNT KEY PAGE

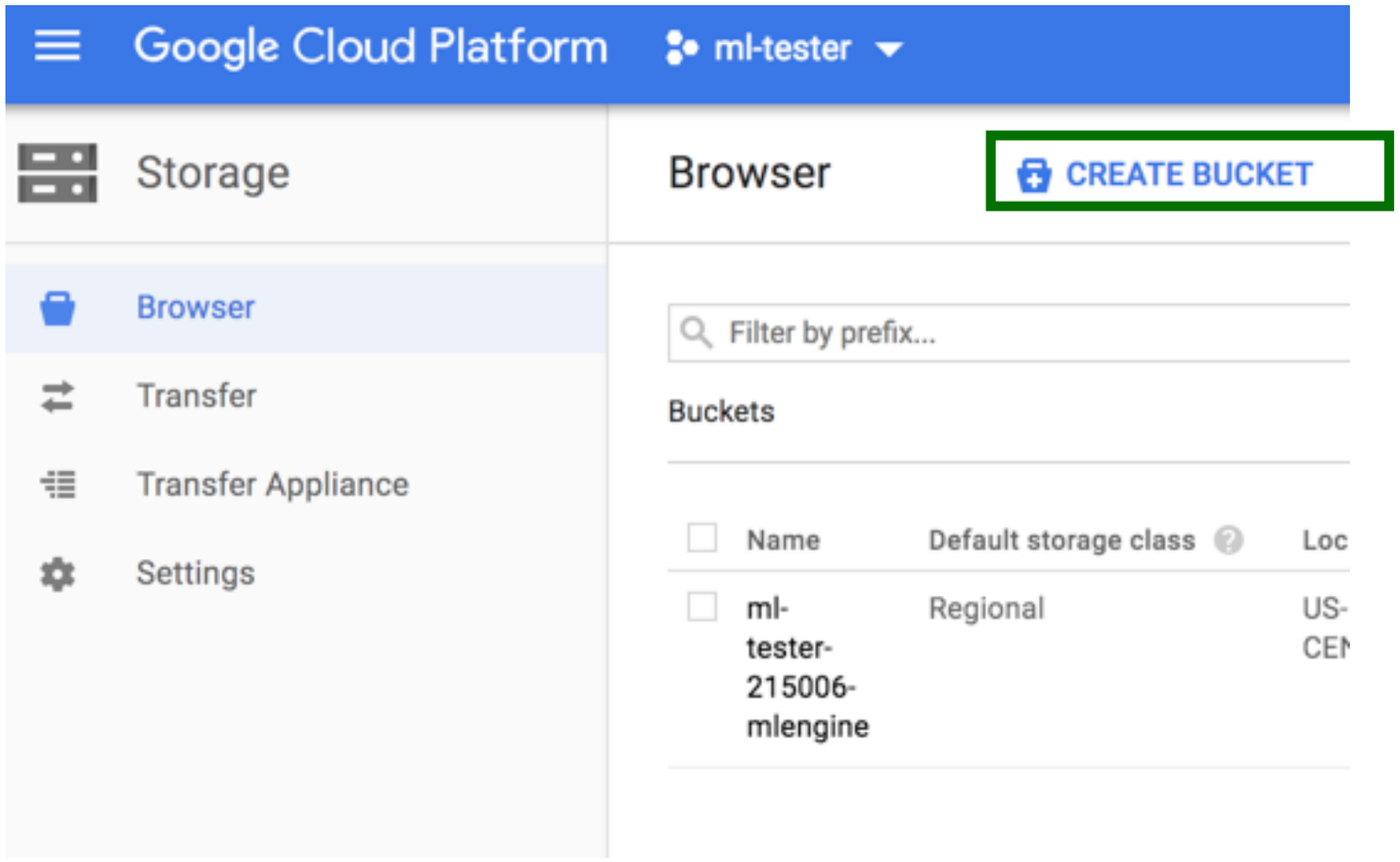
   b. From the **Service account** drop-down list, select **New service account**.
   c. Enter a name into the **Service account name** field.
   d. From the **Role** drop-down list, select **Project** > **Owner**.
   e. Click **Create**. A JSON file that contains your key downloads to your computer.
5. Set the environment variable *GOOGLE_APPLICATION_CREDENTIALS* to the file path of the JSON file that contains your service account key. This variable only applies to your current shell session, so if you open a new session, set the variable again.
   - **Example:** Linux or macOS
   - **Example:** Windows
6. Install and initialize the Cloud SDK.

basic flow introduction

1. create a bucket, and some folders in it
2. upload files to folders
3. create a project source tree
4. create a setup.py if you have any dependencies to install
5. create a config.yaml for your configuration of the project
6. create an empty __init__.py in every other folders except for the root path
7. write some codes
8. or copy some code files from local
    1. change print() function to logging to see a real-time logs
9. submit a job
10. watch the logs


uploading dataset into GCS
Create project
Write codes
Write configs
Run jobs + watch the logs

