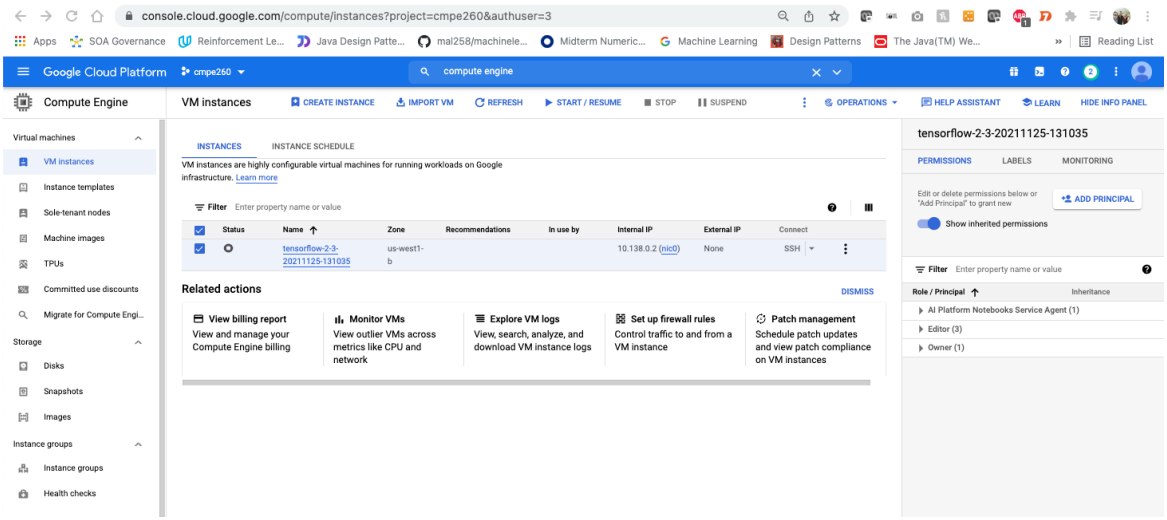
Assignment 7: optional catchup assignment 2 - VERTEX AI - for midterm and quiz - this will catch up midterm.

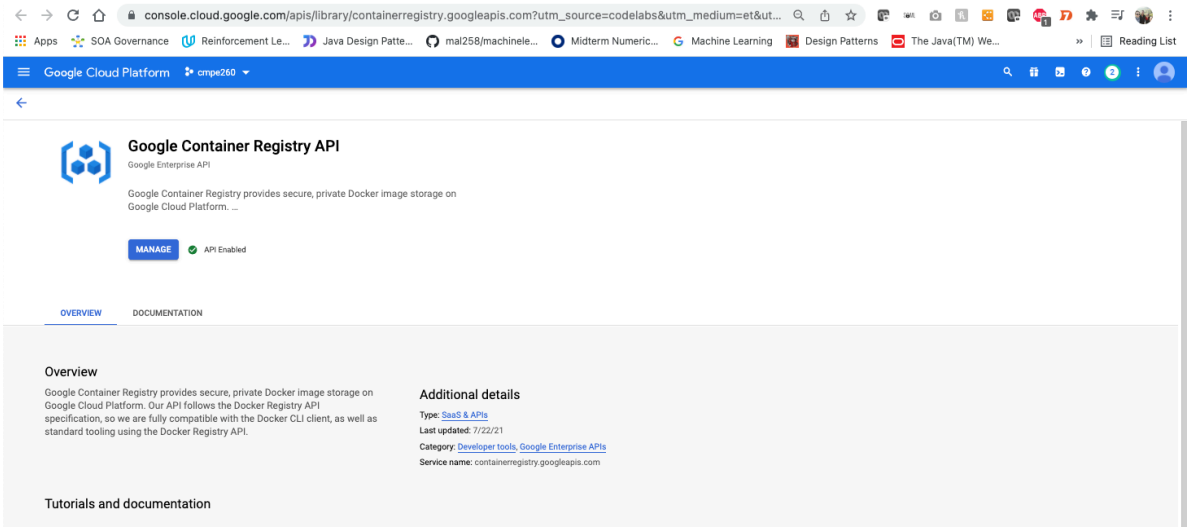
Vertex AI: Training and serving a custom model

