

Cycle Data clean part 01-DBConnect

April 27, 2018

1 6.4 Cycle Data Munging

Declaration : The coding is used was abstract from Kevin mark ham youtube video serie, Introduction to machine learning with scikit-learn video series. You can find link under resources section.

2 6.4.1 Data Munging steps

1. Handling missing Data
2. Encoding, Decoding, and recoding of data
3. Handling Anomalous values
4. Transforming Data
5. Merging Dataset

3 Data Cleaning

```
In [2]: # load libraries and set styles, options
```

```
import os, csv
import numpy as np
import pandas as pd
import seaborn as sns
from IPython.display import HTML
import warnings; warnings.simplefilter('ignore')
```

```
In [3]: %matplotlib inline
```

2. Read and verify data

```
In [4]: # read in a CSV
```

```
df = pd.read_csv('C:/Users/mrferozi/Desktop/GitHub/Bike/cycle-share-dataset-original/tr
```

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 286857 entries, 0 to 286856
Data columns (total 12 columns):
trip_id      286857 non-null int64
```

```

starttime      286857 non-null object
stoptime       286857 non-null object
bikeid         286857 non-null object
tripduration   286857 non-null float64
from_station_name 286857 non-null object
to_station_name 286857 non-null object
from_station_id 286857 non-null object
to_station_id  286857 non-null object
usertype       286857 non-null object
gender         181557 non-null object
birthyear      181553 non-null float64
dtypes: float64(2), int64(1), object(9)
memory usage: 26.3+ MB

```

The data set contains trip duration for less than three minutes, which could be possible because it may happen that the user took the bike from cycle dock and decide not to proceed with the journey and the user return bike to the dock. Another, thing which we are ignoring for the moment is the journey recorded for the short pass holder which does not kick in any meaning in our current analysis, Note, a separate analysis has performed for short pass holders but at this point this study decide to get rid of undesirable data and more focus on the data which contributing significance in our analysis.

```

In [5]: # Filter data and only consider trip duration more than 3 minute and exclude Short-Term
        bikes = df[(df.tripduration >=300) & (df.usertype != 'Short-Term Pass Holder')]

```

```

In [6]: bikes.info()

```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 143543 entries, 0 to 286847
Data columns (total 12 columns):
trip_id      143543 non-null int64
starttime    143543 non-null object
stoptime     143543 non-null object
bikeid       143543 non-null object
tripduration 143543 non-null float64
from_station_name 143543 non-null object
to_station_name 143543 non-null object
from_station_id 143543 non-null object
to_station_id 143543 non-null object
usertype     143543 non-null object
gender       143543 non-null object
birthyear    143541 non-null float64
dtypes: float64(2), int64(1), object(9)
memory usage: 14.2+ MB

```

- After applying fillter our data set shrink from 286857 to 143541 tuples

4 Imputation (Handle missing values)

What does "NaN" mean?

- "NaN" is not a string, rather it's a special value: `numpy.nan`.
- It stands for "Not a Number" and indicates a **missing value**.
- `read_csv` detects missing values (by default) when reading the file, and replaces them with this special value.

```
In [7]: # 'isnull' returns a DataFrame of booleans (True if missing, False if not missing)
        bikes.isnull().tail()
```

```
Out[7]:
```

	trip_id	starttime	stoptime	bikeid	tripduration	from_station_name	\	to_station_name	from_station_id	to_station_id	usertype	gender	birthyear
286842	False	False	False	False	False	False		False			False	False	False
286843	False	False	False	False	False	False		False			False	False	False
286844	False	False	False	False	False	False		False			False	False	False
286845	False	False	False	False	False	False		False			False	False	False
286847	False	False	False	False	False	False		False			False	False	False

```
In [8]: # 'nonnull' returns the opposite of 'isnull' (True if not missing, False if missing)
        bikes.notnull().tail()
```

```
Out[8]:
```

