

# [544] BigQuery Machine Learning

Tyler Caraza-Harter

# 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 its 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

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

features label  
(to predict)

# 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
```

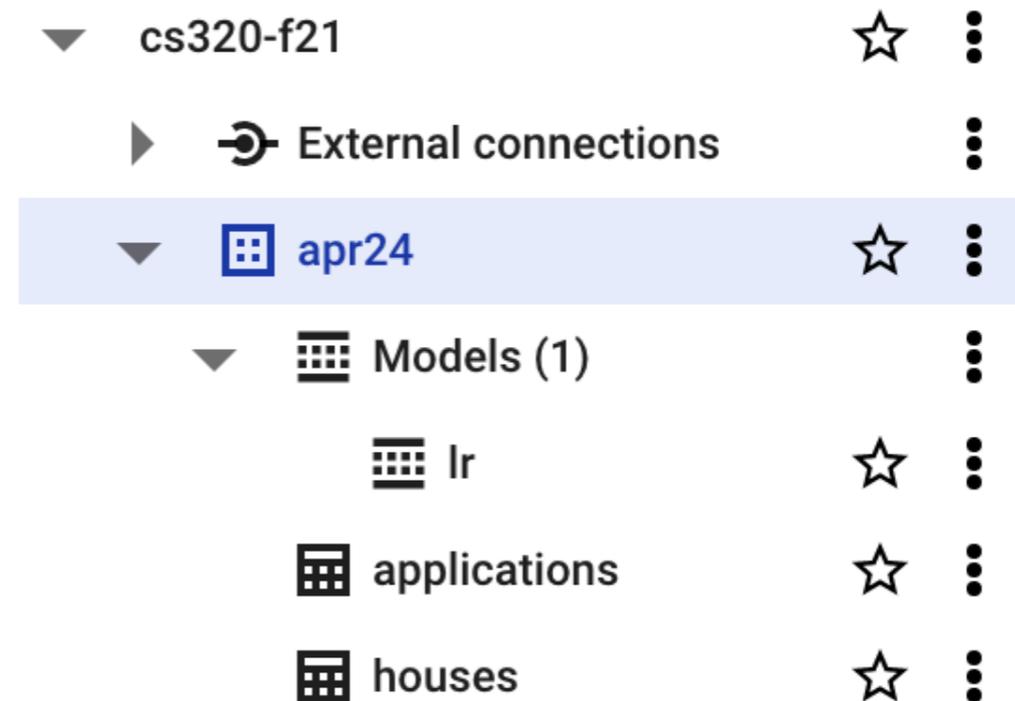
**hierarchy:**

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)
```

