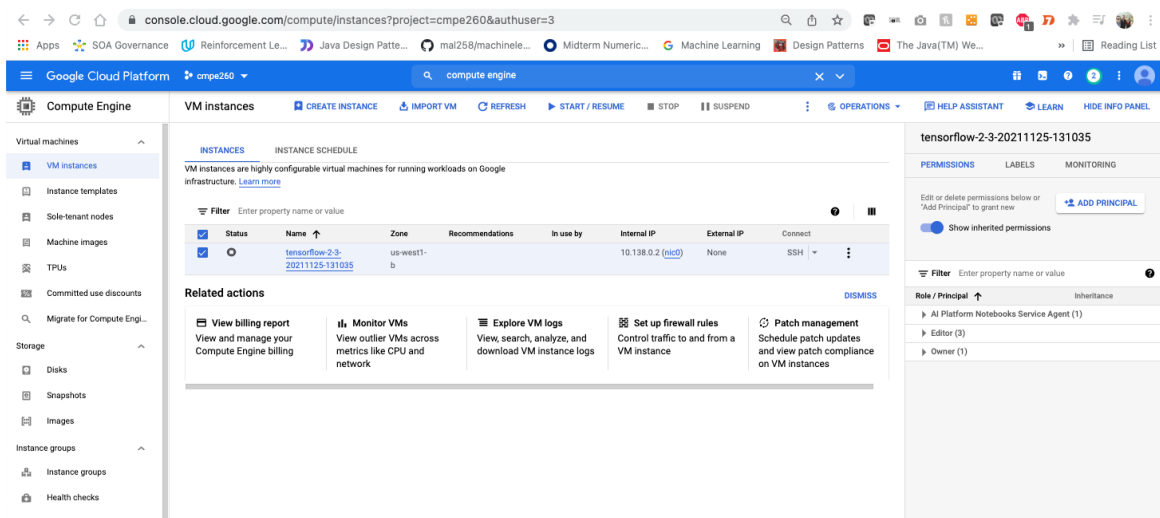


Assignment 7: optional catchup assignment 2 - VERTEX AI - for midterm and quiz - this will catch up midterm.

Vertex AI: Training and serving a custom model

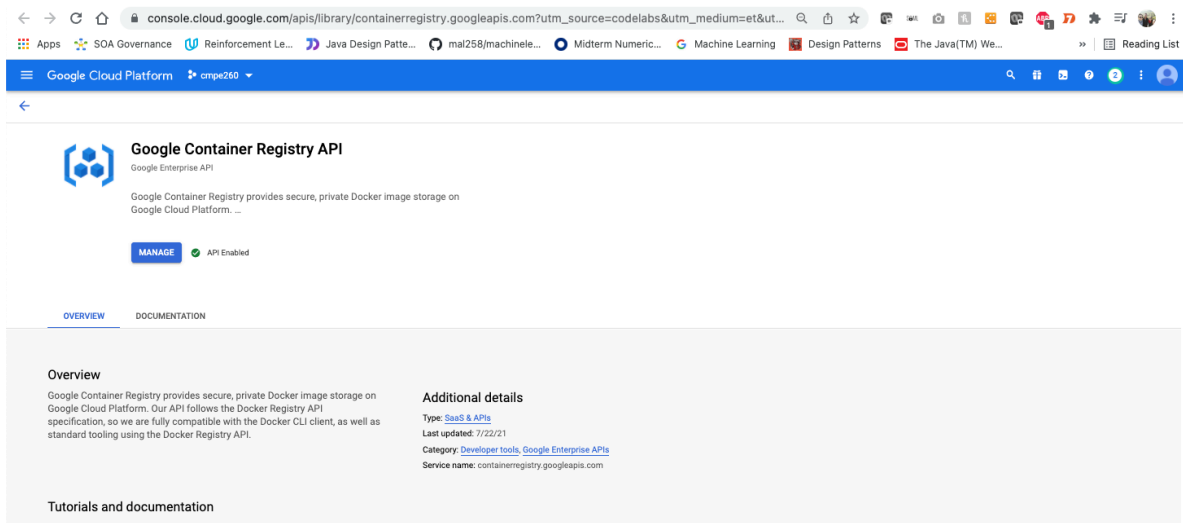
Setting up the environment

1) Enable Compute Engine:



The screenshot shows the Google Cloud Platform console for the project 'cmpe260'. The 'Compute Engine' section is active, displaying a list of VM instances. A single instance is listed with the name 'tensorflow-2-3-20211125-131035', located in the 'us-west1-b' zone. The instance is running the 'tensorflow/tensorflow:2.3.0-gpu' image and has an internal IP address of 10.138.0.2. The page also features a 'Related actions' section with links to view billing reports, monitor VMs, explore VM logs, set up firewall rules, and manage patches.