Setting up the environment

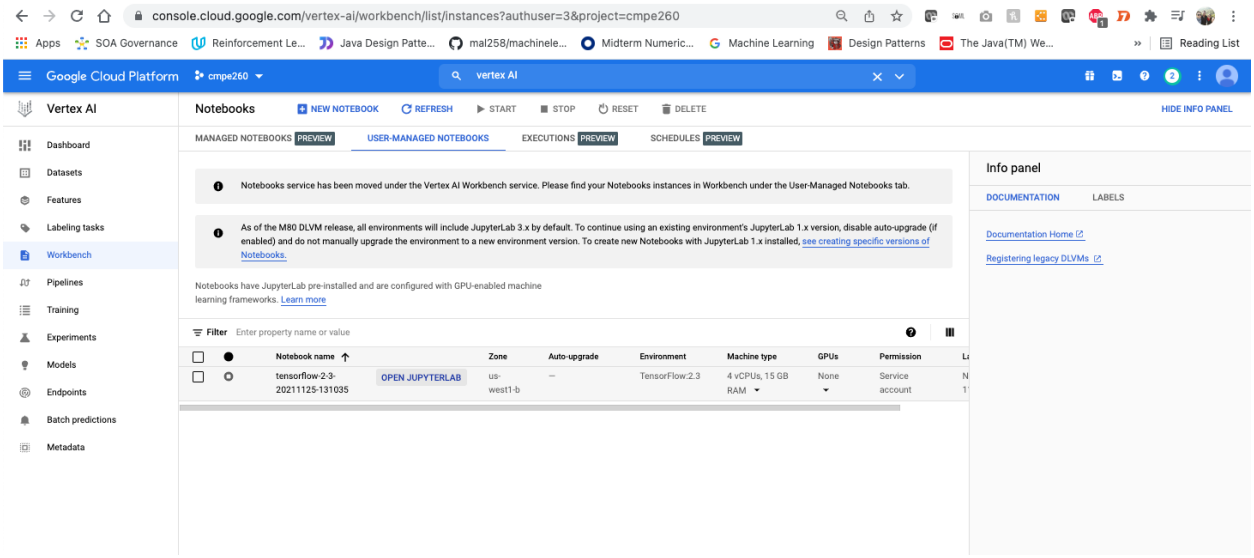
1) Enable Compute Engine:



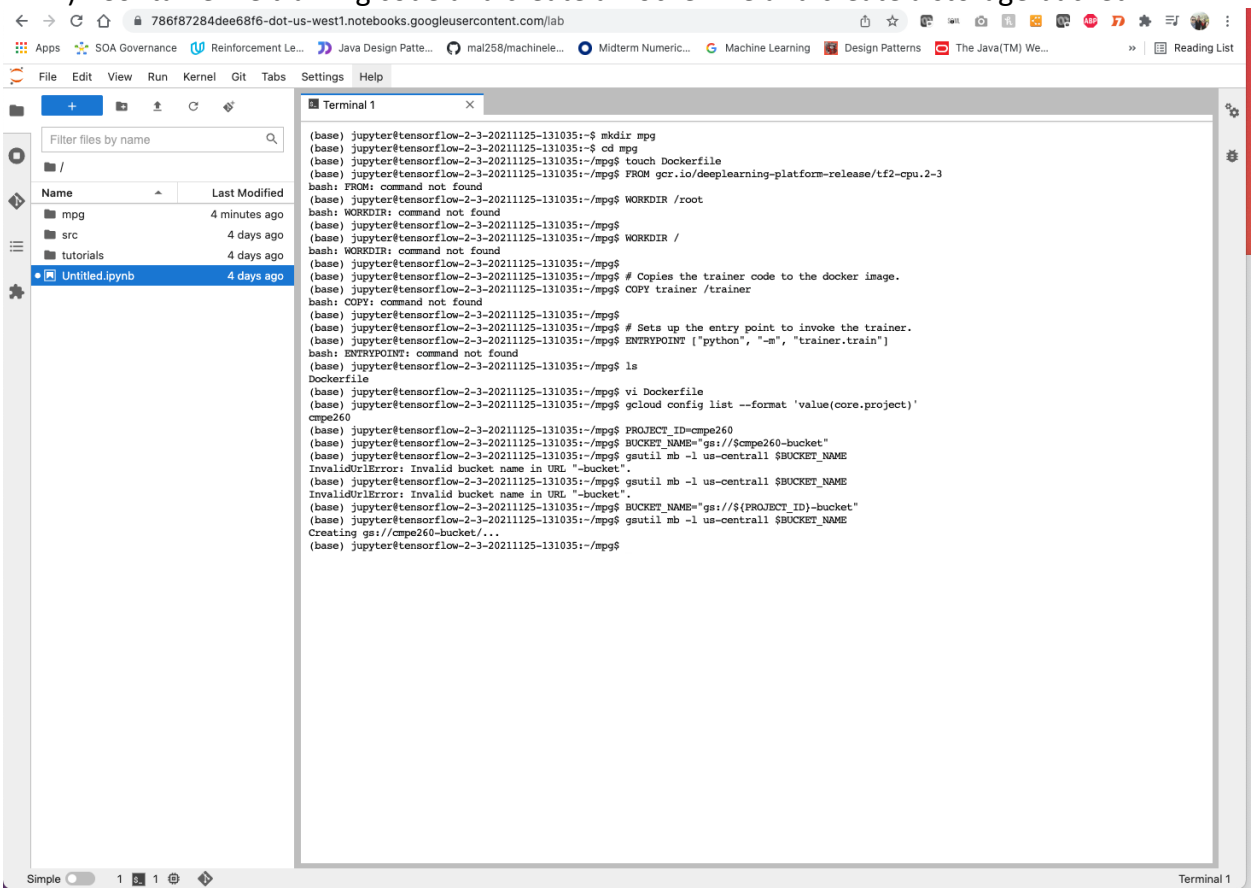
2) Enable Google Container Registry API



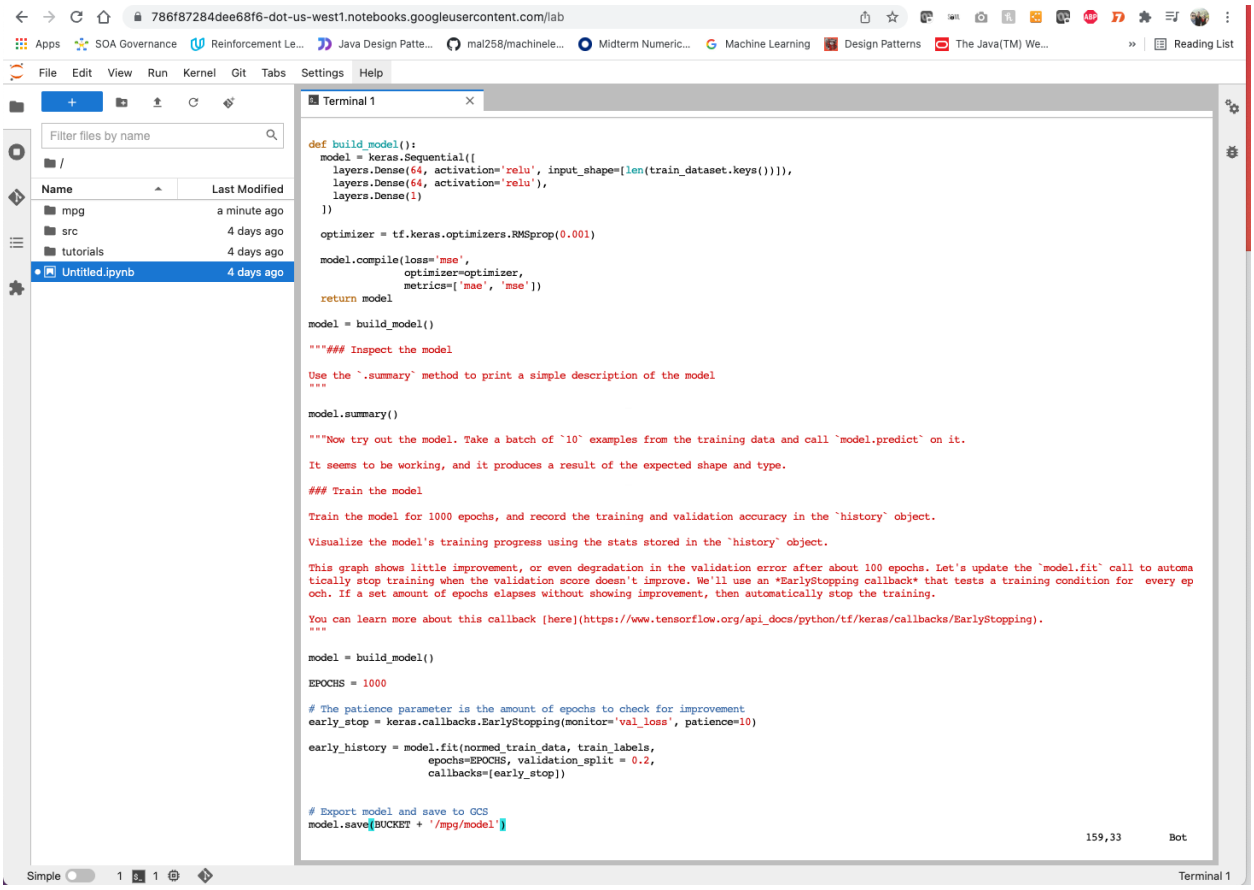