	trip_id	starttime	stoptime	bikeid	tripduration	from_station_name	\	to_station_name	from_station_id	to_station_id	usertype	gender	birthyear
286842	True	True	True	True	True	True		True			True	True	True
286843	True	True	True	True	True	True		True			True	True	True
286844	True	True	True	True	True	True		True			True	True	True
286845	True	True	True	True	True	True		True			True	True	True
286847	True	True	True	True	True	True		True			True	True	True

```
In [9]: # count the number of missing values in each Series
        bikes.isnull().sum()
```

```
Out[9]: trip_id          0
        starttime        0
        stoptime         0
```

```

bikeid          0
tripduration    0
from_station_name 0
to_station_name 0
from_station_id 0
to_station_id   0
usertype        0
gender          0
birthyear       2
dtype: int64

```

handle missing values depends on the dataset as well as the nature of analysis

```
In [10]: bikes.shape
```

```
Out[10]: (143543, 12)
```

```
In [11]: # if 'any' values are missing in a row, then drop that row
         bikes.dropna(how='any').shape
```

```
Out[11]: (143541, 12)
```

```
In [12]: bikes.birthyear.mean()
```

```
Out[12]: 1979.3765474672741
```

```
In [13]: # fill in missing values with a specified value
         bikes['birthyear'].fillna(value='1979', inplace=True)
```

```
In [14]: # count the number of missing values in each Series
         bikes.isnull().sum()
```

```
Out[14]: trip_id          0
         starttime        0
         stoptime         0
         bikeid           0
         tripduration     0
         from_station_name 0
         to_station_name   0
         from_station_id   0
         to_station_id     0
         usertype         0
         gender           0
         birthyear         0
         dtype: int64

```

5 Data Encoding

converting data Categorical to Numaric for further use

- Converting Gender Column

```
In [15]: # create the 'Sex_num' dummy variable using the 'map' method
        bikes['Sex_num'] = bikes.gender.map({'Female':2, 'Male':1})
```

- Converting category From Station ID and To Station ID to nominal

```
In [16]: bikes["from_station_id_cat"] = bikes["from_station_id"].astype('category')
        bikes.dtypes
```

```
Out[16]: trip_id                int64
        starttime              object
        stoptime               object
        bikeid                 object
        tripduration           float64
        from_station_name      object
        to_station_name        object
        from_station_id        object
        to_station_id          object
        usertype               object
        gender                 object
        birthyear              object
        Sex_num                float64
        from_station_id_cat    category
        dtype: object
```

```
In [17]: bikes["from_station_id_num"] = bikes["from_station_id_cat"].cat.codes
```

```
In [18]: bikes["to_station_id_cat"] = bikes["to_station_id"].astype('category')
        bikes.dtypes
```

```
Out[18]: trip_id                int64
        starttime              object
        stoptime               object
        bikeid                 object
        tripduration           float64
        from_station_name      object
        to_station_name        object
        from_station_id        object
        to_station_id          object
        usertype               object
        gender                 object
        birthyear              object
        Sex_num                float64
        from_station_id_cat    category
        from_station_id_num    int8
        to_station_id_cat      category
        dtype: object
```

```
In [19]: bikes["to_station_id_num"] = bikes["to_station_id_cat"].cat.codes
```

