# Structured data prediction using Cloud Al Platform

This notebook illustrates:

- 1. Create a BigQuery Dataset and Google Cloud Storage Bucket
- 2. Export from BigQuery to CSVs in GCS
- 3. Training on Cloud AI Platform
- 4. Deploy trained model

```
In [1]:
         !sudo chown -R jupyter:jupyter /home/jupyter/training-data-analyst
In [2]:
         !pip install --user google-cloud-bigguery==1.25.0
        Collecting google-cloud-bigguery==1.25.0
          Downloading google_cloud_bigquery-1.25.0-py2.py3-none-any.whl (169 kB)
                                       169 kB 7.4 MB/s eta 0:00:01
        Requirement already satisfied: google-auth<2.0dev,>=1.9.0 in /opt/conda/lib/pyth
        on3.7/site-packages (from google-cloud-bigguery==1.25.0) (1.35.0)
        Requirement already satisfied: six<2.0.0dev,>=1.13.0 in /opt/conda/lib/python3.
        7/site-packages (from google-cloud-bigquery==1.25.0) (1.16.0)
        Collecting google-resumable-media<0.6dev,>=0.5.0
          Downloading google resumable media-0.5.1-py2.py3-none-any.whl (38 kB)
        Collecting google-cloud-core<2.0dev,>=1.1.0
          Downloading google cloud core-1.7.2-py2.py3-none-any.whl (28 kB)
        Requirement already satisfied: google-api-core<2.0dev,>=1.15.0 in /opt/conda/li
        b/python3.7/site-packages (from google-cloud-bigquery==1.25.0) (1.31.2)
        Requirement already satisfied: protobuf>=3.6.0 in /opt/conda/lib/python3.7/site-
        packages (from google-cloud-bigguery==1.25.0) (3.16.0)
        Requirement already satisfied: packaging>=14.3 in /opt/conda/lib/python3.7/site-
        packages (from google-api-core<2.0dev,>=1.15.0->google-cloud-bigguery==1.25.0)
        Requirement already satisfied: setuptools>=40.3.0 in /opt/conda/lib/python3.7/si
        te-packages (from google-api-core<2.0dev,>=1.15.0->google-cloud-bigquery==1.25.
        0) (57.4.0)
        Requirement already satisfied: googleapis-common-protos<2.0dev,>=1.6.0 in /opt/c
        onda/lib/python3.7/site-packages (from google-api-core<2.0dev,>=1.15.0->google-c
        loud-bigquery==1.25.0) (1.53.0)
        Requirement already satisfied: pytz in /opt/conda/lib/python3.7/site-packages (f
        rom google-api-core<2.0dev,>=1.15.0->google-cloud-bigguery==1.25.0) (2021.1)
        Requirement already satisfied: requests<3.0.0dev,>=2.18.0 in /opt/conda/lib/pyth
        on3.7/site-packages (from google-api-core<2.0dev,>=1.15.0->google-cloud-bigquery
        ==1.25.0) (2.25.1)
        Requirement already satisfied: rsa<5,>=3.1.4 in /opt/conda/lib/python3.7/site-pa
        ckages (from google-auth<2.0dev,>=1.9.0->google-cloud-bigquery==1.25.0) (4.7.2)
        Requirement already satisfied: cachetools<5.0,>=2.0.0 in /opt/conda/lib/python3.
        7/site-packages (from google-auth<2.0dev,>=1.9.0->google-cloud-bigguery==1.25.0)
        Requirement already satisfied: pyasn1-modules>=0.2.1 in /opt/conda/lib/python3.
        7/site-packages (from google-auth<2.0dev,>=1.9.0->google-cloud-bigguery==1.25.0)
        (0.2.7)
        Requirement already satisfied: pyparsing>=2.0.2 in /opt/conda/lib/python3.7/site
        -packages (from packaging>=14.3->google-api-core<2.0dev,>=1.15.0->google-cloud-b
```

```
igquery==1.25.0) (2.4.7)
Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /opt/conda/lib/python3.7/
site-packages (from pyasn1-modules>=0.2.1->google-auth<2.0dev,>=1.9.0->google-cl
oud-bigguery==1.25.0) (0.4.8)
Requirement already satisfied: idna<3,>=2.5 in /opt/conda/lib/python3.7/site-pac
kages (from requests<3.0.0dev,>=2.18.0->google-api-core<2.0dev,>=1.15.0->google-
cloud-bigguery==1.25.0) (2.10)
Requirement already satisfied: certifi>=2017.4.17 in /opt/conda/lib/python3.7/si
te-packages (from requests<3.0.0dev,>=2.18.0->google-api-core<2.0dev,>=1.15.0->g
oogle-cloud-bigquery==1.25.0) (2021.5.30)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in /opt/conda/lib/python3.
7/site-packages (from requests<3.0.0dev,>=2.18.0->google-api-core<2.0dev,>=1.15.
0->google-cloud-bigguery==1.25.0) (1.26.6)
Requirement already satisfied: chardet<5,>=3.0.2 in /opt/conda/lib/python3.7/sit
e-packages (from requests<3.0.0dev,>=2.18.0->google-api-core<2.0dev,>=1.15.0->go
ogle-cloud-bigquery==1.25.0) (4.0.0)
Installing collected packages: google-resumable-media, google-cloud-core, google
-cloud-bigguery
ERROR: pip's dependency resolver does not currently take into account all the pa
ckages that are installed. This behaviour is the source of the following depende
ncy conflicts.
google-cloud-storage 1.42.0 requires google-resumable-media<3.0dev,>=1.3.0; pyth
on version >= "3.6", but you have google-resumable-media 0.5.1 which is incompat
Successfully installed google-cloud-bigquery-1.25.0 google-cloud-core-1.7.2 goog
le-resumable-media-0.5.1
```