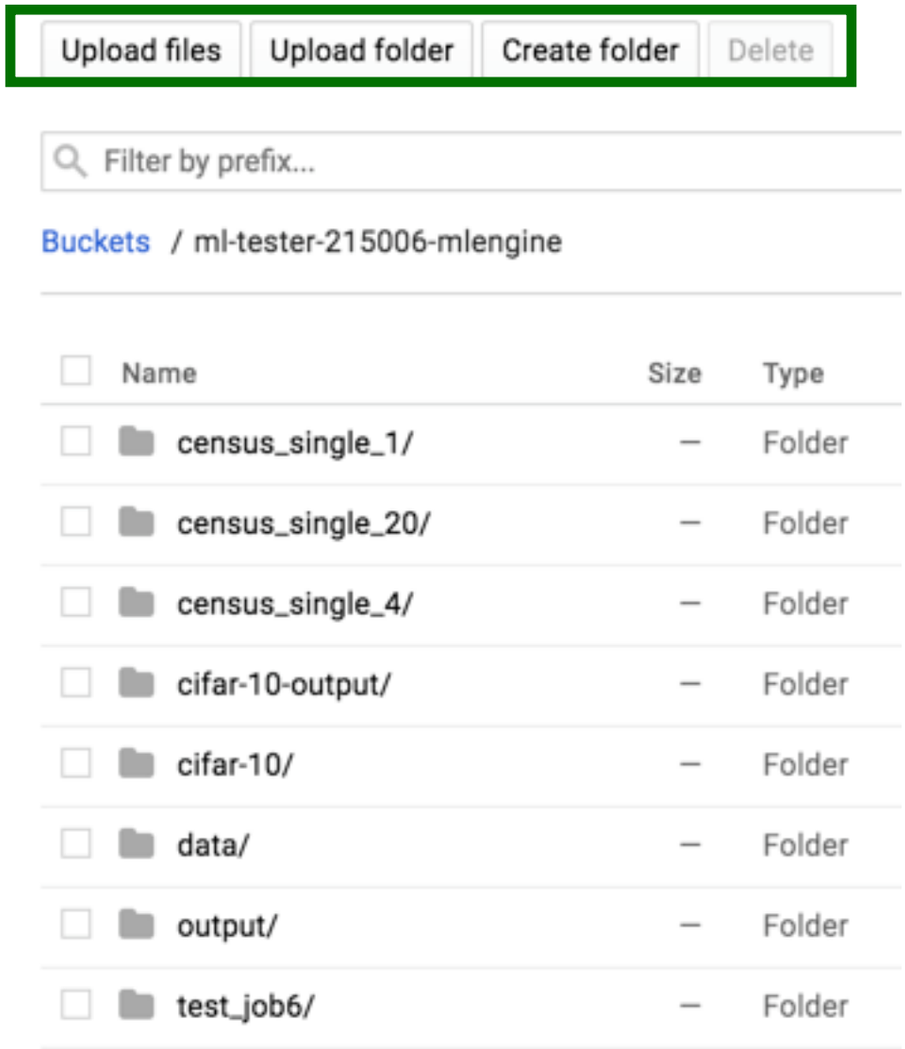
create GCS bucket



make a bucket
@ where cloud ML engine is available

bucket and cloud ML trainer should be placed @the same location

Create folder + Upload files



SUPER EASY

Project Structure

main application module

Create an `__init__.py` file in every subdirectory. These files are used by Setuptools to identify directories with code to package, and may be empty

Dependencies

- runtime versions shows packages that are already installed. You need to specify which version to use.
  * --runtime_version x.x
- standard dependencies can be defined in setup.py file

```
from setuptools import find_packages
from setuptools import setup

REQUIRED_PACKAGES = ['some_PyPI_package>=1.0']

setup(
    name='trainer',
    version='0.1',
    install_requires=REQUIRED_PACKAGES,
    packages=find_packages(),
    include_package_data=True,
    description='My training application package.'
)
```

find packages
of your interest
@PyPI

gcloud cli will automatically recognize the setup.py

configuration (1)

set option “--config" to specify the location of config.yaml

```
--config=config.yaml
```

see the machine types, and here

config.yaml example

```
trainingInput:
  scaleTier: CUSTOM
  masterType: complex_model_m
  workerType: complex_model_m
  parameterServerType: large_model
  workerCount: 9
  parameterServerCount: 3
  packageUris: gs://my/trainer/path/package-0.0.0.tar.gz
  pythonModule: trainer.task
  region: us-central1
  jobDir: gs://my/training/job/directory
  runtimeVersion: 1.10
  pythonVersion: 3.5
```

`standard_gpu`	A machine equivalent to standard that also includes a single NVIDIA Tesla K80 GPU. **Compute Engine machine name:** n1-standard-8 with one k80 GPU
$1.2118 *(2.4731)*	
`standard_p100`	A machine equivalent to standard that also includes a single NVIDIA Tesla P100 GPU. **Compute Engine machine name:** n1-standard-8-p100x1
$2.6864 *(5.4824)*	
`standard_v100`	A machine equivalent to a standard that also includes a single NVIDIA Tesla V100 GPU. The availability of these GPUs is in *Beta* launch stage. **Compute Engine machine name:** n1-standard-8-v100x1

Job directory (`jobDir`)

The path to a Cloud Storage location to use for job output.

configuration (2)

other common training parameters

```
gcloud ml-engine jobs submit training JOB --module-name=MODULE_NAME [--config=CONFIG]
  [--job-dir=JOB_DIR] [--labels=[KEY=VALUE,...]] [--package-path=PACKAGE_PATH]
  [--packages=[PACKAGE,...]] [--python-version=PYTHON_VERSION] [--region=REGION]
  [--runtime-version=RUNTIME_VERSION] [--scale-tier=SCALE_TIER]
  [--staging-bucket=STAGING_BUCKET] [--async | --stream-logs] [GCLOUD_WIDE_FLAG …]
  [-- USER_ARGS ...]
```

further description for each parameter

read/write file from GCS?

```
from tensorflow.python.lib.io import file_io
```

python standard library can’t recognize GCS location,
but, tensorflow provides file_io packages for it

```
bucket_name = 'ml-tester-215006-mlengine'
path = 'cifar-10'
filename = 'cifar10_preprocess_batch_' + str(batch_id) + '.p'
full_path = 'gs://' + bucket_name + '/' + path + '/' + filename
```