6 Data Encoding

For further, for our analysis we are dividing date columns in small portion. Which, could help us to understand dataset more clearly.

```
In [20]: # convert 'Time' to datetime format
        bikes['starttime'] = pd.to_datetime(bikes.starttime)
```

```
In [21]: bikes.dtypes
```

```
Out[21]: trip_id          int64
        starttime      datetime64[ns]
        stoptime        object
        bikeid          object
        tripduration    float64
        from_station_name object
        to_station_name  object
        from_station_id  object
        to_station_id    object
        usertype        object
        gender          object
        birthyear        object
        Sex_num          float64
        from_station_id_cat category
        from_station_id_num int8
        to_station_id_cat category
        to_station_id_num int8
        dtype: object
```

```
In [22]: bikes['Day'] = bikes.starttime.dt.weekday_name
```

```
In [23]: # convert 'Time' to datetime format
        bikes['stoptime'] = pd.to_datetime(bikes.stoptime)
```

```
In [24]: bikes["Day_cat"] = bikes["Day"].astype('category')
        bikes.dtypes
```

```
Out[24]: trip_id          int64
        starttime      datetime64[ns]
        stoptime      datetime64[ns]
        bikeid          object
        tripduration    float64
        from_station_name object
        to_station_name  object
        from_station_id  object
        to_station_id    object
        usertype        object
        gender          object
        birthyear        object
```

```

Sex_num                float64
from_station_id_cat    category
from_station_id_num     int8
to_station_id_cat      category
to_station_id_num      int8
Day                    object
Day_cat                category
dtype: object

```

```
In [25]: bikes["Day_num"] = bikes["Day_cat"].cat.codes
```

```
In [26]: # convenient Series attributes are now available
bikes["sthours"] = bikes.starttime.dt.hour
```

```
In [27]: # convenient Series attributes are now available
bikes["stphours"] = bikes.stoptime.dt.hour
```

```
In [28]: bikes['tripduration_minutes']=bikes['tripduration']/60
desred_decimals = 2
bikes['tripduration_minutes'] = bikes['tripduration_minutes'].apply(lambda x: round(x
```

```
In [29]: bikes['birthyear'] = bikes['birthyear'].astype(int)
```

```
In [30]: bikes['age'] = 2018 - bikes['birthyear']
desred_decimals = 2
bikes['birthyear'] = bikes['birthyear'].apply(lambda x: round(x,desred_decimals))
bikes.age.head()
```

```
Out[30]: 0    58
         1    48
         2    30
         3    41
         4    47
         Name: age, dtype: int32
```

Some statistics about age Variable

```
In [31]: bikes.age.describe()
```

```
Out[31]: count    143543.000000
         mean       38.623458
         std        10.251359
         min        19.000000
         25%        31.000000
         50%        36.000000
         75%        44.000000
         max        87.000000
         Name: age, dtype: float64
```

```
In [32]: bikes.info()
```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 143543 entries, 0 to 286847
Data columns (total 24 columns):
trip_id                143543 non-null int64
starttime              143543 non-null datetime64[ns]
stoptime               143543 non-null datetime64[ns]
bikeid                 143543 non-null object
tripduration           143543 non-null float64
from_station_name      143543 non-null object
to_station_name        143543 non-null object
from_station_id        143543 non-null object
to_station_id          143543 non-null object
usertype               143543 non-null object
gender                 143543 non-null object
birthyear              143543 non-null int64
Sex_num                140905 non-null float64
from_station_id_cat    143543 non-null category
from_station_id_num    143543 non-null int8
to_station_id_cat      143543 non-null category
to_station_id_num      143543 non-null int8
Day                    143543 non-null object
Day_cat                143543 non-null category
Day_num                143543 non-null int8
sthours                143543 non-null int64
stphours               143543 non-null int64
tripduration_minutes   143543 non-null float64
age                    143543 non-null int32
dtypes: category(3), datetime64[ns](2), float64(3), int32(1), int64(4), int8(3), object(8)
memory usage: 21.1+ MB

```

7 Removing Duplicate values

```

In [33]: # detect duplicate trip_id codes: True if an item is identical to a previous item
         bikes.trip_id.duplicated().tail()

```

```

Out[33]: 286842    False
         286843    False
         286844    False
         286845    False
         286847    False
         Name: trip_id, dtype: bool

```

The above codes shows that we do not have any duplicated rows in our dataset

```

In [34]: # Breaking date column for further Analysis
         bikes['bmonth'] = bikes['starttime'].dt.month

```



```
In [35]: # Breaking date column for further Analysis
bikes['Date'] = bikes['starttime'].dt.date
```