**Note**: Restart your kernel to use updated packages.

Kindly ignore the deprecation warnings and incompatibility errors related to google-cloudstorage.

## Set up environment variables and load necessary libraries

Set environment variables so that we can use them throughout the entire notebook. We will be using our project name for our bucket, so you only need to change your project and region.

```
In [3]:
         # change these to try this notebook out
         BUCKET = 'qwiklabs-gcp-01-44a3211a105a' # Replace with the your bucket name
         PROJECT = 'qwiklabs-gcp-01-44a3211a105a' # Replace with your project-id
         REGION = 'us-central1'
In [4]:
         import os
         from google.cloud import bigquery
In [5]:
         os.environ["PROJECT"] = PROJECT
         os.environ["BUCKET"] = BUCKET
         os.environ["REGION"] = REGION
         os.environ["TFVERSION"] = "2.3"
         os.environ["PYTHONVERSION"] = "3.7"
In [6]:
         %%bash
         export PROJECT=$(gcloud config list project --format "value(core.project)")
         echo "Your current GCP Project Name is: "$PROJECT
```

Your current GCP Project Name is: qwiklabs-gcp-01-44a3211a105a

#### The source dataset

Our dataset is hosted in BigQuery. The CDC's Natality data has details on US births from 1969 to 2008 and is a publically available dataset, meaning anyone with a GCP account has access. Click here to access the dataset.

The natality dataset is relatively large at almost 138 million rows and 31 columns, but simple to understand. weight\_pounds is the target, the continuous value we'll train a model to predict.

## Create a BigQuery Dataset and Google Cloud Storage Bucket

A BigQuery dataset is a container for tables, views, and models built with BigQuery ML. Let's create one called **babyweight**. We'll do the same for a GCS bucket for our project too.

```
In [7]:
         %%bash
         # Create a BiqQuery dataset for babyweight if it doesn't exist
         datasetexists=$(bq ls -d | grep -w babyweight)
         if [ -n "$datasetexists" ]; then
             echo -e "BigQuery dataset already exists, let's not recreate it."
         else
             echo "Creating BigQuery dataset titled: babyweight"
             bq --location=US mk --dataset \
                 --description "Babyweight" \
                 $PROJECT: babyweight
             echo "Here are your current datasets:"
             bq ls
         fi
         ## Create GCS bucket if it doesn't exist already...
         exists=$(gsutil ls -d | grep -w gs://${BUCKET}/)
         if [ -n "$exists" ]; then
             echo -e "Bucket exists, let's not recreate it."
         else
             echo "Creating a new GCS bucket."
             gsutil mb -l ${REGION} gs://${BUCKET}
             echo "Here are your current buckets:"
             gsutil ls
         fi
```

```
Creating BigQuery dataset titled: babyweight

Dataset 'qwiklabs-gcp-01-44a3211a105a:babyweight' successfully created.

Here are your current datasets:

datasetId
```

```
babyweight
Bucket exists, let's not recreate it.
```