3) Enable workbench



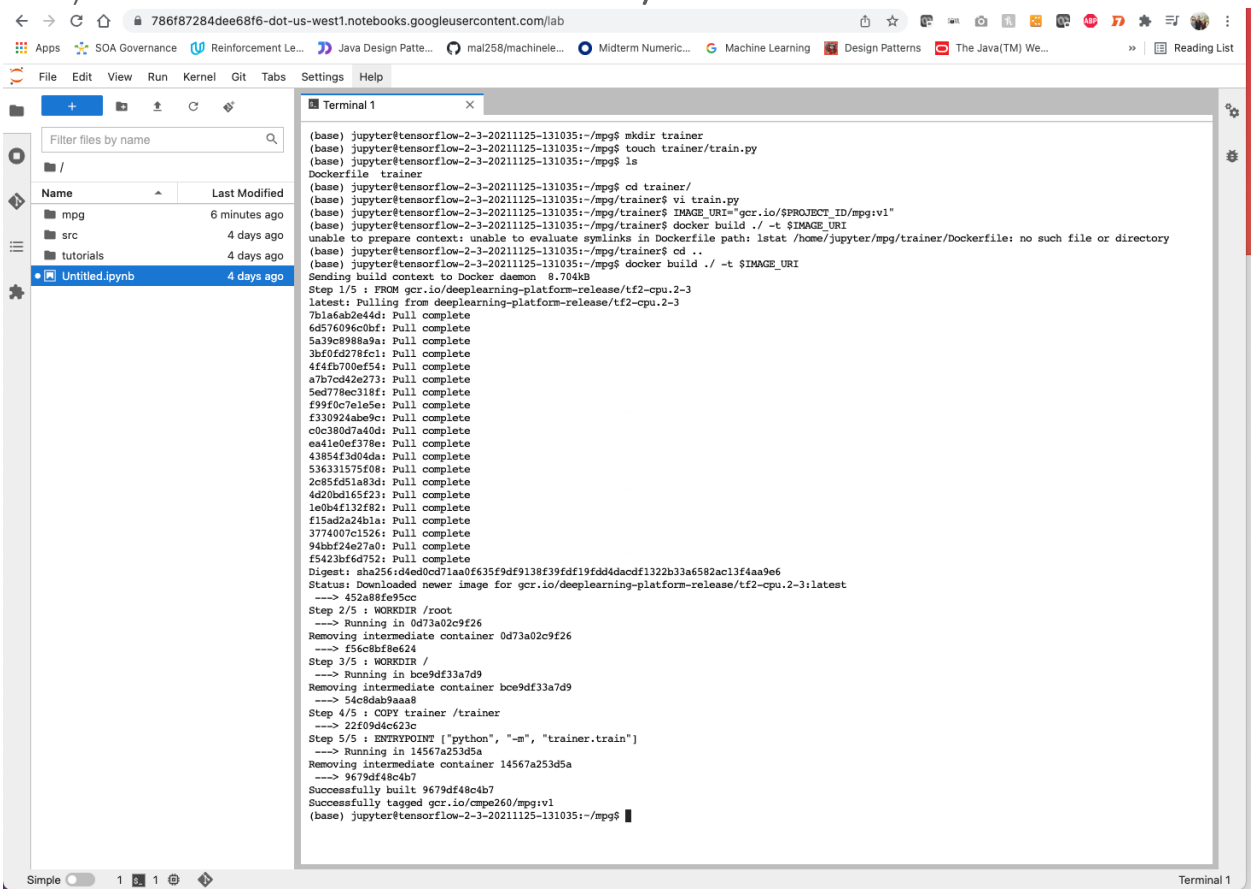
4) Containerize training code and create a Docker file and create a storage bucket