```
In [36]: bikes.dtypes
```

```
Out[36]: trip_id                int64
starttime          datetime64[ns]
stoptime           datetime64[ns]
bikeid             object
tripduration       float64
from_station_name  object
to_station_name    object
from_station_id    object
to_station_id      object
usertype           object
gender             object
birthyear          int64
Sex_num            float64
from_station_id_cat category
from_station_id_num int8
to_station_id_cat  category
to_station_id_num  int8
Day                object
Day_cat            category
Day_num            int8
sthours            int64
stphours           int64
tripduration_minutes float64
age                int32
bmonth             int64
Date               object
dtype: object
```

```
In [37]: bikes.Day_num.tail()
```

```
Out[37]: 286842    6
         286843    6
         286844    6
         286845    6
         286847    6
         Name: Day_num, dtype: int8
```

8 Filtering Data for year 2015 and 2016

```
In [38]: # convert 'Time' to datetime format
bikes['starttime'] = pd.to_datetime(bikes.starttime)
```

```
In [39]: # Breaking date column for further Analysis
bikes['year'] = bikes['starttime'].dt.year
```

```
In [40]: # convert 'Time' to datetime format
        bikes['Date'] = pd.to_datetime(bikes.Date)
        bikes.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 143543 entries, 0 to 286847
Data columns (total 27 columns):
trip_id                143543 non-null int64
starttime              143543 non-null datetime64[ns]
stoptime               143543 non-null datetime64[ns]
bikeid                 143543 non-null object
tripduration           143543 non-null float64
from_station_name      143543 non-null object
to_station_name        143543 non-null object
from_station_id        143543 non-null object
to_station_id          143543 non-null object
usertype               143543 non-null object
gender                 143543 non-null object
birthyear              143543 non-null int64
Sex_num                140905 non-null float64
from_station_id_cat    143543 non-null category
from_station_id_num     143543 non-null int8
to_station_id_cat      143543 non-null category
to_station_id_num      143543 non-null int8
Day                    143543 non-null object
Day_cat                143543 non-null category
Day_num                143543 non-null int8
sthours                143543 non-null int64
stphours               143543 non-null int64
tripduration_minutes   143543 non-null float64
age                    143543 non-null int32
bmonth                 143543 non-null int64
Date                   143543 non-null datetime64[ns]
year                   143543 non-null int64
dtypes: category(3), datetime64[ns](3), float64(3), int32(1), int64(6), int8(3), object(8)
memory usage: 24.4+ MB
```

```
In [41]: # df2 belong to bikes dataset
        df2 = bikes[(bikes.year != 2014) ]
        df2.head(100)
```