GCS location can be recognized by three parts,
“gs://” protocol, “bucket_name”, and “file_path”

```
from tensorflow.python.lib.io import file_io

with file_io.FileIO('gs://.....', mode='w+') as f:
    cPickle.dump(self.words, f)
```

Or you can read pickle file in like this:

```
file_stream = file_io.FileIO(train_file, mode='r')
x_train, y_train, x_test, y_test  = pickle.load(file_stream)
```

logging

- standard python’s print function shows up “AFTER” everything is done. This is a known buffering issue.

- If you want to check a real time logs, you need to use “logging” package instead.

```python
import logging

logger = logging.getLogger('simple_example')
logger.debug('debug message')
logger.info('info message')
logger.warn('warn message')
logger.error('error message')
logger.critical('critical message')
```

submitting a job (1)

with common parameters in option flags

```
gcloud ml-engine jobs submit training $JOB_NAME \
        --package-path $TRAINER_PACKAGE_PATH \
        --module-name $MAIN_TRAINER_MODULE \
        --job-dir $JOB_DIR \
        --region $REGION \
        --config config.yaml \
        -- \
        --user_first_arg=first_arg_value \
        --user_second_arg=second_arg_value
```

```
gcloud ml-engine jobs submit training $JOB_NAME \
        --scale-tier basic \
        --package-path $TRAINER_PACKAGE_PATH \
        --module-name $MAIN_TRAINER_MODULE \
        --job-dir $JOB_DIR \
        --region $REGION \
        -- \
        --user_first_arg=first_arg_value \
        --user_second_arg=second_arg_value
```

```
TRAINER_PACKAGE_PATH="/path/to/your/application/sources"
now=$(date +"%Y%m%d_%H%M%S")
JOB_NAME="your_name_$now"
MAIN_TRAINER_MODULE="trainer.task"
JOB_DIR="gs://your/chosen/job/output/path"
PACKAGE_STAGING_PATH="gs://your/chosen/staging/path"
REGION="us-east1"
RUNTIME_VERSION="1.10"
```

submitting a job (2)

with only config.yaml configuration file,
in which every parameters are defined

```
gcloud ml-engine jobs submit training $JOB_NAME \
        --config config.yaml \
        -- \
        --user_first_arg=first_arg_value \
        --user_second_arg=second_arg_value
```

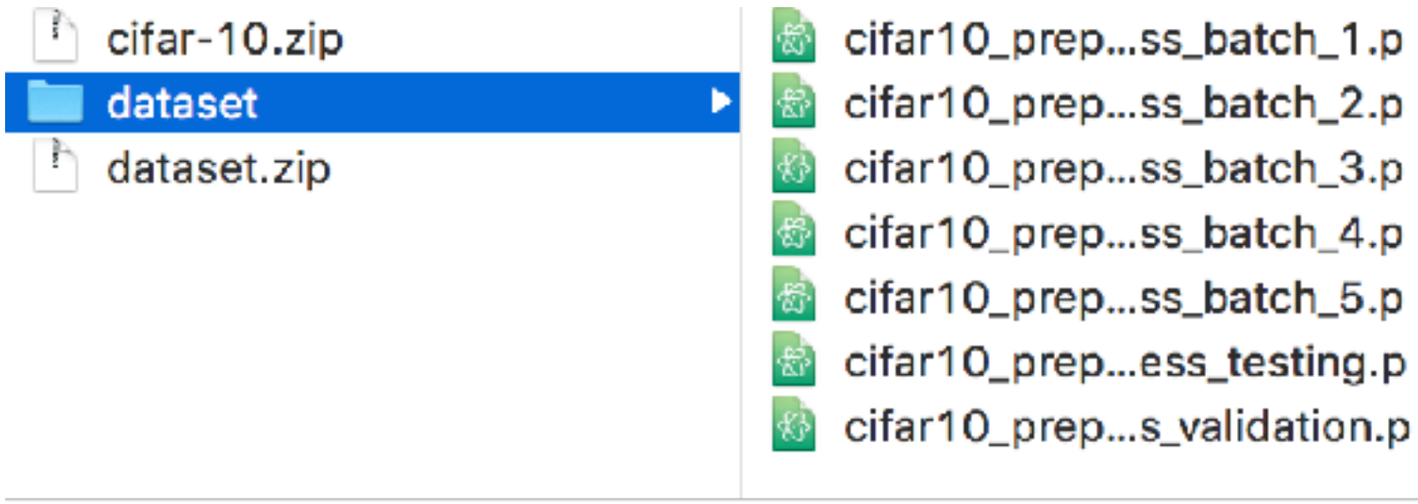
Example

- showing a simple example of CIFAR10 image classification with GoogLeNet (InceptionV1) model.

