

[544] BigQuery Machine Learning

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Learning Objectives

- write "CREATE MODEL" queries to train models on BigQuery query results
- use a BigQuery TRANSFORM clause to pre-process data prior to training
- use BigQuery's "ML.????" tabular functions to inspect models, make predictions, and evaluate performance

Outline

BigQuery ML Basics

Feature Transformation

Train/Test Split

BigQuery provides a `DATA_SPLIT_METHOD` config, but it's a bit unusual.

Default behavior depends on dataset

- <500 rows: 100% training data
- <50K rows: 80% training data
- bigger: 10K rows for test, rest for training

Documentation: "When there is a data split, you can find the temporary split results (Training Data, Evaluation Data) on the Model Details page in the BigQuery Console and the model API `data_split_result` field. **These split tables will be saved for 48 hours.** If you will need them for longer than 48 hours, copy them out of the anonymous dataset for longer retention."

Recommendation:

- split manually using `rand()<ratio` in SQL (`rand` gives num between 0 and 1)
- disable BigQuery splitting: **`DATA_SPLIT_METHOD="NO_SPLIT"`**

Training

Step 1: write a query to select both features and label

label
(to predict)

features

```
SELECT yesterday_temp, humidity, temp  
FROM weather
```

Training

Step 2: choose a model name and create it

```
CREATE OR REPLACE MODEL myproj.mydataset.mymodel  
OPTIONS (...) name
```

AS

```
SELECT yesterday_temp, humidity, temp  
FROM weather
```

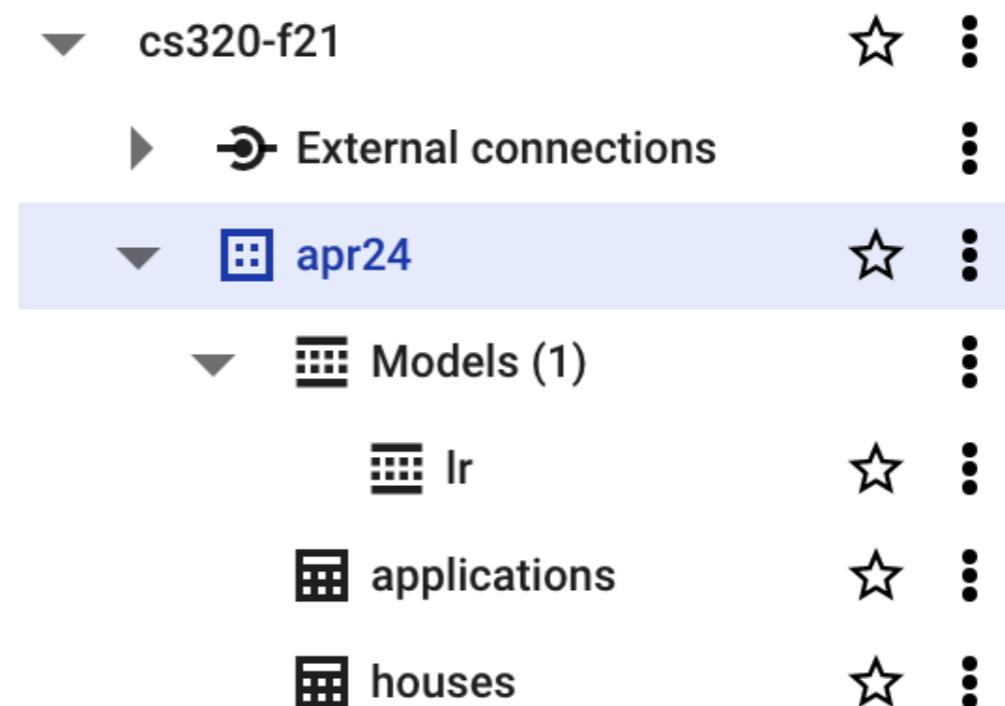
hierarchy:

projects

datasets

tables

models



Training

Step 3: choose type of model

```
CREATE OR REPLACE MODEL myproj.mydataset.mymodel
OPTIONS (MODEL_TYPE='LINEAR_REG')
```

AS

```
SELECT yesterday_temp, humidity, temp
FROM weather
```

Options: LINEAR_REG, LOGISTIC_REG, KMEANS, MATRIX_FACTORIZATION, PCA, AUTOENCODER, AUTOML_CLASSIFIER, AUTOML_REGRESSOR, BOOSTED_TREE_CLASSIFIER, BOOSTED_TREE_REGRESSOR, RANDOM_FOREST_CLASSIFIER, RANDOM_FOREST_REGRESSOR, DNN_CLASSIFIER, DNN_REGRESSOR, DNN_LINEAR_COMBINED_CLASSIFIER, DNN_LINEAR_COMBINED_REGRESSOR, ARIMA_PLUS, ARIMA_PLUS_XREG, TENSORFLOW, TENSORFLOW_LITE, ONNX, XGBOOST