```
Out[41]:
```

	trip_id	starttime	stoptime	bikeid \
20240	25092	2015-01-01 00:37:00	2015-01-01 00:44:00	SEA00267
20252	25114	2015-01-01 03:07:00	2015-01-01 03:18:00	SEA00472
20255	25127	2015-01-01 08:12:00	2015-01-01 08:20:00	SEA00245
20260	25132	2015-01-01 10:55:00	2015-01-01 11:03:00	SEA00391
20263	25135	2015-01-01 11:31:00	2015-01-01 11:55:00	SEA00079

20265	25137	2015-01-01	11:45:00	2015-01-01	11:54:00	SEA00401
20270	25143	2015-01-01	12:13:00	2015-01-01	12:22:00	SEA00347
20271	25144	2015-01-01	12:14:00	2015-01-01	12:22:00	SEA00326
20274	25147	2015-01-01	12:16:00	2015-01-01	12:27:00	SEA00140
20276	25150	2015-01-01	12:52:00	2015-01-01	13:16:00	SEA00212
20277	25151	2015-01-01	12:52:00	2015-01-01	12:59:00	SEA00390
20281	25155	2015-01-01	12:58:00	2015-01-01	13:16:00	SEA00210
20282	25156	2015-01-01	12:59:00	2015-01-01	13:16:00	SEA00141
20292	25168	2015-01-01	13:17:00	2015-01-01	13:32:00	SEA00255
20325	25203	2015-01-01	14:22:00	2015-01-01	14:31:00	SEA00390
20327	25205	2015-01-01	14:27:00	2015-01-01	14:46:00	SEA00212
20330	25210	2015-01-01	14:49:00	2015-01-01	14:57:00	SEA00319
20331	25211	2015-01-01	15:00:00	2015-01-01	15:15:00	SEA00264
20332	25212	2015-01-01	15:02:00	2015-01-01	15:12:00	SEA00206
20333	25213	2015-01-01	15:15:00	2015-01-01	15:25:00	SEA00418
20337	25217	2015-01-01	15:19:00	2015-01-01	15:27:00	SEA00056
20342	25222	2015-01-01	15:27:00	2015-01-01	15:36:00	SEA00312
20346	25226	2015-01-01	15:40:00	2015-01-01	15:57:00	SEA00215
20347	25228	2015-01-01	15:46:00	2015-01-01	15:54:00	SEA00056
20348	25229	2015-01-01	15:45:00	2015-01-01	15:58:00	SEA00382
20350	25231	2015-01-01	16:04:00	2015-01-01	16:12:00	SEA00212
20351	25232	2015-01-01	16:04:00	2015-01-01	16:13:00	SEA00135
20356	25237	2015-01-01	16:17:00	2015-01-01	16:33:00	SEA00421
20357	25238	2015-01-01	16:19:00	2015-01-01	16:27:00	SEA00278
20358	25239	2015-01-01	16:23:00	2015-01-01	16:28:00	SEA00116
...
20437	25324	2015-01-02	10:42:00	2015-01-02	11:04:00	SEA00299
20438	25325	2015-01-02	11:21:00	2015-01-02	11:34:00	SEA00100
20439	25326	2015-01-02	11:25:00	2015-01-02	11:32:00	SEA00309
20440	25327	2015-01-02	11:38:00	2015-01-02	11:55:00	SEA00168
20443	25330	2015-01-02	11:52:00	2015-01-02	11:58:00	SEA00208
20446	25333	2015-01-02	12:13:00	2015-01-02	12:20:00	SEA00405
20447	25334	2015-01-02	12:21:00	2015-01-02	12:32:00	SEA00277
20448	25335	2015-01-02	12:26:00	2015-01-02	12:49:00	SEA00153
20449	25336	2015-01-02	12:26:00	2015-01-02	12:43:00	SEA00496
20450	25337	2015-01-02	12:39:00	2015-01-02	12:44:00	SEA00340
20451	25338	2015-01-02	12:43:00	2015-01-02	12:49:00	SEA00198
20453	25340	2015-01-02	12:50:00	2015-01-02	13:01:00	SEA00196
20454	25341	2015-01-02	12:52:00	2015-01-02	13:09:00	SEA00496
20455	25342	2015-01-02	12:52:00	2015-01-02	12:59:00	SEA00056
20457	25344	2015-01-02	12:56:00	2015-01-02	13:03:00	SEA00309
20458	25345	2015-01-02	13:02:00	2015-01-02	13:15:00	SEA00100