## Create the training and evaluation data tables

Since there is already a publicly available dataset, we can simply create the training and evaluation data tables using this raw input data. First we are going to create a subset of the data limiting our columns to weight\_pounds, is\_male, mother\_age, plurality, and gestation\_weeks as well as some simple filtering and a column to hash on for repeatable splitting.

 Note: The dataset in the create table code below is the one created previously, e.g. "babyweight".

#### Preprocess and filter dataset

We have some preprocessing and filtering we would like to do to get our data in the right format for training.

#### Preprocessing:

- Cast is\_male from BOOL to STRING
- Cast plurality from INTEGER to STRING where [1, 2, 3, 4, 5] becomes
   ["Single(1)", "Twins(2)", "Triplets(3)", "Quadruplets(4)",
   "Quintuplets(5)"]
- Add hashcolumn hashing on year and month

#### Filtering:

- Only want data for years later than 2000
- Only want baby weights greater than 0
- Only want mothers whose age is greater than 0
- Only want plurality to be greater than 0
- Only want the number of weeks of gestation to be greater than 0

```
In [8]:
```

```
%%bigquery
CREATE OR REPLACE TABLE
   babyweight.babyweight data AS
SELECT
   weight pounds,
   CAST(is male AS STRING) AS is male,
   mother age,
        WHEN plurality = 1 THEN "Single(1)"
        WHEN plurality = 2 THEN "Twins(2)"
        WHEN plurality = 3 THEN "Triplets(3)"
        WHEN plurality = 4 THEN "Quadruplets(4)"
        WHEN plurality = 5 THEN "Quintuplets(5)"
   END AS plurality,
    gestation weeks,
   FARM_FINGERPRINT(
        CONCAT (
```

```
CAST(year AS STRING),
CAST(month AS STRING)

)

) AS hashmonth

FROM

publicdata.samples.natality

WHERE

year > 2000

AND weight_pounds > 0

AND mother_age > 0

AND plurality > 0

AND gestation_weeks > 0
```

```
Query complete after 0.00s: 100% | 3/3 [00:00<00:00, 1822.55query/s]
Out[8]: —
```

#### Augment dataset to simulate missing data

Now we want to augment our dataset with our simulated babyweight data by setting all gender information to Unknown and setting plurality of all non-single births to Multiple(2+).

```
In [9]:
         %%bigquery
         CREATE OR REPLACE TABLE
             babyweight.babyweight_augmented_data AS
         SELECT
             weight_pounds,
             is male,
             mother age,
             plurality,
             gestation weeks,
             hashmonth
         FROM
             babyweight.babyweight data
         UNION ALL
         SELECT
             weight pounds,
             "Unknown" AS is male,
             mother age,
             CASE
                  WHEN plurality = "Single(1)" THEN plurality
                 ELSE "Multiple(2+)"
             END AS plurality,
             gestation weeks,
             hashmonth
         FROM
             babyweight.babyweight_data
```

```
Query complete after 0.01s: 100% 3/3 [00:00<00:00, 320.25query/s]
```

### Split augmented dataset into train and eval sets

Using hashmonth, apply a module to get approximately a 75/25 train-eval split.