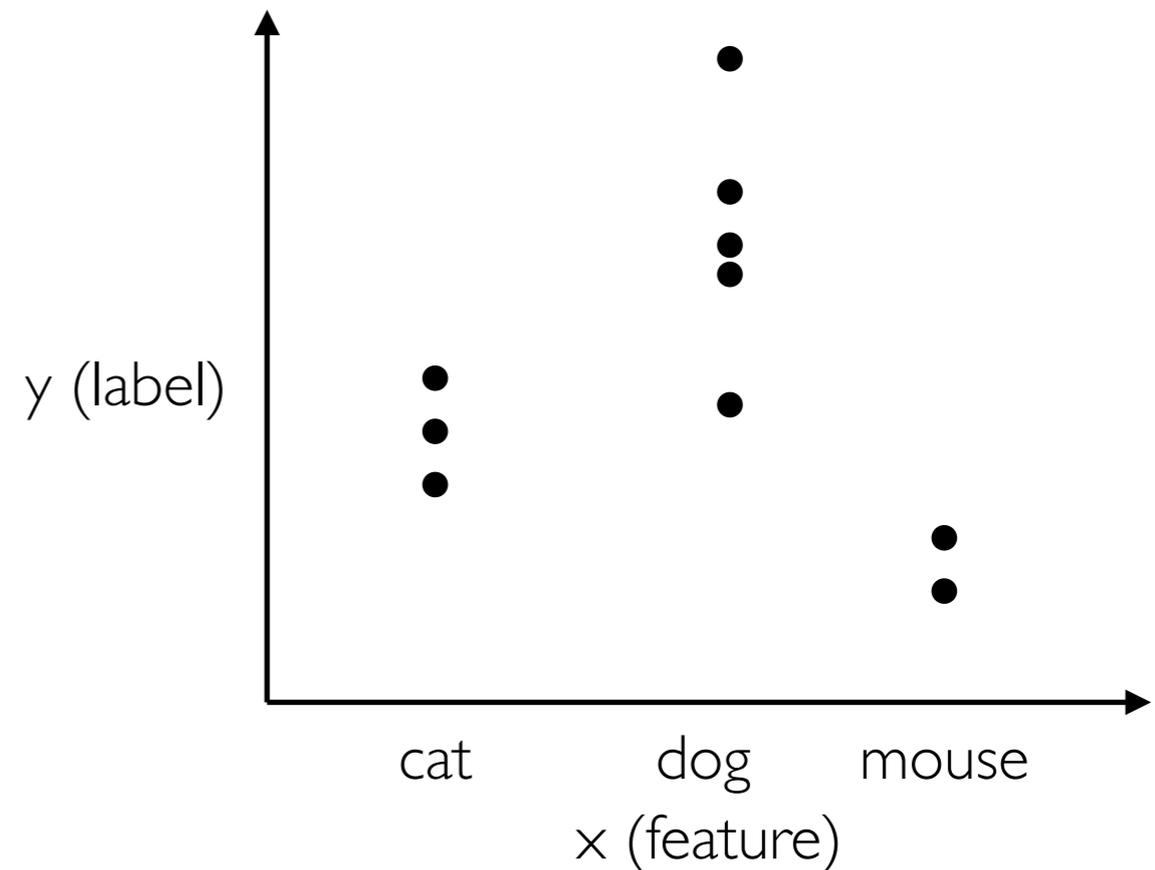
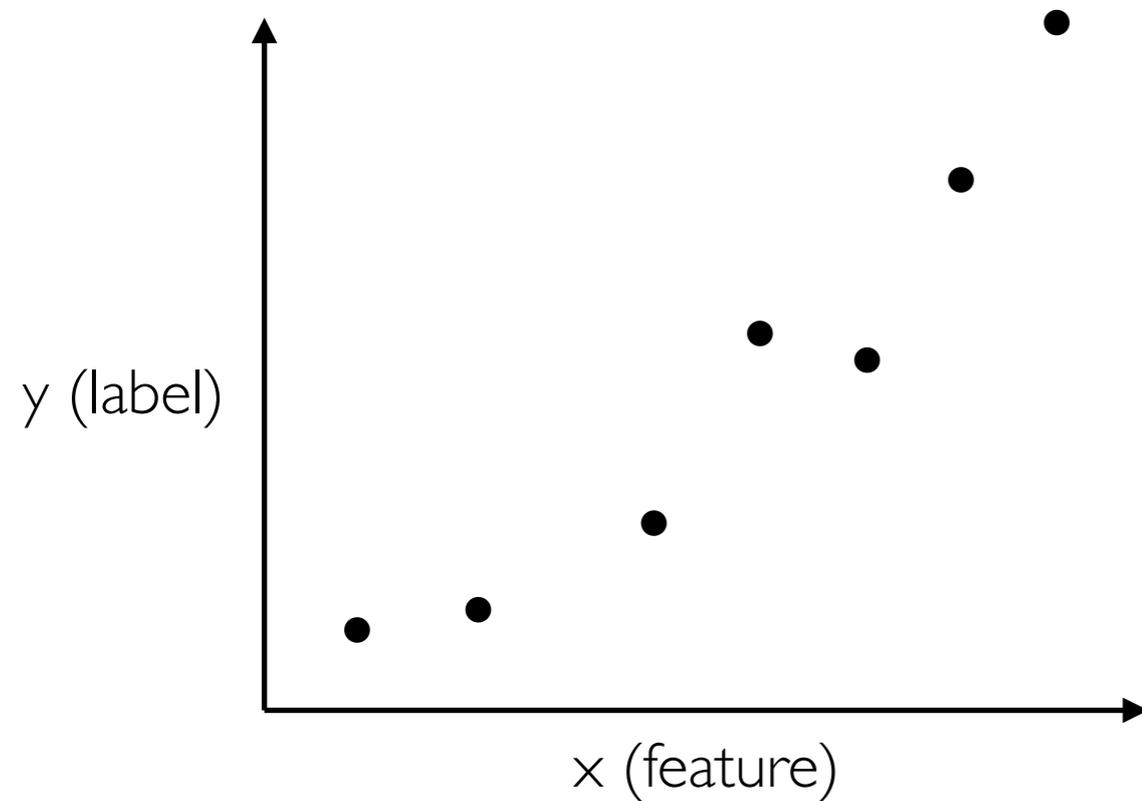
**TopHat, Demos**