- This example will use TESLA V100/P100

- All the materials needed for this example can be downloaded from HERE

Step1 - extract materials



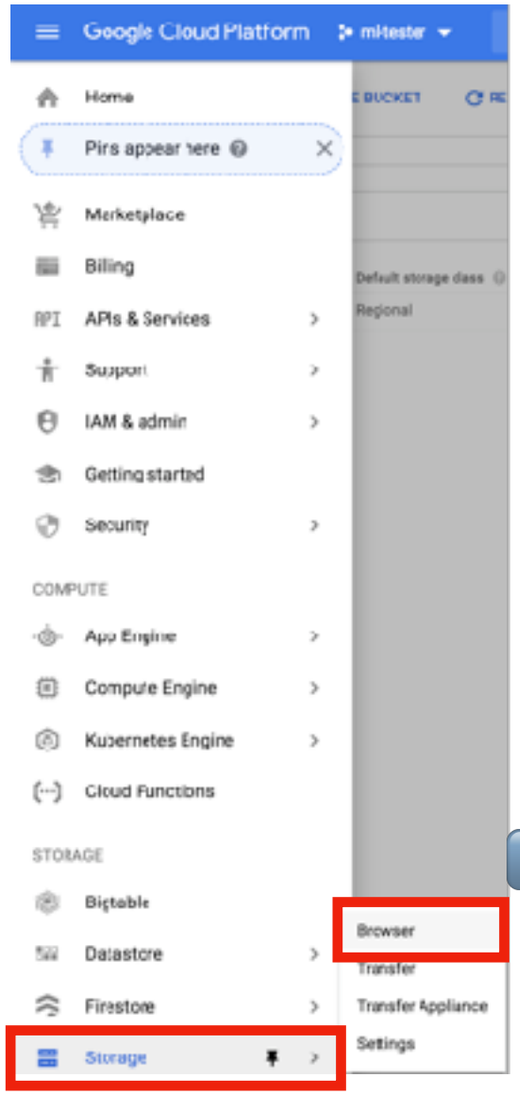
cifar-10 file contains source codes to train

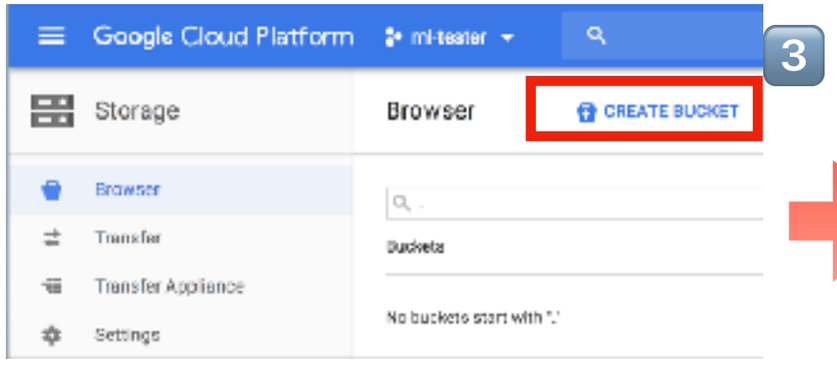
dataset folder contains dataset to train on. these files should be uploaded into the GCP bucket