#### Split augmented dataset into train dataset

Query complete after 0.00s: 100% | 3/3 [00:00<00:00, 1651.73query/s]

#### Split augmented dataset into eval dataset

## Verify table creation

Out[11]: -

Verify that you created the dataset and training data table.

```
In [12]:

**bigquery
-- LIMIT 0 is a free query; this allows us to check that the table exists.

SELECT * FROM babyweight_babyweight_data_train
LIMIT 0

Query complete after 0.00s: 100% | 1/1 [00:00<00:00, 555.61query/s]
Downloading: 0rows [00:00, ?rows/s]

Out[12]: weight_pounds is_male mother_age plurality gestation_weeks

In [13]:

**bigquery
-- LIMIT 0 is a free query; this allows us to check that the table exists.

SELECT * FROM babyweight_babyweight_data_eval
LIMIT 0
```

```
Query complete after 0.00s: 100% | 1/1 [00:00<00:00, 446.30query/s]
Downloading: 0rows [00:00, ?rows/s]

Out[13]: weight_pounds is_male mother_age plurality gestation_weeks
```

## **Export from BigQuery to CSVs in GCS**

Use BigQuery Python API to export our train and eval tables to Google Cloud Storage in the CSV format to be used later for TensorFlow/Keras training. We'll want to use the dataset we've been using above as well as repeat the process for both training and evaluation data.

```
In [14]:
          # Construct a BigQuery client object.
          client = bigquery.Client()
          dataset name = "babyweight"
          # Create dataset reference object
          dataset_ref = client.dataset(
              dataset_id=dataset_name, project=client.project)
          # Export both train and eval tables
          for step in ["train", "eval"]:
              destination uri = os.path.join(
                  "gs://", BUCKET, dataset_name, "data", "{}*.csv".format(step))
              table_name = "babyweight_data_{}".format(step)
              table ref = dataset ref.table(table name)
              extract job = client.extract table(
                  table ref,
                  destination_uri,
                  # Location must match that of the source table.
                  location="US",
                # API request
              extract job.result() # Waits for job to complete.
              print("Exported {}:{}.{} to {}".format(
                  client.project, dataset name, table name, destination uri))
```

Exported qwiklabs-gcp-01-44a3211a105a:babyweight.babyweight\_data\_train to gs://q wiklabs-gcp-01-44a3211a105a/babyweight/data/train\*.csv

Exported qwiklabs-gcp-01-44a3211a105a:babyweight.babyweight\_data\_eval to gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/eval\*.csv

## Verify CSV creation

Verify that we correctly created the CSV files in our bucket.

```
In [15]: %%bash gsutil ls gs://${BUCKET}/babyweight/data/*.csv

gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/eval000000000000.csv
gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/eval0000000000001.csv
gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/train000000000000.csv
gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/train000000000001.csv
gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/train0000000000002.csv
gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/train000000000003.csv
```

gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/train0000000000004.csvgs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/train000000000005.csv

#### Check data exists

Verify that you previously created CSV files we'll be using for training and evaluation.

```
In [16]: %%bash gsutil ls gs://${BUCKET}/babyweight/data/*00000000000csv
```

gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/eval000000000000.csv gs://qwiklabs-gcp-01-44a3211a105a/babyweight/data/train00000000000.csv

## Training on Cloud AI Platform

Now that we see everything is working locally, it's time to train on the cloud!

To submit to the Cloud we use gcloud ai-platform jobs submit training [jobname] and simply specify some additional parameters for AI Platform Training Service:

- jobname: A unique identifier for the Cloud job. We usually append system time to ensure uniqueness
- job-dir: A GCS location to upload the Python package to
- runtime-version: Version of TF to use.
- python-version: Version of Python to use. Currently only Python 3.7 is supported for TF 2.3.
- region: Cloud region to train in. See here for supported Al Platform Training Service regions

Below the -- \ we add in the arguments for our task.py file.

```
In [17]:
          %%bash
          OUTDIR=gs://${BUCKET}/babyweight/trained model
          JOBID=babyweight $(date -u +%y%m%d %H%M%S)
          gcloud ai-platform jobs submit training ${JOBID} \
              --region=${REGION} \
              --module-name=trainer.task \
              --package-path=$(pwd)/babyweight/trainer \
              --job-dir=${OUTDIR} \
              --staging-bucket=gs://${BUCKET} \
              --master-machine-type=n1-standard-8 \
              --scale-tier=CUSTOM \
              --runtime-version=${TFVERSION} \
              --python-version=${PYTHONVERSION} \
              --train data path=gs://${BUCKET}/babyweight/data/train*.csv \
              --eval data path=gs://${BUCKET}/babyweight/data/eval*.csv \
              --output dir=${OUTDIR} \
              --num epochs=10 \
              --train examples=10000 \
              --eval steps=100 \
              --batch size=32 \
              --nembeds=8
```