2) Enable Google Container Registry API



The screenshot shows the Google Cloud Platform console for the project 'cmpe260'. The 'Google Container Registry API' page is displayed, indicating that the API is enabled. The page includes an overview section describing the API as a secure, private Docker image storage solution on Google Cloud Platform. Additional details include the API type (SaaS & APIs), last updated date (7/22/21), category (Developer tools, Google Enterprise APIs), and service name (containregistry.googleapis.com).

3) Enable workbench

Google Cloud Platform console showing the Vertex AI Workbench interface. The 'Notebooks' tab is active, displaying a table of managed notebooks. A notification states that the Notebooks service has been moved under the Vertex AI Workbench service. The table lists a notebook named 'tensorflow-2-3-20211125-131035' with an 'OPEN JUPYTERLAB' button. The right sidebar shows the 'Info panel' with documentation links.

4) Containerize training code and create a Docker file and create a storage bucket

```
(base) jupyter@tensorflow-2-3-20211125-131035:~$ mkdir mpg
(base) jupyter@tensorflow-2-3-20211125-131035:~$ cd mpg
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ touch Dockerfile
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ FROM gcr.io/deeplearning-platform-release/tf2-cpu.2-3
bash: FROM: command not found
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ WORKDIR /root
bash: WORKDIR: command not found
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ WORKDIR /
bash: WORKDIR: command not found
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ # Copies the trainer code to the docker image.
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ COPY trainer /trainer
bash: COPY: command not found
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ # Sets up the entry point to invoke the trainer.
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ ENTRYPOINT ["python", "-m", "trainer.train"]
bash: ENTRYPOINT: command not found
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ ls
Dockerfile
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ vi Dockerfile
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ gcloud config list --format 'value(core.project)'
cmpe260
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ PROJECT_ID=cmpe260
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ BUCKET_NAME=gs://cmpe260-bucket
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ gsutil mb -l us-central1 $BUCKET_NAME
InvalidUrlError: Invalid bucket name in URL "-bucket".
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ gsutil mb -l us-central1 $BUCKET_NAME
InvalidUrlError: Invalid bucket name in URL "-bucket".
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ BUCKET_NAME=gs://cmpe260-bucket
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ gsutil mb -l us-central1 $BUCKET_NAME
Creating gs://cmpe260-bucket/...
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$
```

5) Create training code:

786f87284dee68f6-dot-us-west1.notebooks.googleusercontent.com/lab

Apps SOA Governance Reinforcement Le... Java Design Patte... mal258/machinele... Midterm Numeric... Machine Learning Design Patterns The Java(TM) We... Reading List

File Edit View Run Kernel Git Tabs Settings Help

Filter files by name

Name	Last Modified
/	
mpg	a minute ago
src	4 days ago
tutorials	4 days ago
Untitled.ipynb	4 days ago

Terminal 1

```
def build_model():
    model = keras.Sequential([
        layers.Dense(64, activation='relu', input_shape=[len(train_dataset.keys())]),
        layers.Dense(64, activation='relu'),
        layers.Dense(1)
    ])

    optimizer = tf.keras.optimizers.RMSprop(0.001)

    model.compile(loss='mse',
                  optimizer=optimizer,
                  metrics=['mae', 'mse'])

    return model

model = build_model()

#### Inspect the model

Use the `.summary` method to print a simple description of the model
...

model.summary()

""Now try out the model. Take a batch of `10` examples from the training data and call `model.predict` on it.

It seems to be working, and it produces a result of the expected shape and type.

### Train the model

Train the model for 1000 epochs, and record the training and validation accuracy in the `history` object.

Visualize the model's training progress using the stats stored in the `history` object.

This graph shows little improvement, or even degradation in the validation error after about 100 epochs. Let's update the `model.fit` call to automa
tically stop training when the validation score doesn't improve. We'll use an "EarlyStopping callback" that tests a training condition for every ep
och. If a set amount of epochs elapses without showing improvement, then automatically stop the training.

You can learn more about this callback [here](https://www.tensorflow.org/api_docs/python/tf/keras/callbacks/EarlyStopping).
...

model = build_model()

EPOCHS = 1000

# The patience parameter is the amount of epochs to check for improvement
early_stop = keras.callbacks.EarlyStopping(monitor='val_loss', patience=10)

early_history = model.fit(normed_train_data, train_labels,
                           epochs=EPOCHS, validation_split = 0.2,
                           callbacks=[early_stop])

# Export model and save to GCS
model.save(BUCKET + '/mpg/model')
```