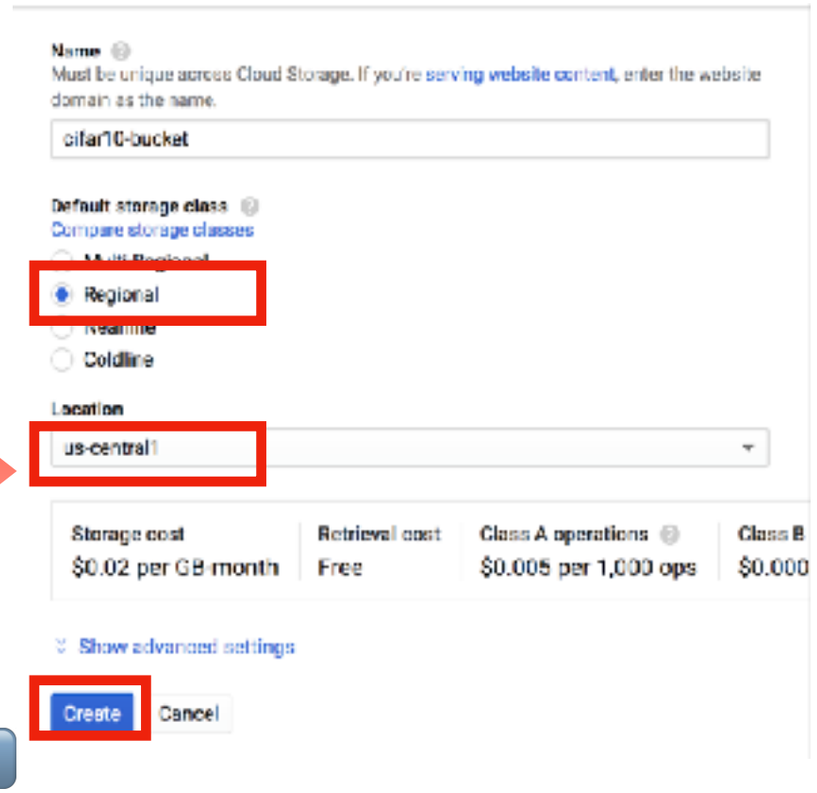
Step2 - upload data to GCP

Also possible to download dataset from source code and save it to the bucket

here, I will show how to upload manually though







locations supporting GPUs/TPU

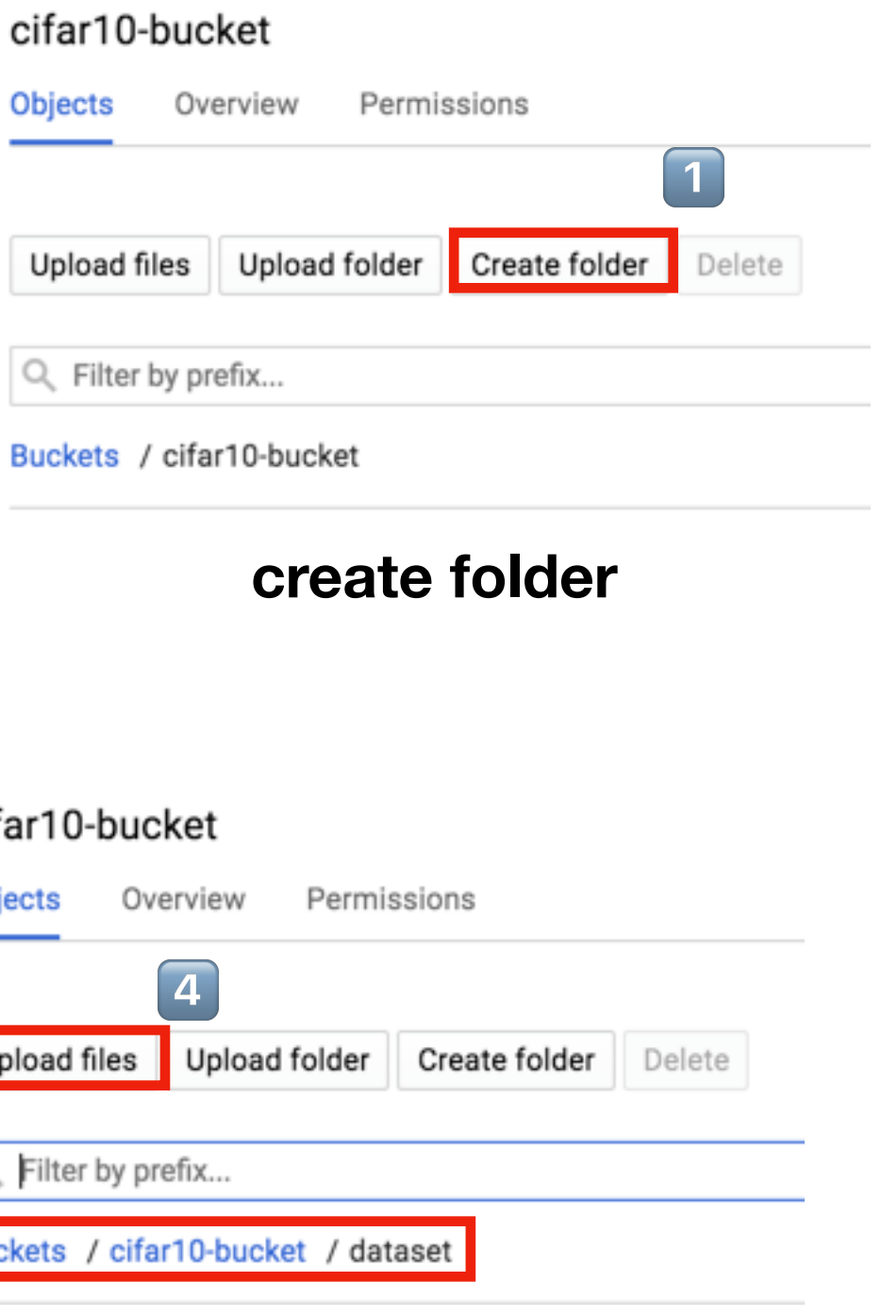
US EUROPE ASIA PACIFIC

Region	us-west1	us-central1	us-east1
NVIDIA TESLA K80		✓	✓
NVIDIA TESLA P100	✓	✓	✓
NVIDIA TESLA V100	✓	✓	
TPU_V2		✓	

US EUROPE ASIA PACIFIC

Region	asia-south1	asia-southeast1	asia-east1
NVIDIA TESLA K80			✓
NVIDIA TESLA P100			✓
NVIDIA TESLA V100			✓
TPU_V2			

Step2 - upload data to GCP(2)



create folder

name folder

check if folder has been created successfully, and click the folder

upload dataset files into the folder

Step2 - upload data to GCP(3)

cifar10-bucket

Objects Overview Permissions

Upload files | Upload folder | Create folder | Delete

Filter by prefix...