ERROR: (gcloud.ai-platform.jobs.submit.training) FAILED PRECONDITION: Field: job dir Error: The provided GCS path qs://qwiklabs-qcp-01-44a3211a105a/babyweight/t rained model cannot be written by service account service-204959009116@cloud-ml. google.com.iam.gserviceaccount.com. - '@type': type.googleapis.com/google.rpc.BadRequest fieldViolations: - description: The provided GCS path qs://qwiklabs-qcp-01-44a3211a105a/babywei ght/trained model cannot be written by service account service-204959009116@cloud-ml.google. com.iam.gserviceaccount.com. field: job\_dir CalledProcessError Traceback (most recent call last) /tmp/ipykernel 19723/2586862429.py in <module> ----> 1 get\_ipython().run\_cell\_magic('bash', '', '\nOUTDIR=gs://\${BUCKET}/babywe ight/trained\_model\nJOBID=babyweight\_\$(date -u +%y%m%d\_%H%M%S)\n\ngcloud ai-plat form jobs submit training \${JOBID} \\\n --region=\${REGION} \\\n ame=trainer.task \\\n --package-path=\$(pwd)/babyweight/trainer \\\n --jobdir=\${OUTDIR} \\n --staging-bucket=gs://\${BUCKET} \\n --master-machine-t ype=n1-standard-8 \\\n --scale-tier=CUSTOM \\\n --runtime-version=\${TFVERS ION} \\\n --python-version=\${PYTHONVERSION} \\\n -- \\\n --train\_data\_p ath=gs://\${BUCKET}/babyweight/data/train\*.csv \\n --eval data path=gs://\${BU CKET}/babyweight/data/eval\*.csv \\n --output\_dir=\${OUTDIR} \\\n --num\_epo chs=10 \\n --train\_examples=10000 \\n --eval\_steps=100 \\n ize=32 \\\n --nembeds=8\n') /opt/conda/lib/python3.7/site-packages/IPython/core/interactiveshell.py in run c ell\_magic(self, magic\_name, line, cell) 2401 with self.builtin trap: 2402 args = (magic arg s, cell) result = fn(\*args, \*\*kwargs) -> 2403 return result 2404 2405 /opt/conda/lib/python3.7/site-packages/IPython/core/magics/script.py in named sc ript magic(line, cell) 140 else: 141 line = script --> 142 return self.shebang(line, cell) 143 144 # write a basic docstring: /opt/conda/lib/python3.7/site-packages/decorator.py in fun(\*args, \*\*kw) 230 if not kwsyntax: 231 args, kw = fix(args, kw, sig)--> 232 return caller(func, \*(extras + args), \*\*kw) fun. name \_ = func.\_\_name\_\_ 233 fun.\_\_doc\_\_ = func.\_\_doc\_\_ /opt/conda/lib/python3.7/site-packages/IPython/core/magic.py in <lambda>(f, \*a, \*\*k) 185 # but it's overkill for just that one bit of state. 186 def magic deco(arg): call = lambda f, \*a, \*\*k: f(\*a, \*\*k) --> 187 188 189 if callable(arg):

/opt/conda/lib/python3.7/site-packages/IPython/core/magics/script.py in shebang

if args.raise\_error and p.returncode!=0:

sys.stderr.flush()

(self, line, cell)

