

# River Flow Analysis

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### Objectives

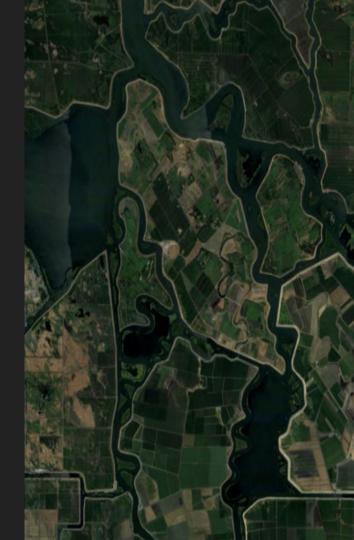
- 1. Outline steps I took to complete coding challenge.
- 2. Share findings and contextualize results.
- 3. What I learned and what I am working towards.

### Structure of Analysis

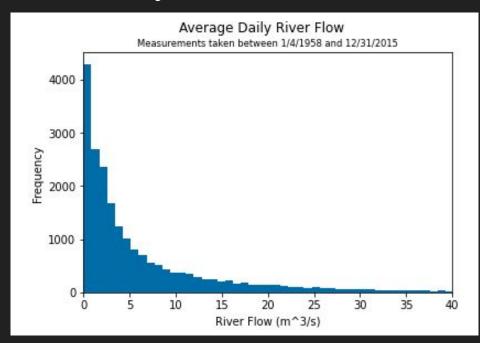
- Data Cleaning and Preparation (in R)
- 2. Summary Statistics and Data Transformation
- 3. Regression Analysis
- 4. Train ML Model

#### Future Steps:

- 5. Test Model
- 6. Apply to Client's Production Scenario



#### **Summary Statistics**



Important metrics for each variable:

- Median, Interquartile Range
- Outliers (min/max)
- Count (should all be the same)
- Data Types

Example histogram of daily river flow frequency

### The Challenge

Predict daily river flow from 9 surrounding temperature and precipitation measurement stations.

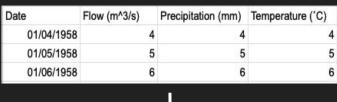
#### Potential applications include:

- Downstream agriculture ventures
- Water reservoirs
- Habitat restoration / Conservation efforts
- Academic research



#### Data Cleaning

- Used R to un-pickle the data, rename variables and merge the two datasets.
- 2. Derived precipitation and temperature values corresponding to N days prior for each date (row):





Date	Flow (m^3/s)	Precip_1	Temp_1	Precip_2	Temp_2
01/06/1958	6	5	5	4	4

#### Function to create new columns:

```
def derive_nth_day_feature(df, feature, N):
rows = df.shape[0]
nth_prior_measurements = [None]*N + [df[feature][i-N] for i in range(N, rows)]
col_name = "{}_{}".format(feature, N)
df[col_name] = nth_prior_measurements
```

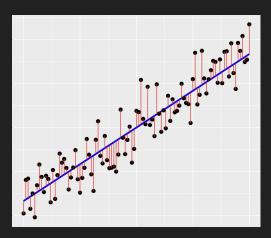
### **Linear Regression Model**

1. Select relevant independent variables.

For each river flow forecast, there are 27 precipitation and 27 temperature predictor variables.

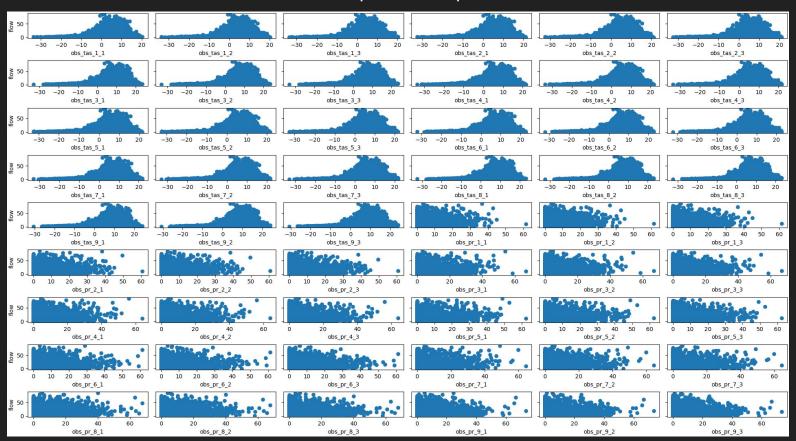
2. Assess Pearson correlation coefficients.

All 54 correlation coefficients have range [0.19 - 0.36]



#### Linear Regression (cont.)

#### Predictor Variable Relationships with Response Flow Variable



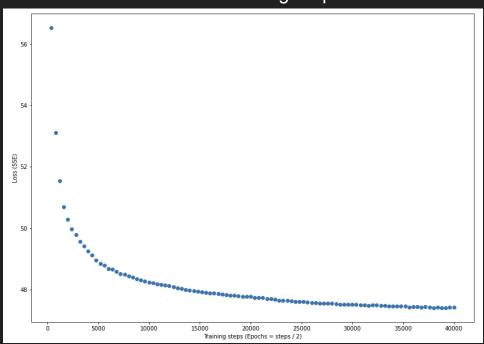
#### Training a ML Model

Using supervised machine learning methods to create a DNNRegressor:

- 1. More data cleaning!
- 2. Split data into (80%) training set, (10%) testing set, and (10%) validation set.
- 3. Instantiate neural network with 2 hidden layers.
- 4. Define reusable function to manage data input.
- 5. Export model.

#### Sources of Error

Loss SSE vs. Training Steps Plot



Indicates that the model was not overfitted since the evaluation losses never exhibit a significant change in direction toward an increasing value.

Bias?

Variance?

### Future Steps

- 1. Create a more robust linear regression model.
- Extract tensors from TensorFlow object.
- 3. Back-test data to assess model efficacy.
- 4. Deliver usable product to client.
- 5. Receive feedback.

#### What I am working towards:

- 1. Gain more hands-on ML experience
- 2. Develop stronger foundation in regression analysis
- 3. Complete Codecademy data science path / other online courses
- 4. Specialize in climate analytics / environmental data science applications



## Thank You!

Questions?