Buckets / cifar10-bucket / dataset

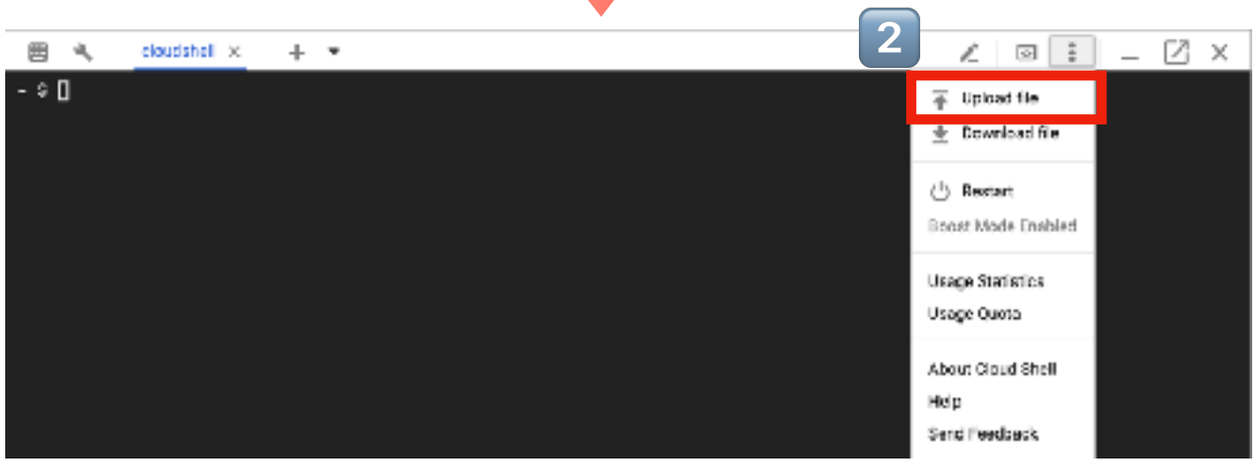
Name	Size	Type	Storage class	Last modified	Public access	Encryption
cifar10_preprocess_batch_1.p	38.7 MB	application/octet-stream	Regional	9/10/18, 5:36 PM	Not public	Google-managed key
cifar10_preprocess_batch_2.p	38.71 MB	application/octet-stream	Regional	9/10/18, 5:36 PM	Not public	Google-managed key
cifar10_preprocess_batch_3.p	38.56 MB	application/octet-stream	Regional	9/10/18, 5:36 PM	Not public	Google-managed key
cifar10_preprocess_batch_4.p	38.68 MB	application/octet-stream	Regional	9/10/18, 5:36 PM	Not public	Google-managed key
cifar10_preprocess_batch_5.p	38.69 MB	application/octet-stream	Regional	9/10/18, 5:35 PM	Not public	Google-managed key
cifar10_preprocess_testing.p	43.12 MB	application/octet-stream	Regional	9/10/18, 5:35 PM	Not public	Google-managed key
cifar10_preprocess_validation.p	21.61 MB	application/octet-stream	Regional	9/10/18, 5:34 PM	Not public	Google-managed key

create “ckpt” folder just like “dataset” folder

- “ckpt” folder will store training checkpoint files

Step3 - upload source files

activate cloud shell



upload cifar-10.zip

check if the file exists

* uploading files will store files only under the home directory

step4-extract source files

```
~ $ mkdir ml-test
~ $ mv cifar-10.zip ./ml-test
~ $ cd ml-test/
~/ml-test $ unzip cifar-10.zip
```

create a folder named “ml-test”

move cifar-10.zip file to the created folder

unzip the cifar-10.zip file after moving into the directory

```
~/ml-test $ find .
.
./cifar-10.zip
./cifar-10
./cifar-10/trainer
./cifar-10/trainer/__init__.py
./cifar-10/trainer/clftrainer.py
./cifar-10/trainer/test.py
./cifar-10/trainer/googlenet.py
./cifar-10/trainer/cifar10_dataset.py
./cifar-10/trainer/imgclfmodel.py
./cifar-10/trainer/dataset.py
./cifar-10/setup.py
./cifar-10/config.yaml
~/ml-test $
```

trainer class (can execute training)

where the main function is

GoogLeNet model definition

CIFAR10 dataset class (knows how to load)

base classes for model and dataset

config/setup files for cloud ml engine

step5-set some variables

```
BUCKET_NAME=cifar10-bucket
REGION=us-central1

OUTPUT_PATH=gs://$BUCKET_NAME/output
JOB_NAME=hello_cifar10_v100
```

```
def get_training_batches_from_preprocessed(self, batch_id, batch_size, scale_to_imagenet=False):
    bucket_name ='cifar10-bucket'
    path = 'dataset'
    filename = 'cifar10_preprocess_batch_' + str(batch_id) + '.p'
    full_path = 'gs://' + bucket_name + '/' + path + '/' + filename

    features, labels = pickle.load(file_io.FileIO(full_path, 'r'))

    if scale_to_imagenet:
        features = self.convert_to_imagenet_size(features)

    return self.get_batches_from(features, labels, batch_size)

def get_valid_set(self, scale_to_imagenet=False):
    bucket_name = 'cifar10-bucket'
    path = 'dataset'
    filename = 'cifar10_preprocess_validation.p'
    full_path = 'gs://' + bucket_name + '/' + path + '/' + filename

    valid_features, valid_labels = pickle.load(file_io.FileIO(full_path, 'r'))

    if scale_to_imagenet:
        valid_features = self.convert_to_imagenet_size(valid_features)

    return valid_features, valid_labels
```