# Outline

BigQuery ML Basics

Feature Transformation

# Patterns and Features



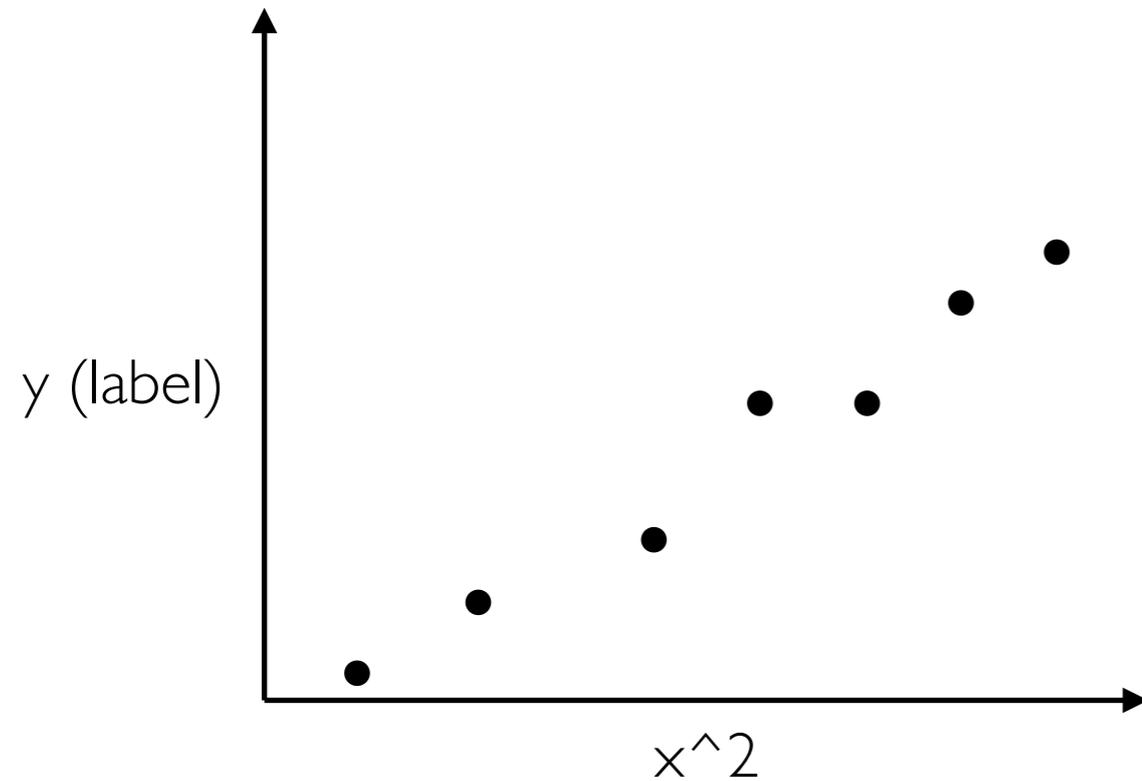
## non-linear patterns

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

## categorical features

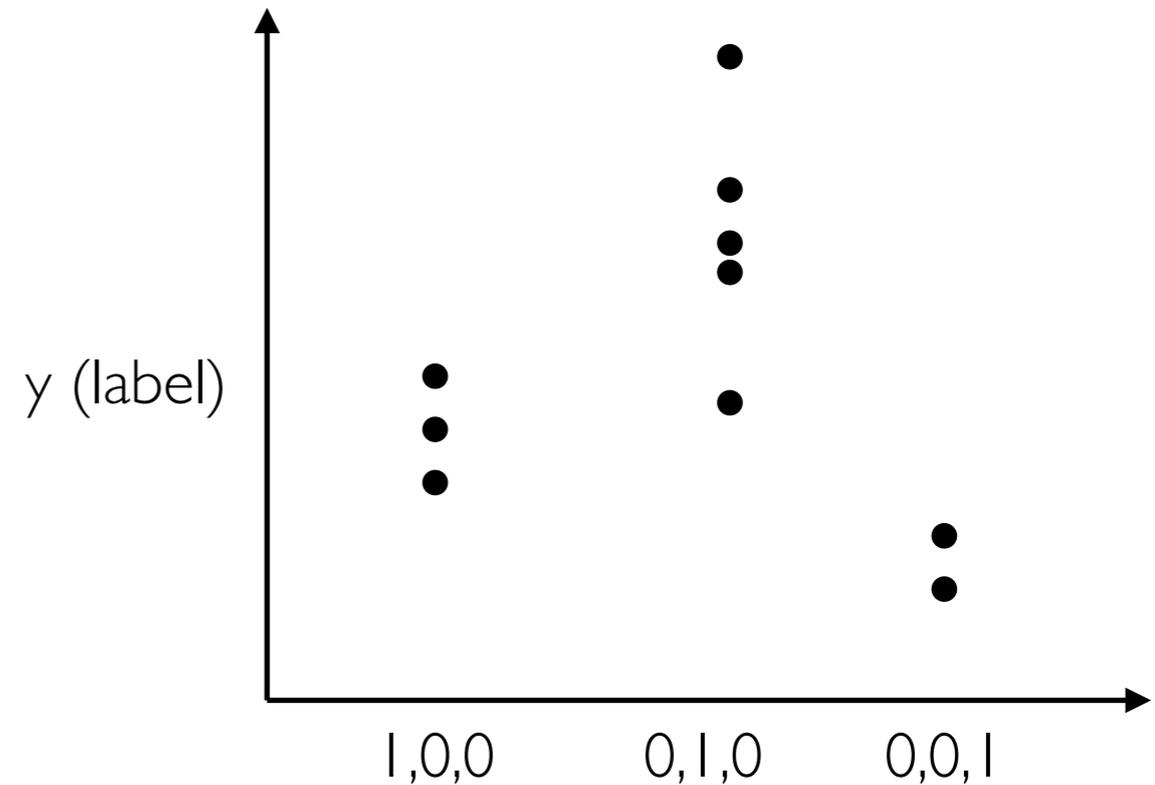
- some models (e.g., DTs) naturally handle this
- others (e.g., LinearRegression) do not

# 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**