243

```
--> 245
                   raise CalledProcessError(p.returncode, cell, output=out, std
err=err)
    246
   247
           def _run_script(self, p, cell, to_close):
CalledProcessError: Command 'b'\nOUTDIR=gs://${BUCKET}/babyweight/trained_model
\nJOBID=babyweight $(date -u +%y%m%d %H%M%S)\n\ngcloud ai-platform jobs submit t
raining ${JOBID} \\\n
                      --region=${REGION} \\n
                                                --module-name=trainer.task
 \\\n
        --package-path=$(pwd)/babyweight/trainer \\n
                                                        --job-dir=${OUTDIR}
 \\\n
        --staging-bucket=gs://${BUCKET} \\n
                                               --master-machine-type=n1-standa
            --scale-tier=CUSTOM \\\n
rd-8 \\\n
                                     --runtime-version=${TFVERSION} \\n
 --python-version=${PYTHONVERSION} \\n -- \\n
                                                    --train data path=gs://${B
UCKET}/babyweight/data/train*.csv \\n
                                       --eval data path=gs://${BUCKET}/babywe
ight/data/eval*.csv \\n
                          --output dir=${OUTDIR} \\\n
                                                         --num epochs=10 \\\n
--train examples=10000 \\n --eval steps=100 \\n
                                                      --batch size=32 \\\n
 --nembeds=8\n'' returned non-zero exit status 1.
```

The training job should complete within 15 to 20 minutes. You do not need to wait for this training job to finish before moving forward in the notebook, but will need a trained model.

#### Check our trained model files

Let's check the directory structure of our outputs of our trained model in folder we exported. We'll want to deploy the saved\_model.pb within the timestamped directory as well as the variable values in the variables folder. Therefore, we need the path of the timestamped directory so that everything within it can be found by Cloud AI Platform's model deployment service.

```
In [18]:
         %%bash
         gsutil ls gs://${BUCKET}/babyweight/trained model
         CommandException: One or more URLs matched no objects.
          -----
         CalledProcessError
                                                  Traceback (most recent call last)
         /tmp/ipykernel 19723/3083723745.py in <module>
         ---> 1 get_ipython().run_cell_magic('bash', '', 'gsutil ls gs://${BUCKET}/babyw
         eight/trained model\n')
         /opt/conda/lib/python3.7/site-packages/IPython/core/interactiveshell.py in run c
         ell_magic(self, magic_name, line, cell)
            2401
                            with self.builtin trap:
            2402
                                args = (magic arg s, cell)
         -> 2403
                                result = fn(*args, **kwargs)
           2404
                            return result
            2405
         /opt/conda/lib/python3.7/site-packages/IPython/core/magics/script.py in named sc
         ript magic(line, cell)
             140
                            else:
                                line = script
             141
         --> 142
                            return self.shebang(line, cell)
             143
             144
                        # write a basic docstring:
         /opt/conda/lib/python3.7/site-packages/decorator.py in fun(*args, **kw)
             230
                            if not kwsyntax:
             231
                                args, kw = fix(args, kw, sig)
                            return caller(func, *(extras + args), **kw)
```

```
fun. name__ = func.__name__
             233
             234
                     fun.__doc__ = func.__doc__
         /opt/conda/lib/python3.7/site-packages/IPython/core/magic.py in <lambda>(f, *a,
                     # but it's overkill for just that one bit of state.
             185
             186
                     def magic deco(arg):
                         call = lambda f, *a, **k: f(*a, **k)
         --> 187
             188
             189
                         if callable(arg):
         /opt/conda/lib/python3.7/site-packages/IPython/core/magics/script.py in shebang
         (self, line, cell)
             243
                             sys.stderr.flush()
             244
                         if args.raise_error and p.returncode!=0:
         --> 245
                             raise CalledProcessError(p.returncode, cell, output=out, std
         err=err)
             246
             247
                     def run script(self, p, cell, to close):
         CalledProcessError: Command 'b'gsutil ls gs://${BUCKET}/babyweight/trained_model
         \n'' returned non-zero exit status 1.
In [19]:
          %%bash
          MODEL_LOCATION=$(gsutil ls -ld -- gs://${BUCKET}/babyweight/trained_model/2* \
                            tail -1)
          gsutil ls ${MODEL_LOCATION}
         gs://qwiklabs-gcp-01-44a3211a105a/
```

## Deploy trained model

Deploying the trained model to act as a REST web service is a simple gcloud call.

