

# **Project Introduction**

# Your Project for the Week

## Goals:

• Use Capital Bike Share Kaggle data set to predict hourly demand for bicycle rentals based on time and weather, e.g.

"Given the forecasted weather conditions, how many bicycles can we expect to be rented out (city-wide) this Saturday at 2pm?"

• Submit your predictions to Kaggle.

#### Notes:

- We're predicting a continuous variable, not a class!
- We're predicting the **count** column. It is a sum of **casual** and **registered** columns. The latter two should not be in our features!

### Data:

train.csv: Your dataset for this weeks project. This is the dataset you are training and evaluating your model on. *It contains the first 19 days of every month.* 

test.csv : The dataset Kaggle uses to evaluate your model. *It contains days 20 to end of month.* 

## **Evaluation:**

 The metric you are optimizing for this week is RMSLE, Root Mean Squared Log Error.

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- Unlike last week, smaller value is better!
- Penalizes under-estimates more than over-estimates
- These are the scores you can expect from the model you build:
  - EASY: ~1.1
  - MEDIUM: <0.9
  - MEDIUM-HARD: <0.7
  - HARD: <0.5
  - VERY HARD: <0.45

# **Machine Learning Workflow**

- 1. Define business goal:
  - Create a model that predicts bike sharing demand; optimize for RMSLE.
- 2. Get data:
  - From Kaggle, or week\_03/data folder on Github.
- 3. Train-test-split:
  - You already know this!
- 4. Explore data:
  - Find out what all the variables in your dataset mean
  - Extract time features from the DatetimeIndex
  - Inspect descriptive statistics of your features and the target variable
  - Visualize the relationships between single features and the target variable
  - Look at correlations
  - Think about how to use **ColumnTransformer** and **Pipeline** with your features
- 5. Train model:
  - This afternoon we'll look at linear regression

6-...

Project Introduction 2

• Later this week

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