20459	25346	2015-01-02	13:08:00	2015-01-02	13:28:00	SEA00447
20461	25349	2015-01-02	13:16:00	2015-01-02	13:26:00	SEA00135
20464	25353	2015-01-02	13:26:00	2015-01-02	13:31:00	SEA00483
20465	25354	2015-01-02	13:38:00	2015-01-02	13:55:00	SEA00063
20467	25356	2015-01-02	13:56:00	2015-01-02	14:03:00	SEA00232
20468	25357	2015-01-02	14:03:00	2015-01-02	14:09:00	SEA00356

20469	25358	2015-01-02	14:10:00	2015-01-02	14:16:00	SEA00278
20470	25359	2015-01-02	14:12:00	2015-01-02	14:33:00	SEA00477
20474	25363	2015-01-02	14:32:00	2015-01-02	14:49:00	SEA00168
20476	25365	2015-01-02	14:48:00	2015-01-02	14:57:00	SEA00235
20477	25366	2015-01-02	14:53:00	2015-01-02	15:07:00	SEA00477
20479	25371	2015-01-02	15:39:00	2015-01-02	15:52:00	SEA00383
20481	25373	2015-01-02	15:51:00	2015-01-02	15:56:00	SEA00208
20489	25381	2015-01-02	16:31:00	2015-01-02	16:43:00	SEA00365

	tripduration		from_station_name	\
20240	459.469		Harvard Ave & E Pine St	
20252	614.453		9th Ave N & Mercer St	
20255	504.420		E Pine St & 16th Ave	
20260	470.801		E Pine St & 16th Ave	
20263	1461.638		12th Ave & E Denny Way	
20265	524.885		Harvard Ave & E Pine St	
20270	514.110		E Harrison St & Broadway Ave E	
20271	472.876		E Harrison St & Broadway Ave E	
20274	684.994		2nd Ave & Vine St	
20276	1463.476		E Pine St & 16th Ave	
20277	397.101		Cal Anderson Park / 11th Ave & Pine St	
20281	1101.863		E Blaine St & Fairview Ave E	
20282	1057.551		E Blaine St & Fairview Ave E	
20292	929.421		Lake Union Park / Valley St & Boren Ave N	
20325	515.097		E Harrison St & Broadway Ave E	
20327	1149.250		6th Ave S & S King St	
20330	485.702		E Blaine St & Fairview Ave E	
20331	910.564		E Harrison St & Broadway Ave E	
20332	628.489		Bellevue Ave & E Pine St	
20333	549.486		Summit Ave E & E Republican St	
20337	465.429		3rd Ave & Broad St	
20342	495.314		E Harrison St & Broadway Ave E	
20346	983.226		15th Ave E & E Thomas St	
20347	474.824		2nd Ave & Pine St	
20348	737.699		15th Ave E & E Thomas St	
20350	480.341		E Pine St & 16th Ave	
20351	491.257		E Pine St & 16th Ave	
20356	939.485		City Hall / 4th Ave & James St	
20357	451.038		12th Ave & E Denny Way	
20358	311.966		PATH / 9th Ave & Westlake Ave	
...	
20437	1265.463	Occidental Park / Occidental Ave S & S Washing...		
20438	835.595		Harvard Ave & E Pine St	
20439	406.067	Fred Hutchinson Cancer Research Center / Fairv...		
20440	993.430		Eastlake Ave E & E Allison St	
20443	330.189		12th Ave & NE Campus Pkwy	
20446	416.075		2nd Ave & Vine St	
20447	652.883	King Street Station Plaza / 2nd Ave Extension ...		

20448	1372.678	Burke-Gilman Trail / NE Blakeley St & 24th Ave NE
20449	1012.147	Key Arena / 1st Ave N & Harrison St
20450	342.515	2nd Ave & Pine St
20451	373.090	3rd Ave & Broad St
20453	639.008	King Street Station Plaza / 2nd Ave Extension ...
20454	1048.554	7th Ave & Union St
20455	403.213	3rd Ave & Broad St
20457	376.005	PATH / 9th Ave & Westlake Ave
20458	795.976	15th Ave E & E Thomas St