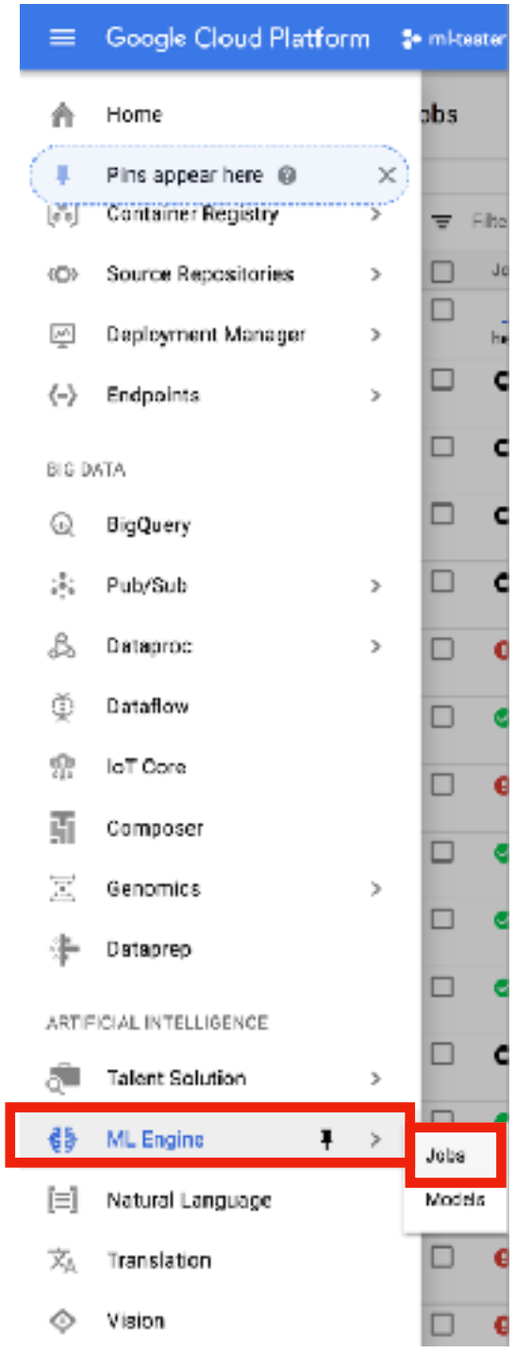
step5-set some variables(2)

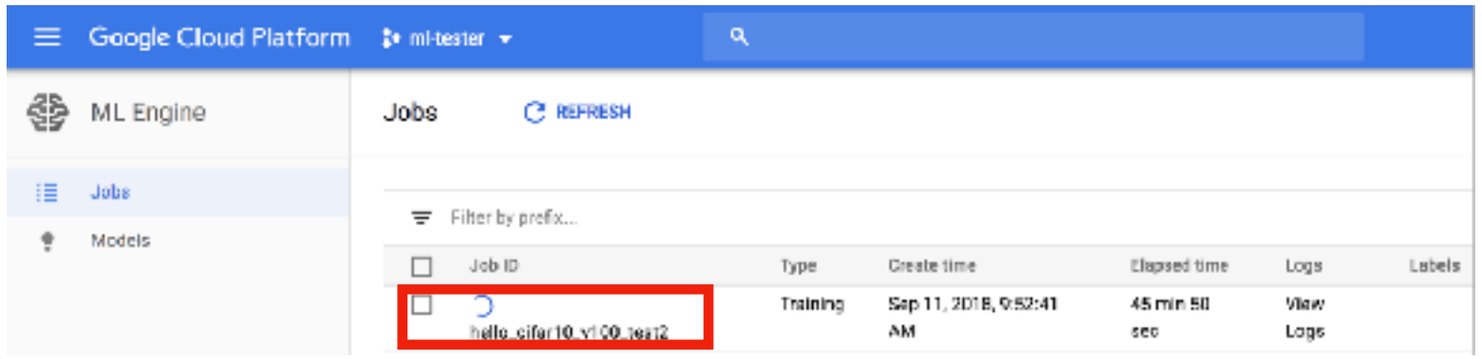
```
def main():

    learning_rate = 0.0001
    epochs = 3
    batch_size = 64

    cifar10_dataset = Cifar10()
    model = GoogLeNet()
    trainer = ClfTrainer(model, cifar10_dataset)

    trainer.run_training(epochs, batch_size, learning_rate,'gs://cifar10-bucket/ckpt/inceptionv1.ckpt')
```

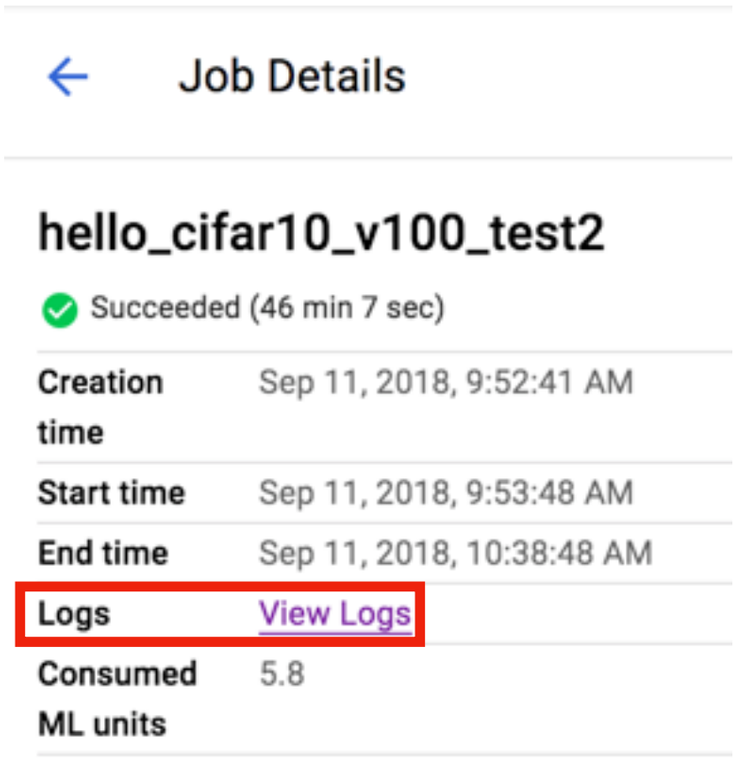
step6-run training

```
ml-test/cifar-10 $ gcloud ml-engine jobs submit training $JOB_NAME \
                    --package-path trainer \
                    --module-name trainer.test \
                    --job-dir $OUTPUT_PATH \
                    --region $REGION \
                    --config config.yaml \
                    --runtime-version 1.9
```