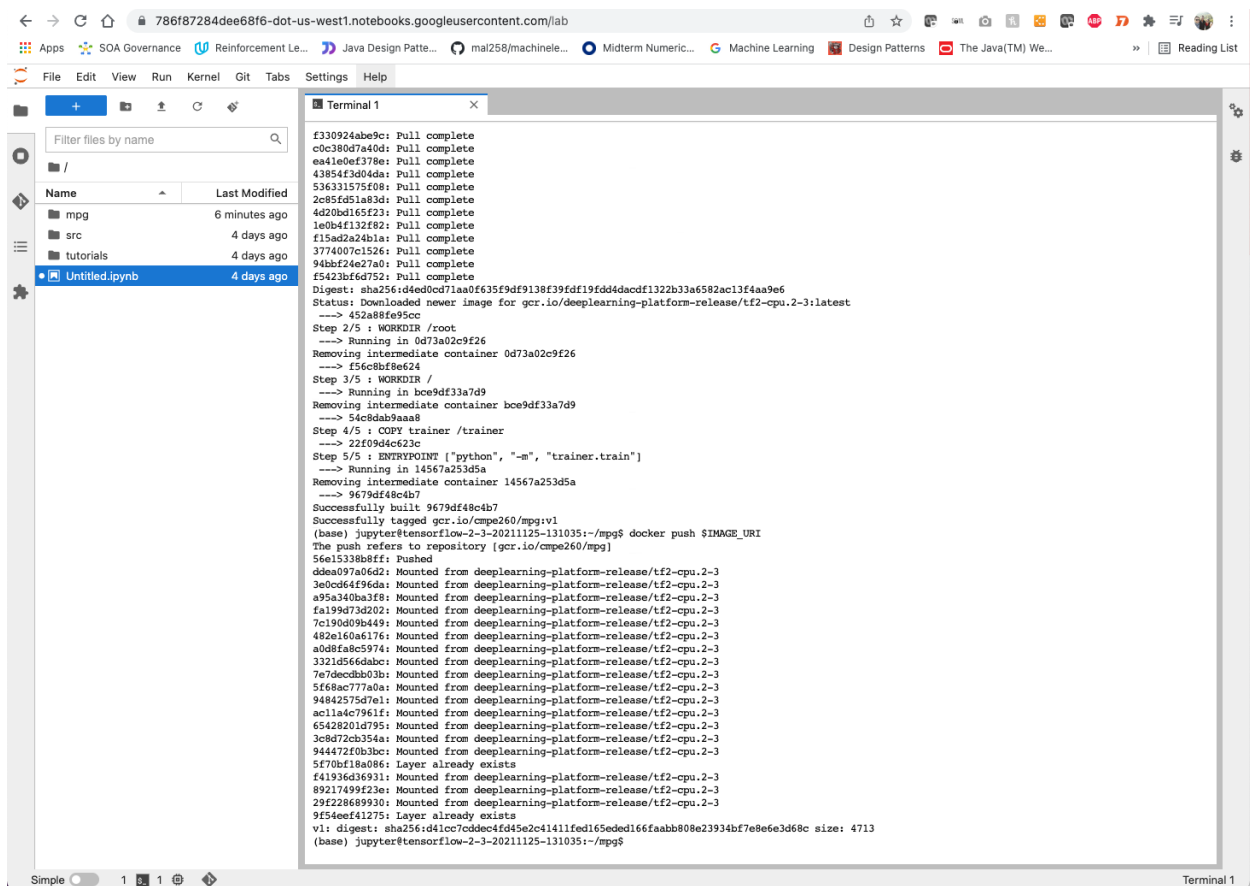
159,33 Bot

Simple 1 1

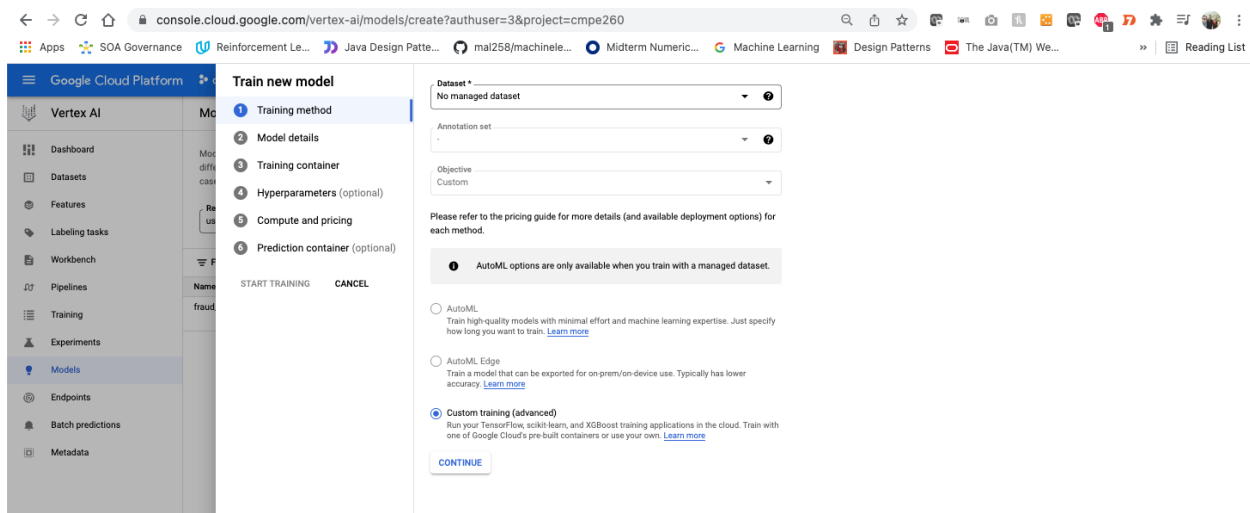
6) Build and test the container locally

The screenshot shows a JupyterLab interface with a file browser on the left and a terminal window on the right. The file browser shows a directory structure with files like 'mpg', 'src', 'tutorials', and 'Untitled.ipynb'. The terminal window shows the following commands and output:

```
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ mkdir trainer
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ touch trainer/train.py
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ ls
Dockerfile  trainer
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ cd trainer/
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg/trainer$ vi train.py
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg/trainer$ docker build . -t $IMAGE_URI
unable to prepare context: unable to evaluate symlinks in Dockerfile path: lstat /home/jupyter/mpg/trainer/Dockerfile: no such file or directory
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg/trainer$ cd ..
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$ docker build . -t $IMAGE_URI
Sending build context to Docker daemon 8.704kB
Step 1/5 : FROM gcr.io/deeplearning-platform-release/tf2-cpu.2-3