20459	1150.649	E Blaine St & Fairview Ave E
20461	649.109	E Harrison St & Broadway Ave E
20464	350.746	2nd Ave & Vine St
20465	993.728	E Harrison St & Broadway Ave E
20467	425.978	2nd Ave & Spring St
20468	330.987	Westlake Ave & 6th Ave
20469	365.864	PATH / 9th Ave & Westlake Ave
20470	1269.127	Occidental Park / Occidental Ave S & S Washing...
20474	1023.856	Lake Union Park / Valley St & Boren Ave N
20476	563.844	UW Engineering Library / E Stevens Way NE & Je...
20477	836.574	Cal Anderson Park / 11th Ave & Pine St
20479	785.410	2nd Ave & Pine St
20481	318.126	UW Engineering Library / E Stevens Way NE & Je...
20489	752.358	2nd Ave & Blanchard St

	to_station_name	from_station_id	\
20240	Cal Anderson Park / 11th Ave & Pine St	CH-09	
20252	E Blaine St & Fairview Ave E	DPD-01	
20255	2nd Ave & Pine St	CH-07	
20260	7th Ave & Union St	CH-07	
20263	Key Arena / 1st Ave N & Harrison St	CH-06	
20265	Summit Ave E & E Republican St	CH-09	
20270	Seattle University / E Columbia St & 12th Ave	CH-02	
20271	Seattle University / E Columbia St & 12th Ave	CH-02	
20274	Republican St & Westlake Ave N	BT-03	
20276	6th Ave S & S King St	CH-07	
20277	E Harrison St & Broadway Ave E	CH-08	
20281	Westlake Ave & 6th Ave	EL-03	
20282	Westlake Ave & 6th Ave	EL-03	
20292	Key Arena / 1st Ave N & Harrison St	SLU-17	
20325	Seattle University / E Columbia St & 12th Ave	CH-02	
20327	E Pine St & 16th Ave	ID-04	
20330	Lake Union Park / Valley St & Boren Ave N	EL-03	
20331	12th Ave & NE Campus Pkwy	CH-02	
20332	15th Ave E & E Thomas St	CH-12	
20333	15th Ave E & E Thomas St	CH-03	
20337	2nd Ave & Pine St	BT-01	
20342	Cal Anderson Park / 11th Ave & Pine St	CH-02	
20346	Eastlake Ave E & E Allison St	CH-05	

20347	3rd Ave & Broad St	CBD-13
20348	REI / Yale Ave N & John St	CH-05
20350	E Harrison St & Broadway Ave E	CH-07
20351	E Harrison St & Broadway Ave E	CH-07
20356	E Harrison St & Broadway Ave E	CBD-07
20357	PATH / 9th Ave & Westlake Ave	CH-06
20358	REI / Yale Ave N & John St	SLU-07
...
20437	Cal Anderson Park / 11th Ave & Pine St	PS-04
20438	15th Ave E & E Thomas St	CH-09
20439	PATH / 9th Ave & Westlake Ave	EL-01
20440	Lake Union Park / Valley St & Boren Ave N	EL-05
20443	UW Engineering Library / E Stevens Way NE & Je...	UD-04
20446	Republican St & Westlake Ave N	BT-03
20447	Seattle Aquarium / Alaskan Way S & Elliott Bay...	PS-05
20448	15th Ave NE & NE 40th St	UD-01
20449	7th Ave & Union St	SLU-19
20450	1st Ave & Marion St	CBD-13
20451	PATH / 9th Ave & Westlake Ave	BT-01
20453	2nd Ave & Spring St	PS-05
20454	Key Arena / 1st Ave N & Harrison St	CBD-03
20455	2nd Ave & Spring St	BT-01
20457	Fred Hutchinson Cancer Research Center / Fairv...	SLU-07
20458	Bellevue Ave & E Pine St	CH-05
20459	2nd Ave & Spring St	EL-03
20461	E Pine St & 16th Ave	CH-02
20464	Republican St & Westlake Ave N	BT-03
20465	2nd Ave & Pine St	CH-02
20467	King Street Station Plaza / 2nd Ave Extension ...	CBD-06