step7-watch/monitor

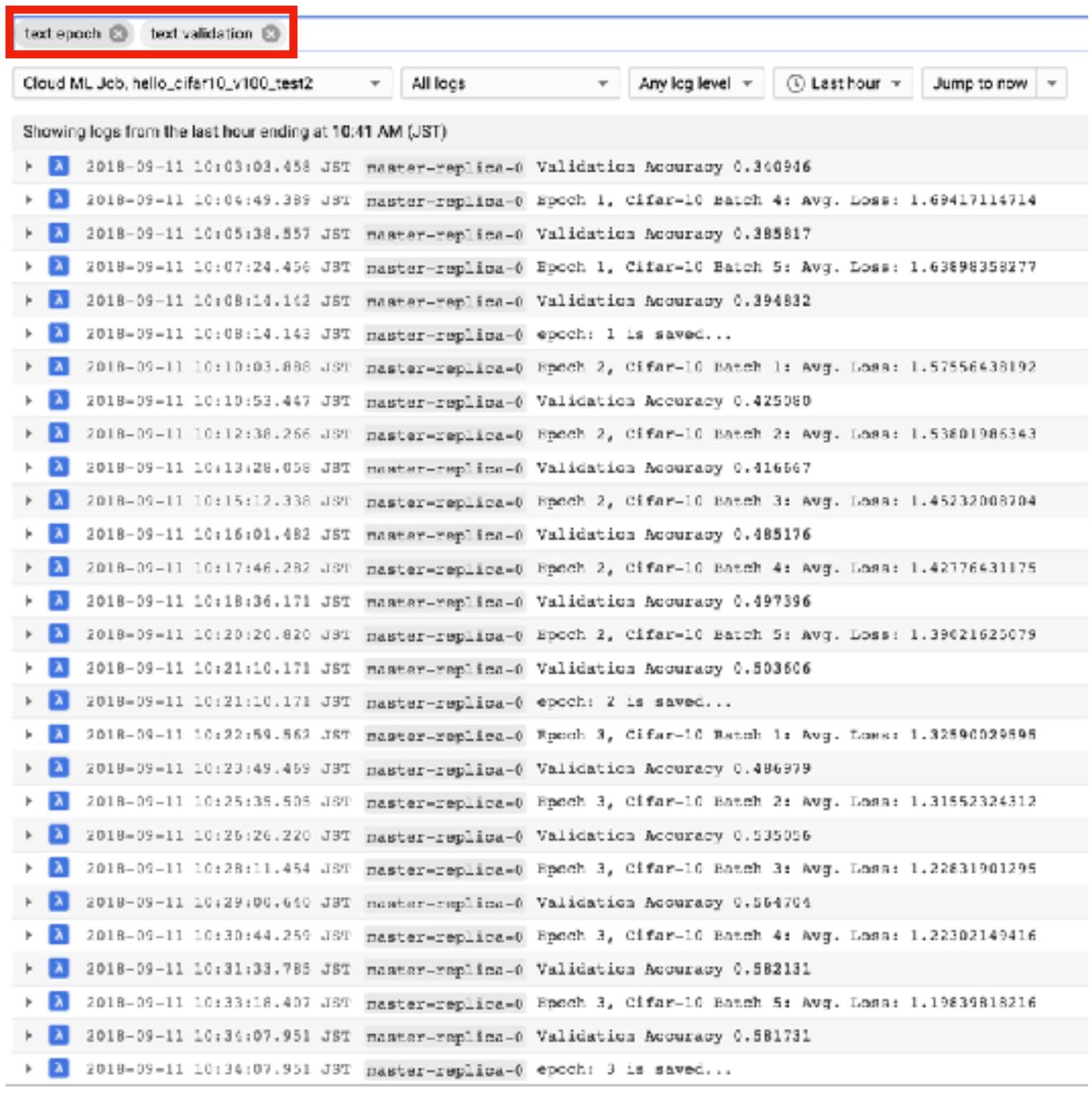






step7-watch/monitor(2)

```
master-replica-0 name: Tesla V100-SXM2-16GB major: 7 minor: 0 memoryClockRate(GHz): 1.53
master-replica-0
master-replica-0 +-----------------------------------------------------------------------------+
master-replica-0 | NVIDIA-SMI 384.111 Driver Version: 384.111 |
master-replica-0 |-------------------------------+----------------------+----------------------+
master-replica-0 | GPU Name Persistence-M| Bus-Id Disp.A | Volatile Uncorr. ECC |
master-replica-0 | Fan Temp Perf Pwr:Usage/Cap| Memory-Usage | GPU-Util Compute M. |
master-replica-0 |===============================+======================+======================|
master-replica-0 | 0 Tesla V100-SXM2... Off | 00000000:00:04.0 Off | 0 |
master-replica-0 | N/A 39C P0 38W / 300W | 15370MiB / 16152MiB | 0% Default |
master-replica-0 +-------------------------------+----------------------+----------------------+
```



appendix

- how to get current project ID?
  $(gcloud config list project --format “value(core.project)”)

- with tf.device(‘/device:GPU:0’):

Google
Cloud Platform

Cloud
ML Engine