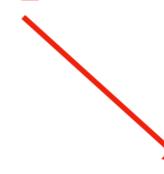
Training

Step 4: indicate label column (others are assumed features)

```
CREATE OR REPLACE MODEL myproj.mydataset.mymodel
OPTIONS (MODEL_TYPE='LINEAR_REG',
         INPUT_LABEL_COLS=['temp'])
```

AS

```
SELECT yesterday_temp, humidity, temp
FROM weather
```



Using Trained Models

Each of these functions return a table related to a model.

what are the coefficients used to multiply features?

`ML.WEIGHTS(MODEL ???)`

what are the predictions given the features?

`ML.PREDICT(MODEL ???, (????))`

SQL query to get features

how well do we predict (various metrics) given the features+label?

`ML.EVALUATE(MODEL ???, (????))`

SQL query to get features and label

Using Trained Models

Each of these functions return a table related to a model.

what are the coefficients used to multiply features?

ML.WEIGHTS(MODEL ???)

example:

```
SELECT *
FROM ML.WEIGHTS(MODEL mymodel)
```

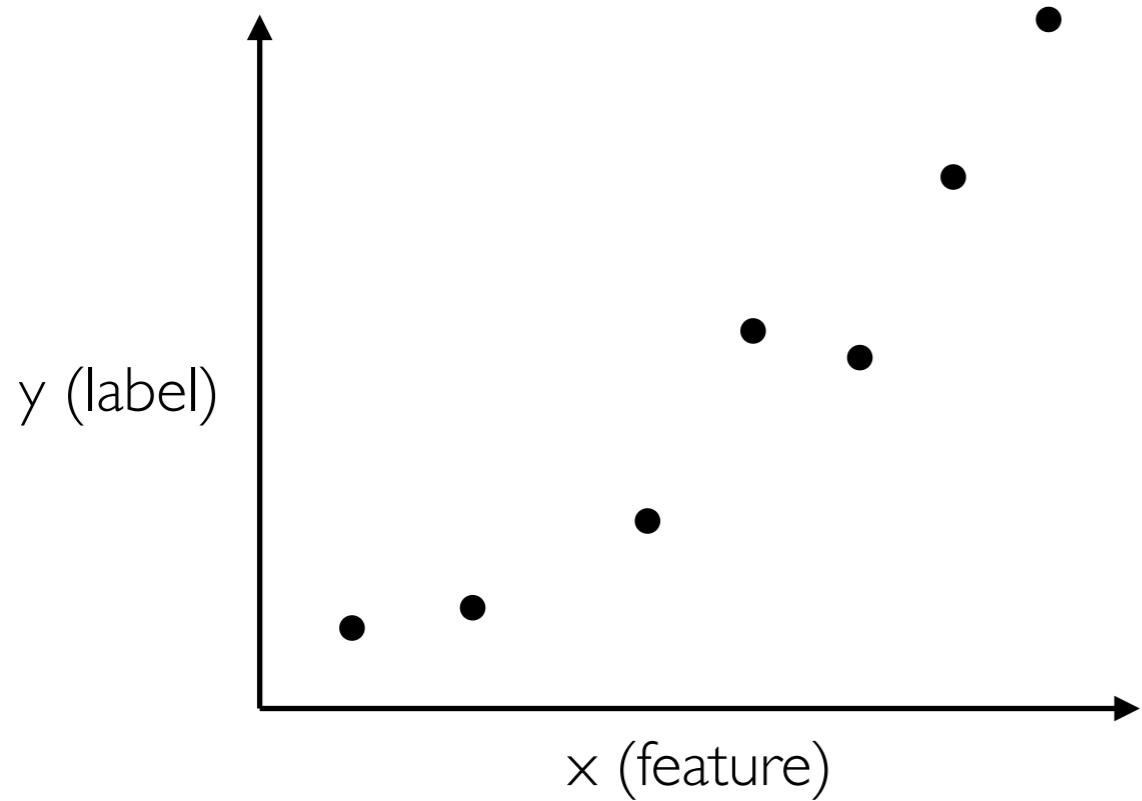
TopHat, Demos

Outline

BigQuery ML Basics

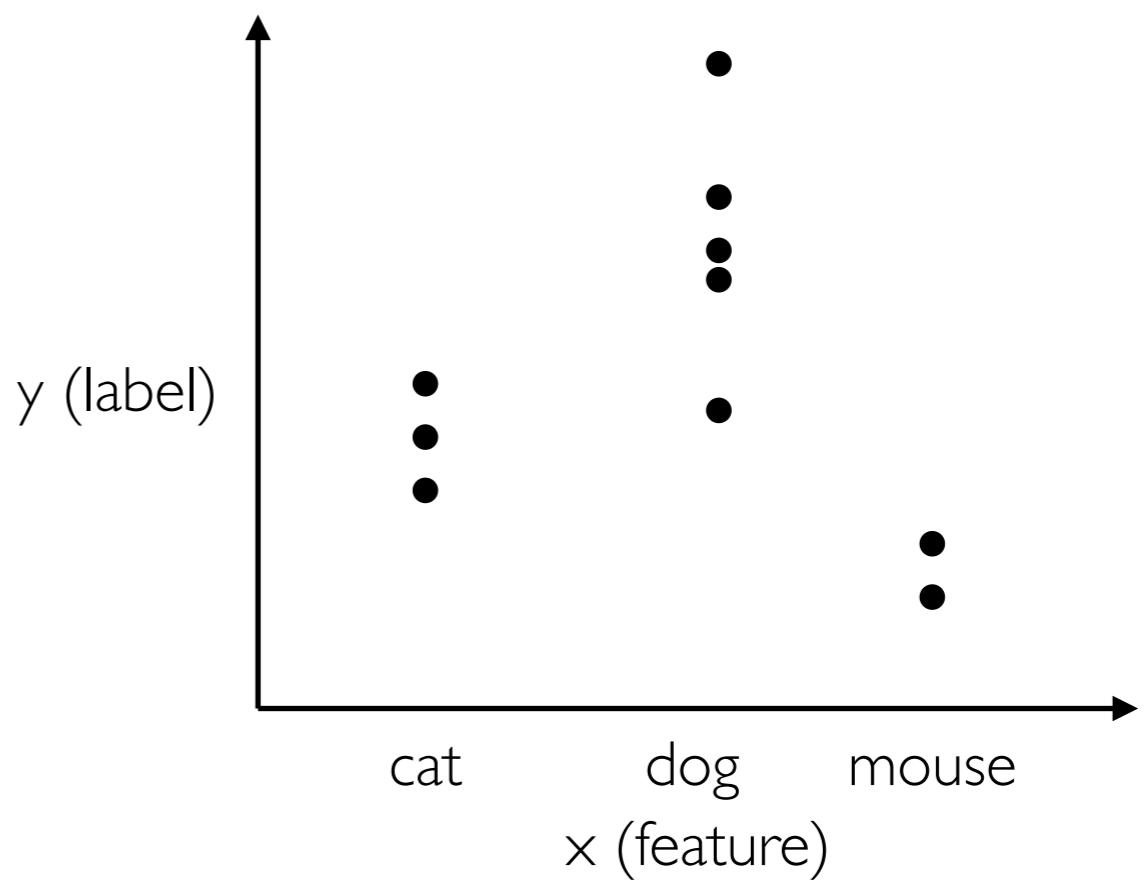
Feature Transformation

Patterns and Features



non-linear patterns

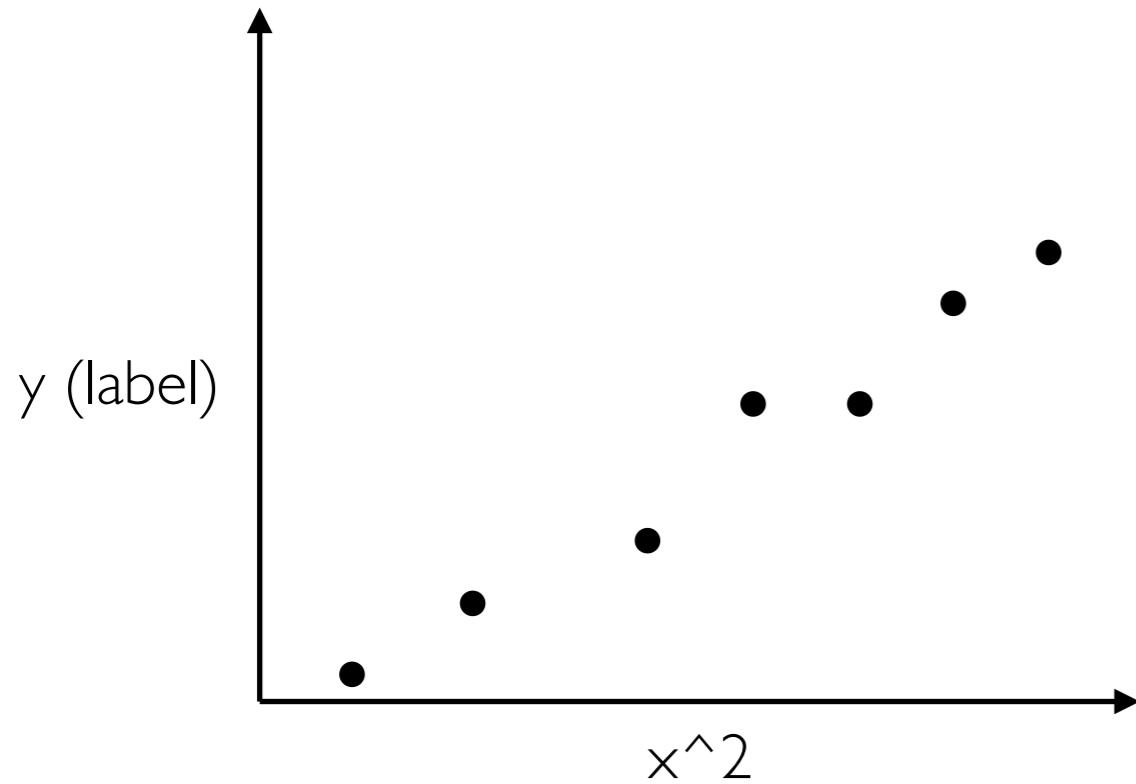
- some models (e.g., DNNs) naturally handle this
- others (e.g., LinearRegression) do not



