Assignment 5 Notebook

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Test Setup details:

Jupyter verison: 4.3.0

```
In [2]: import pandas as pd
# numpy has a lots of useful math related modules
import numpy as np
# Helpful function to display intermittent result
from IPython.display import display
```

Step 0: Take a peek at the dataset

```
In [3]: # Load the data and display only first five rows
# NOTE replace the following value with the actual path to the csv fil
e
data_file = "AQI.csv" # e.g ~/cs564/p5/AQI.csv"
display(pd.read_csv(data_file, nrows=5).head())
```

	State Code	County Code	Latitude	Longitude	Date Local	AQI	Address	State Name	County Name	(
0	55	59	42.504722	-87.8093	1997- 04-28	82	CHIWAUKEE PRAIRIE, 11838 FIRST COURT	Wisconsin	Kenosha	F
1	55	59	42.504722	-87.8093	1997- 05-10	33	CHIWAUKEE PRAIRIE, 11838 FIRST COURT	Wisconsin	Kenosha	F
2	55	59	42.504722	-87.8093	1997- 05-16	25	CHIWAUKEE PRAIRIE, 11838 FIRST COURT	Wisconsin	Kenosha	F
3	55	59	42.504722	-87.8093	1997- 05-22	25	CHIWAUKEE PRAIRIE, 11838 FIRST COURT	Wisconsin	Kenosha	F
4	55	59	42.504722	-87.8093	1997- 05-28	46	CHIWAUKEE PRAIRIE, 11838 FIRST COURT	Wisconsin	Kenosha	F

Step 1 : Load the dataset

In [4]: from sqlalchemy import create_engine import sqlite3

create a database where we'll load the dataset from the csv file
db_conn = create_engine('sqlite://AirQualityIndex.db')

```
In [5]: import string
# load in a batch of 5000 tuples. Modify this value to your needs
chunks = 5000
for data in pd.read_csv(data_file, chunksize=chunks,
    iterator=True, encoding='utf-8'):

    data = data.rename(columns={col: col.replace('-', '') for col in dat
    a.columns})
    data = data.rename(columns={col: col.strip() for col in data.columns})
)
    data = data.rename(columns={col: string.capwords(col) for col in data
    .columns})
    data = data.rename(columns={col: col.replace('', '') for col in data
    .columns})

    data = data.rename(columns={col: col.replace('', '') for col in data
    .columns})
```

Step 2: Data Exploration

Find the average air quality index for each city recorded during the year 1997

	avg_latitude	city
0	43.020075	Waukesha
1	1 43.016667 Milwaukee	
2	42.504722	Pleasant Prairie

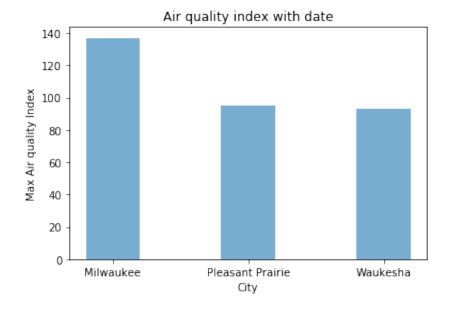
	avg_longitude	city	
0	-87.809300	Pleasant Prairie	
1	1 -87.933333 Milwaukee		
2	-88.215070	Waukesha	

	max_aqi	city	date	
0	137 Milwaukee		1997-04-04	
1	95	Pleasant Prairie	1997-10-07	
2	93	Waukesha	1997-04-28	

Step 3 : Data Visualization

Visualize the data you received above as plots

```
In [9]:
        # Use matplotlib library to plot graph - a simple barchart in this cas
        # We encourage you to explore different and more complex types of char
        import matplotlib.pyplot as plt
        # First we convert the Dataframe object to Python list (that Matplotli
        b understands)
        from pandas import DataFrame
        # this is our 'y-axis' value - obtained by flattening the DataFrame ob
        iect
        aqis list = list(max air_quality['max_aqi'].values.flatten())
        # This is our x-axis value
        cities = list(max air quality['city'].values.flatten())
        dates = list(max air quality['date'].values.flatten())
        # Specify the plot type and formatting
        plt.title('Air quality index with date')
        y pos = np.arange(len(cities))
        # Type of graph
        plt.bar(y pos, agis list, align='center', alpha=0.6, width=0.4)
        # Specify 'ticks' on the x-axis for each item in the list for 'x-axis'
        values
        plt.xticks(y_pos, cities)
        plt.xlabel('City')
        plt.ylabel('Max Air quality Index')
        # Save this diagram(by default at the directory from where this applic
        ation is launched)
        plt.savefig("max air quality.png")
```