CommandException: One or more URLs matched no objects.

```
In [20]:
          %%bash
          gcloud config set ai platform/region global
         Updated property [ai platform/region].
In [21]:
          %%bash
          MODEL NAME="babyweight"
          MODEL VERSION="ml on gcp"
          MODEL LOCATION=$(gsutil ls -ld -- gs://${BUCKET}/babyweight/trained model/2* \
                            | tail -1 | tr -d '[:space:]')
          echo "Deleting and deploying $MODEL NAME $MODEL VERSION from $MODEL LOCATION"
          # qcloud ai-platform versions delete ${MODEL VERSION} --model ${MODEL NAME}
          # gcloud ai-platform models delete ${MODEL NAME}
          gcloud ai-platform models create ${MODEL NAME} --regions ${REGION}
          gcloud ai-platform versions create ${MODEL VERSION} \
              --model=${MODEL NAME} \
              --origin=${MODEL LOCATION} \
              --runtime-version=2.3 \
              --python-version=3.7
```

Deleting and deploying babyweight ml\_on\_gcp from

CommandException: One or more URLs matched no objects.

Using endpoint [https://ml.googleapis.com/]

```
Created ai platform model [projects/qwiklabs-gcp-01-44a3211a105a/models/babyweig
ht].
Using endpoint [https://ml.googleapis.com/]
ERROR: (gcloud.ai-platform.versions.create) Either `--origin` must be provided o
r `deploymentUri` must be provided in the file given by `--config`.
CalledProcessError
                                          Traceback (most recent call last)
/tmp/ipykernel_19723/281812598.py in <module>
----> 1 get_ipython().run_cell_magic('bash', '', 'MODEL_NAME="babyweight"\nMODEL
_VERSION="ml_on_gcp"\nMODEL_LOCATION=$(gsutil ls -ld -- gs://${BUCKET}/babyweigh
                                        | tail -1 | tr -d \'[:space:]\')\necho
t/trained model/2* \\n
 "Deleting and deploying $MODEL NAME $MODEL VERSION from $MODEL LOCATION"\n# gcl
oud ai-platform versions delete ${MODEL_VERSION} --model ${MODEL_NAME}\n# gcloud
ai-platform models delete ${MODEL NAME}\ngcloud ai-platform models create ${MODE
L_NAME } -- regions ${REGION} \ ngcloud ai-platform versions create ${MODEL_VERSION}
\\\n --model=${MODEL_NAME} \\\n --origin=${MODEL_LOCATION} \\\n
me-version=2.3 \\\n
                      --python-version=3.7\n')
/opt/conda/lib/python3.7/site-packages/IPython/core/interactiveshell.py in run c
ell magic(self, magic name, line, cell)
   2401
                   with self.builtin_trap:
   2402
                        args = (magic_arg_s, cell)
-> 2403
                        result = fn(*args, **kwargs)
   2404
                   return result
   2405
/opt/conda/lib/python3.7/site-packages/IPython/core/magics/script.py in named sc
ript magic(line, cell)
    140
                    else:
    141
                        line = script
--> 142
                    return self.shebang(line, cell)
    143
    144
                # write a basic docstring:
/opt/conda/lib/python3.7/site-packages/decorator.py in fun(*args, **kw)
                    if not kwsyntax:
    230
    231
                        args, kw = fix(args, kw, sig)
--> 232
                    return caller(func, *(extras + args), **kw)
    233
            fun. name = func. name
    234
            fun.__doc__ = func.__doc__
/opt/conda/lib/python3.7/site-packages/IPython/core/magic.py in <lambda>(f, *a,
 **k)
    185
            # but it's overkill for just that one bit of state.
    186
            def magic deco(arg):
                call = lambda f, *a, **k: f(*a, **k)
--> 187
    188
               if callable(arg):
/opt/conda/lib/python3.7/site-packages/IPython/core/magics/script.py in shebang
(self, line, cell)
    243
                    sys.stderr.flush()
    244
                if args.raise error and p.returncode!=0:
--> 245
                    raise CalledProcessError(p.returncode, cell, output=out, std
err=err)
    246
            def run script(self, p, cell, to close):
```

CalledProcessError: Command 'b'MODEL NAME="babyweight"\nMODEL VERSION="ml on gc

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