latest: Pulling from deeplearning-platform-release/tf2-cpu.2-3
7b1a6ab2e4d4: Pull complete
6d576096c0bf: Pull complete
5a39c8988a9a: Pull complete
3bf0fd278fc1: Pull complete
4f4fb700ef54: Pull complete
a7b7cd42e273: Pull complete
5ed778ec318f: Pull complete
f99f0c7e1e5e: Pull complete
f330924abe9c: Pull complete
c0c380d7a40d: Pull complete
ea41e0ef378e: Pull complete
43854f3d04da: Pull complete
536331575f08: Pull complete
2c85fd51a83d: Pull complete
4d20bd165f23: Pull complete
1e0b4f132f82: Pull complete
f15ad2a24b1a: Pull complete
3774007c1526: Pull complete
94bbf24e27a0: Pull complete
f5423bf6d752: Pull complete
Digest: sha256:d4ed0cd71aa0f635f9df9138f39fd19fdd4dacf1322b33a6582ac13f4aa9e6
Status: Downloaded newer image for gcr.io/deeplearning-platform-release/tf2-cpu.2-3:latest
--> 452a88fe95cc
Step 2/5 : WORKDIR /root
--> Running in 0d73a02c9f26
Removing intermediate container 0d73a02c9f26
--> f56c8bf8e624
Step 3/5 : WORKDIR /
--> Running in bce9df33a7d9
Removing intermediate container bce9df33a7d9
--> 54c8dab9aaa8
Step 4/5 : COPY trainer /trainer
--> 22f09d4c623c
Step 5/5 : ENTRYPOINT ["python", "-m", "trainer.train"]
--> Running in 14567a253d5a
Removing intermediate container 14567a253d5a
--> 9679df48c4b7
Successfully built 9679df48c4b7
Successfully tagged gcr.io/cmp260/mpg:v1
(base) jupyter@tensorflow-2-3-20211125-131035:~/mpg$
```