categorical features

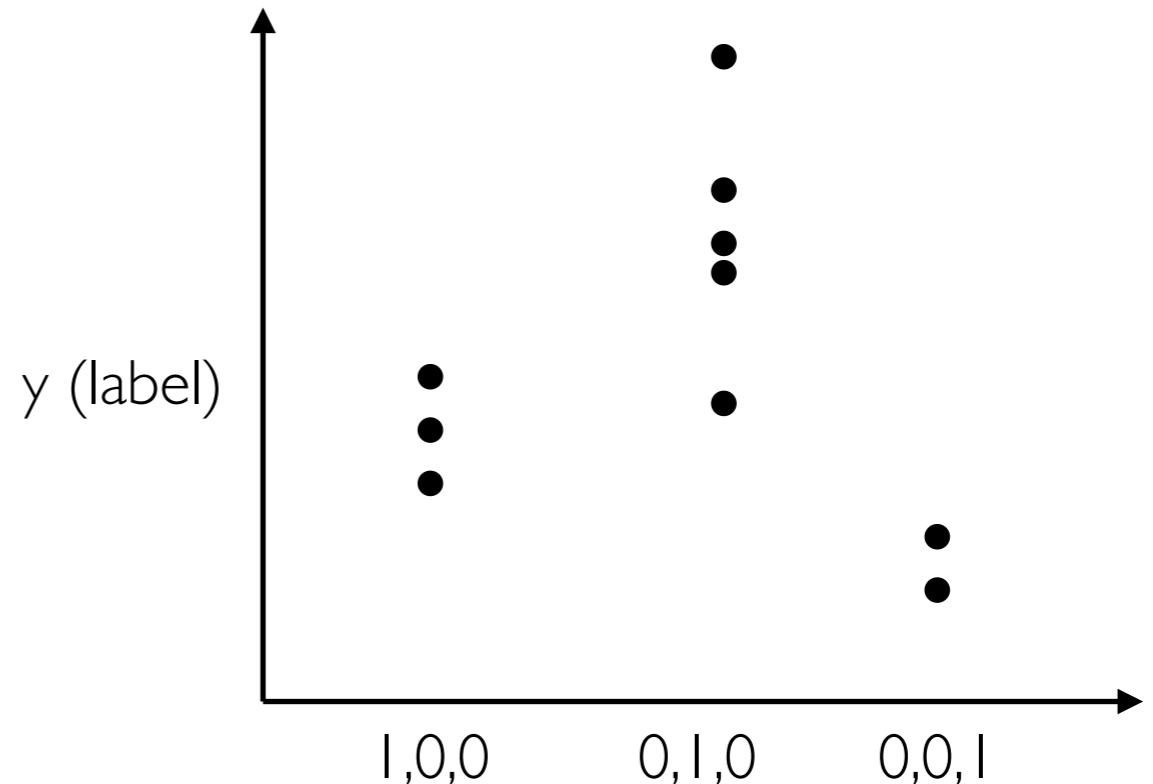
- some models (e.g., DTs) naturally handle this
- others (e.g., LinearRegression) do not

Feature Transformation



non-linear patterns

- can introduce new features than are computed as functions of originals (e.g., $x_2=x^2$)
- a linear model over the new features corresponds to a non-linear model over the originals



categorical features

- encode categorical features as numbers (e.g., as matrix of zeros and ones for OneHot encoding)

Demos