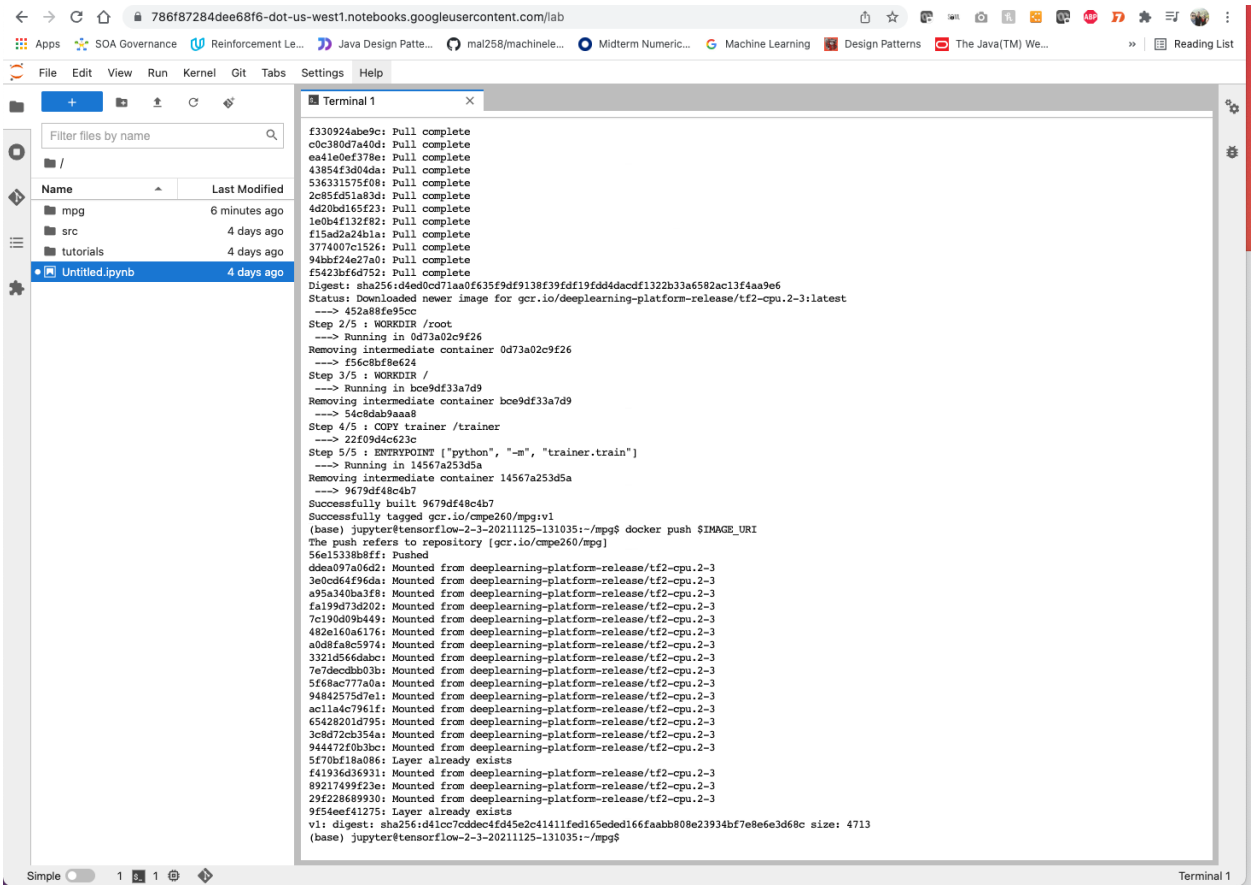


5) Create training code:

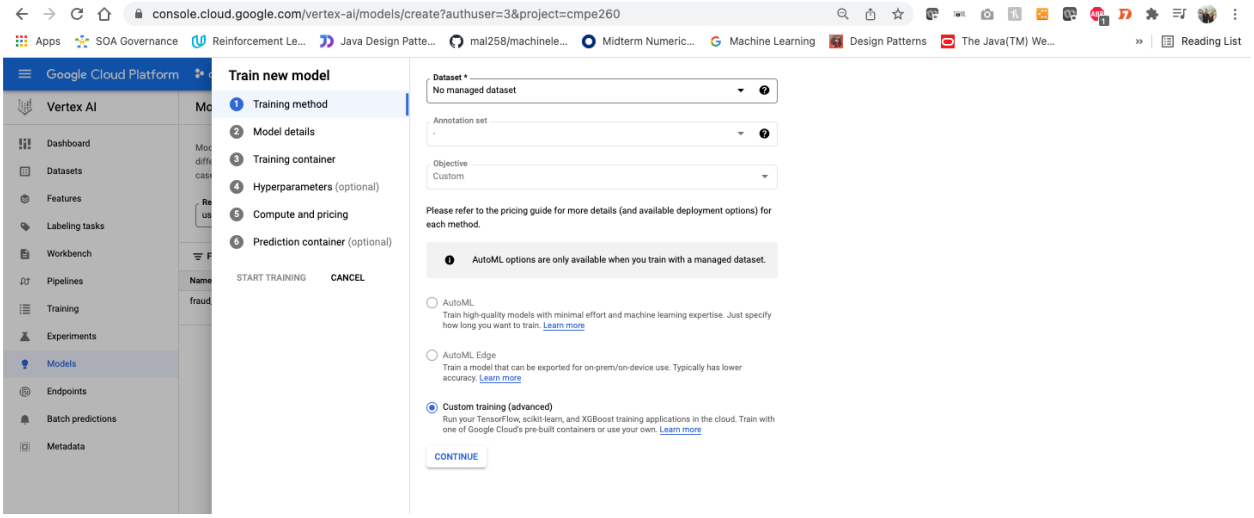

786f87284dee68f6-dot-us-west1.notebooks.googleusercontent.com/lab
Apps
SOA Governance
Reinforcement Le...
Java Design Patte...
mal258/machinele...
Midterm Numeric...
Machine Learning
Design Patterns
The Java(TM) We...
Reading List
File Edit View Run Kernel Git Tabs Settings Help
Filter files by name
/
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
Terminal 1