7) Run a training job in vertex AI



console.cloud.google.com/vertex-ai/models/create?authuser=3&project=cmpe260

Google Cloud Platform

Vertex AI

Dashboard

Datasets

Features

Labeling tasks

Workbench

Pipelines

Training

Experiments

Models

Endpoints

Batch predictions

Metadata

Train new model

Training method

Model details

Training container

Hyperparameters (optional)

Compute and pricing

Prediction container (optional)

START TRAINING

CANCEL

Select a pre-built container or build a custom container using ML frameworks (as well as non-ML dependencies, libraries and binaries) that are not otherwise supported. [Learn more](#)

☐ Pre-built container

View the list of [supported runtimes](#) including TensorFlow and scikit-learn versions

☒ Custom container

Build a custom Docker container. Must be stored in [Container Registry](#)

Custom container settings

Container image *

BROWSE

Container image URL is required.

gs:// Model output directory

BROWSE

Your model artifacts and other data needed for training will be stored on Cloud Storage. You should specify a path here if you do not set an output directory in your application code or arguments.

Arguments

Optional. Add arguments for the command that runs when the container starts. Overrides the container's CMD instruction. Enter one parameter and its argument per line.

-flag_XXXXX
-flag2
flag3

Select container image

CONTAINER REGISTRY

ARTIFACT REGISTRY

Project: cmpe260 [CHANGE](#)

gcr.io/cmpe260/mpg

d41cc7cddc v1 3 minutes ago

SELECT

CANCEL

console.cloud.google.com/vertex-ai/models/create?authuser=3&project=cmpe260

Google Cloud Platform

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Hyperparameters (optional)

Compute and pricing

Prediction container (optional)

START TRAINING

CANCEL

You can associate your custom-trained model with a container in order to serve prediction requests using Vertex AI. [Learn more about getting predictions.](#)

☐ No prediction container

You can always import your model artifact later to serve prediction requests

☒ Pre-built container

View the list of [supported runtimes](#) including TensorFlow and PyTorch versions

☐ Custom container

Build a custom Docker container. Must be stored in [Container Registry](#) or [Artifact Registry](#)

Pre-built container settings

Vertex AI provides Docker container images for serving predictions. To use a pre-built container, your trained model code must be in Python 3.7. [Learn more about pre-built containers](#)

In order to run in a pre-built container, your code needs to be in Python 3.7

Model framework *

TensorFlow

Model framework version *

2.1

Accelerator type *

None

Model directory *

gs:// cmpe260-bucket/mpg BROWSE

Cloud Storage location containing the model artifact and any supporting files

Predict schemata

Optional. [Learn more about the predict schemata](#)

gs:// Instances BROWSE

Cloud Storage location to a YAML file that defines the format of a single instance used in prediction and explanation requests.

gs:// Parameters BROWSE

Cloud Storage location to a YAML file that defines the prediction and explanation parameters.

gs:// Predictions BROWSE

Cloud Storage location to a YAML file that defines the format of a single prediction or explanation.

Vertex AI Training Pipelines

Training pipelines are the primary model training workflow in Vertex AI. You can use training pipelines to create an AutoML-trained model or a custom-trained model. For custom-trained models, training pipelines orchestrate custom training jobs and hyperparameter tuning with additional steps like adding a dataset or uploading the model to Vertex AI for prediction serving. [Learn More](#)

Region: us-central1 (Iowa)

Name	ID	Status	Job type	Model type	Created	Elapsed time
fraud_detection_2021112512351	850544811852169216	Finished	Training pipeline	Tabular classification	Nov 25, 2021, 1:26:32 PM	1 hr 56 min
autorm-beans1637862305	1811500382342348800	Finished	Training pipeline	Tabular classification	Nov 25, 2021, 9:49:58 AM	2 hr 22 min

Deploy a model endpoint and get prediction on deployed model

```

from google.cloud import aiplatform

endpoint = aiplatform.Endpoint(
    endpoint_name="projects/YOUR-PROJECT-NUMBER/locations/us-central1/endpoints/YOUR-ENDPOINT-ID"
)

[ ]: test_mpg = [1.483887183355929,
1.8659883497083019,
2.234626276849616,
1.8187816540094983,
-2.538980710602246,
-1.6846416850441676,
-0.4651483719733302,
-0.4952254887173721,
0.7746763768735953]

response = endpoint.predict([test_mpg])
print('API response: ', response)
print('Predicted MPG: ', response.predictions[0][0])

```