Step 4: Optional (recommended): Build Machine Learning models

```
from sklearn.model selection import cross val predict
In [10]:
         from sklearn import linear model
         import matplotlib.pyplot as plt
         df = pd.read csv('AQI.csv')
         df = df.dropna()
         lr = linear model.LinearRegression()
         x = df['Latitude']
         y = df.AQI
         # cross val predict returns an array of the same size as `y` where eac
         # is a prediction obtained by cross validation:
         predicted = cross val predict(lr, x, y, cv=10)
         fig, ax = plt.subplots()
         ax.scatter(y, predicted)
         ax.plot([y.min(), y.max()], [y.min(), y.max()], 'k--', lw=4)
         ax.set xlabel('Date')
         ax.set ylabel('AQI')
         plt.show()
```

/Users/yuhanliu/anaconda/lib/python3.6/site-packages/IPython/core/in teractiveshell.py:2717: DtypeWarning: Columns (0) have mixed types. Specify dtype option on import or set low_memory=False.

interactivity=interactivity, compiler=compiler, result=result)
/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/utils/v
alidation.py:395: DeprecationWarning: Passing 1d arrays as data is d
eprecated in 0.17 and will raise ValueError in 0.19. Reshape your da
ta either using X.reshape(-1, 1) if your data has a single feature o
r X.reshape(1, -1) if it contains a single sample.

DeprecationWarning)

```
-----
```

/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/model_s election/ validation.py in cross val predict(estimator, X, y, groups

```
, cv, n jobs, verbose, fit params, pre dispatch, method)
            prediction blocks = parallel(delayed( fit and predict)(
    400
                clone(estimator), X, y, train, test, verbose, fit pa
rams, method)
--> 401
                for train, test in cv iter)
    402
    403
            # Concatenate the predictions
/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/externa
ls/joblib/parallel.py in call (self, iterable)
    756
                    # was dispatched. In particular this covers the
edge
    757
                    # case of Parallel used with an exhausted iterat
or.
--> 758
                    while self.dispatch one batch(iterator):
    759
                        self. iterating = True
    760
                    else:
/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/externa
ls/joblib/parallel.py in dispatch one batch(self, iterator)
    606
                        return False
    607
                    else:
--> 608
                        self. dispatch(tasks)
    609
                        return True
    610
/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/externa
ls/joblib/parallel.py in _dispatch(self, batch)
    569
                dispatch timestamp = time.time()
    570
                cb = BatchCompletionCallBack(dispatch timestamp, len
(batch), self)
--> 571
                job = self. backend.apply async(batch, callback=cb)
    572
                self. jobs.append(job)
    573
/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/externa
ls/joblib/ parallel backends.py in apply async(self, func, callback)
    107
            def apply async(self, func, callback=None):
                """Schedule a func to be run"""
    108
--> 109
                result = ImmediateResult(func)
    110
                if callback:
    111
                    callback(result)
/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/externa
ls/joblib/_parallel_backends.py in __init__(self, batch)
                # Don't delay the application, to avoid keeping the
    324
input
    325
                # arguments in memory
--> 326
                self.results = batch()
    327
```

328 def get(self):

/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/externa ls/joblib/parallel.py in __call__(self)

```
129
    130     def __call__(self):
--> 131          return [func(*args, **kwargs) for func, args, kwargs
in self.items]
    132
    133     def len (self):
```

/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/externa ls/joblib/parallel.py in stcomp>(.0)

```
129
    130     def __call__(self):
--> 131          return [func(*args, **kwargs) for func, args, kwargs
in self.items]
    132
    133     def len (self):
```

/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/model_s election/_validation.py in _fit_and_predict(estimator, X, y, train, test, verbose, fit_params, method)

/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/linear_model/base.py in fit(self, X, y, sample weight)

e_weight).ndim > 1:

/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/utils/v alidation.py in check_X_y(X, y, accept_sparse, dtype, order, copy, f orce_all_finite, ensure_2d, allow_nd, multi_output, ensure_min_sampl es, ensure_min_features, y_numeric, warn_on_dtype, estimator)

/Users/yuhanliu/anaconda/lib/python3.6/site-packages/sklearn/utils/v

Summary

- 1. Write a brief summary of your findings here.
- 2. Additionally explore by building a machine learning model using scikit-learn to make predictions and analysis

Finally, take a moment to reflect on what we achieved here - we started with a raw dataset, processed them, built models and ran analytics to finally gain insights from the data:)