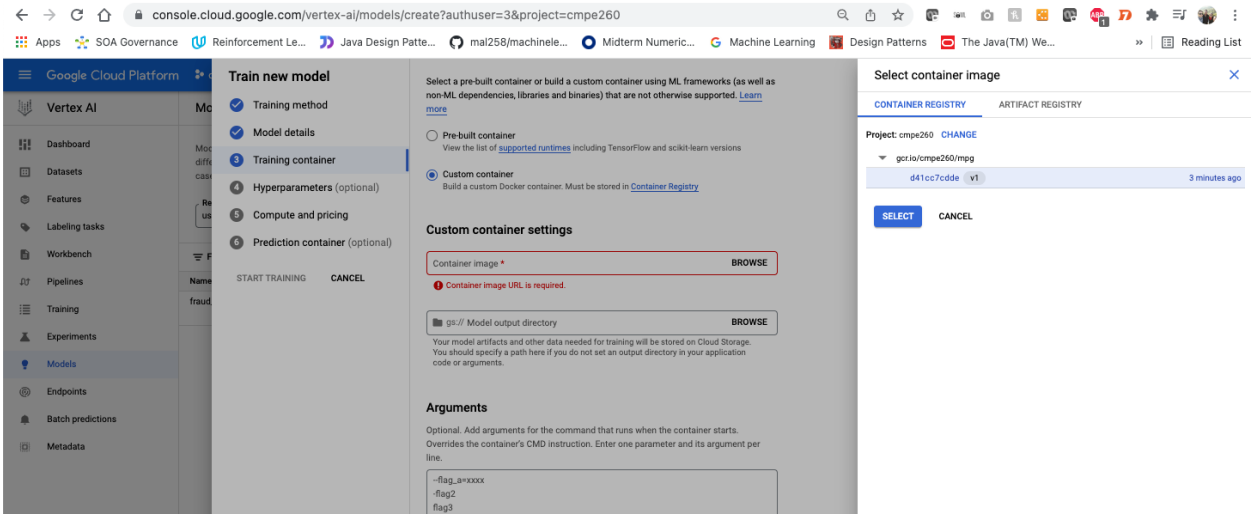
6) Build and test the container locally

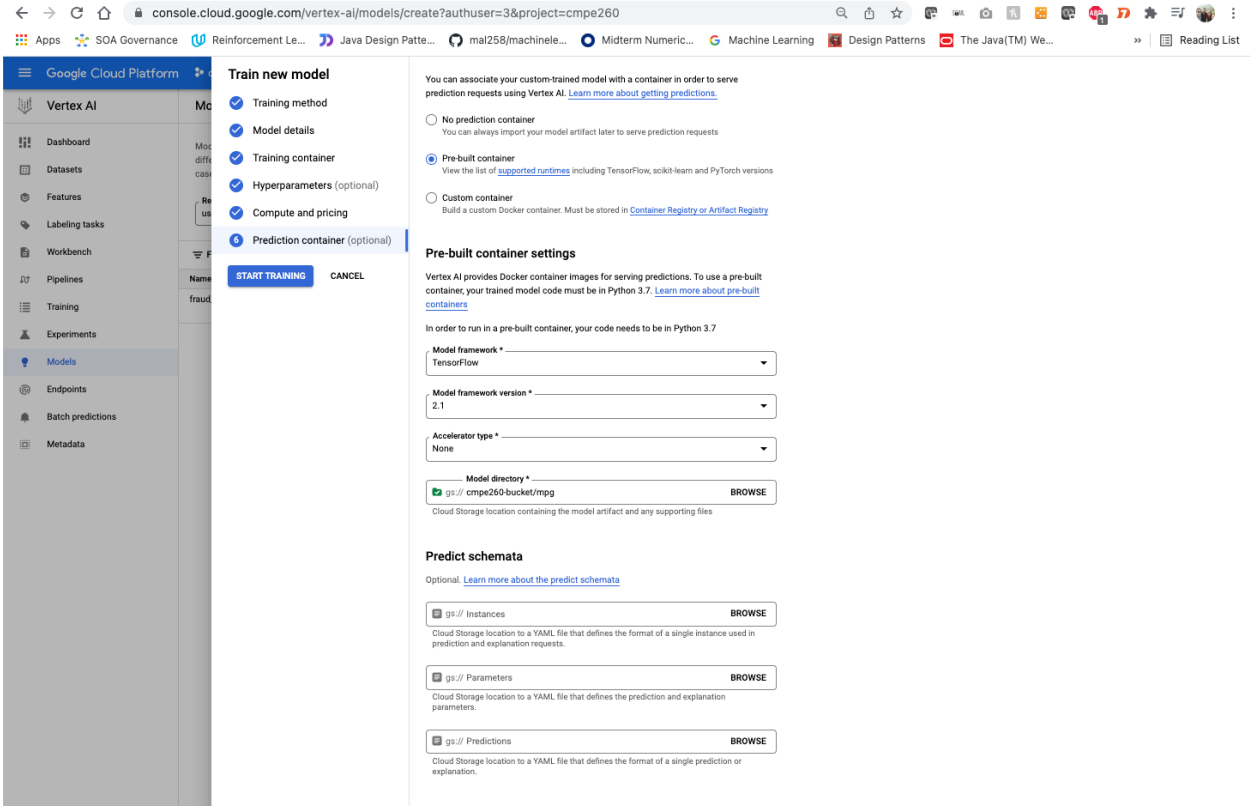


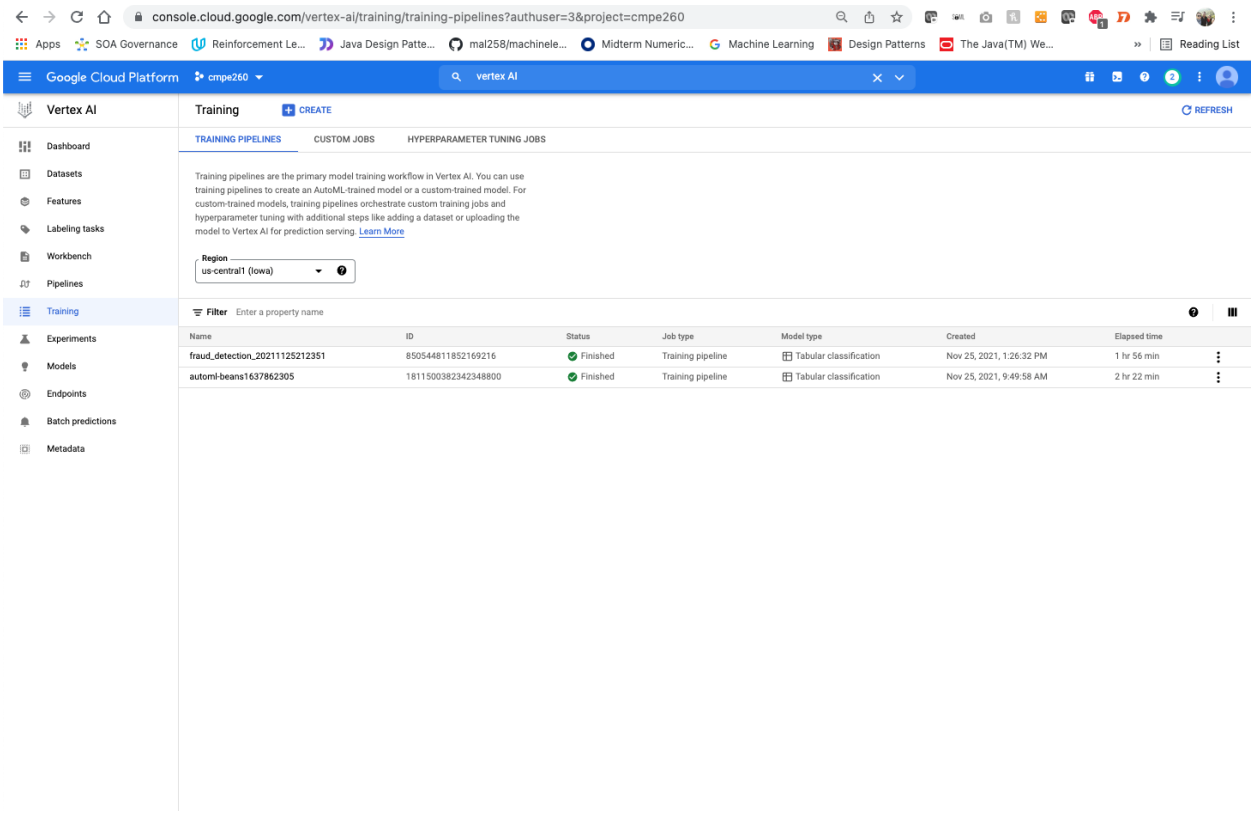


7) Run a training job in vertex AI




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_a=xxxx
-flag2
flag3
Select container image
CONTAINER REGISTRY
ARTIFACT REGISTRY
Project: cmpe260 CHANGE
gcr.io/cmpe260/mpg
d41cc7cdde v1
3 minutes ago
SELECT
CANCEL


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
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, scikit-learn 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.



Deploy a model endpoint and get prediction on deployed model