20468	3rd Ave & Broad St	SLU-15
20469	7th Ave & Union St	SLU-07
20470	Cal Anderson Park / 11th Ave & Pine St	PS-04
20474	Eastlake Ave E & E Allison St	SLU-17
20476	12th Ave & NE Campus Pkwy	UW-06
20477	1st Ave & Marion St	CH-08
20479	E Blaine St & Fairview Ave E	CBD-13
20481	NE 42nd St & University Way NE	UW-06
20489	Summit Ave & E Denny Way	BT-05

	to_station_id	usertype	...	Day	Day_cat	Day_num	sthours	\
20240	CH-08	Member	...	Thursday	Thursday	4	0	
20252	EL-03	Member	...	Thursday	Thursday	4	3	
20255	CBD-13	Member	...	Thursday	Thursday	4	8	
20260	CBD-03	Member	...	Thursday	Thursday	4	10	
20263	SLU-19	Member	...	Thursday	Thursday	4	11	
20265	CH-03	Member	...	Thursday	Thursday	4	11	
20270	FH-04	Member	...	Thursday	Thursday	4	12	
20271	FH-04	Member	...	Thursday	Thursday	4	12	

20274	SLU-04	Member	...	Thursday	Thursday	4	12
20276	ID-04	Member	...	Thursday	Thursday	4	12
20277	CH-02	Member	...	Thursday	Thursday	4	12
20281	SLU-15	Member	...	Thursday	Thursday	4	12
20282	SLU-15	Member	...	Thursday	Thursday	4	12
20292	SLU-19	Member	...	Thursday	Thursday	4	13
20325	FH-04	Member	...	Thursday	Thursday	4	14
20327	CH-07	Member	...	Thursday	Thursday	4	14
20330	SLU-17	Member	...	Thursday	Thursday	4	14
20331	UD-04	Member	...	Thursday	Thursday	4	15
20332	CH-05	Member	...	Thursday	Thursday	4	15
20333	CH-05	Member	...	Thursday	Thursday	4	15
20337	CBD-13	Member	...	Thursday	Thursday	4	15
20342	CH-08	Member	...	Thursday	Thursday	4	15
20346	EL-05	Member	...	Thursday	Thursday	4	15
20347	BT-01	Member	...	Thursday	Thursday	4	15
20348	SLU-01	Member	...	Thursday	Thursday	4	15
20350	CH-02	Member	...	Thursday	Thursday	4	16
20351	CH-02	Member	...	Thursday	Thursday	4	16
20356	CH-02	Member	...	Thursday	Thursday	4	16
20357	SLU-07	Member	...	Thursday	Thursday	4	16
20358	SLU-01	Member	...	Thursday	Thursday	4	16
...
20437	CH-08	Member	...	Friday	Friday	0	10
20438	CH-05	Member	...	Friday	Friday	0	11
20439	SLU-07	Member	...	Friday	Friday	0	11
20440	SLU-17	Member	...	Friday	Friday	0	11
20443	UW-06	Member	...	Friday	Friday	0	11
20446	SLU-04	Member	...	Friday	Friday	0	12
20447	WF-04	Member	...	Friday	Friday	0	12
20448	UW-04	Member	...	Friday	Friday	0	12
20449	CBD-03	Member	...	Friday	Friday	0	12
20450	CBD-05	Member	...	Friday	Friday	0	12
20451	SLU-07	Member	...	Friday	Friday	0	12
20453	CBD-06	Member	...	Friday	Friday	0	12
20454	SLU-19	Member	...	Friday	Friday	0	12
20455	CBD-06	Member	...	Friday	Friday	0	12
20457	EL-01	Member	...	Friday	Friday	0	12
20458	CH-12	Member	...	Friday	Friday	0	13
20459	CBD-06	Member	...	Friday	Friday	0	13
20461	CH-07	Member	...	Friday	Friday	0	13
20464	SLU-04	Member	...	Friday	Friday	0	13
20465	CBD-13	Member	...	Friday	Friday	0	13
20467	PS-05	Member	...	Friday	Friday	0	13
20468	BT-01	Member	...	Friday	Friday	0	14
20469	CBD-03	Member	...	Friday	Friday	0	14
20470	CH-08	Member	...	Friday	Friday	0	14
20474	EL-05	Member	...	Friday	Friday	0	14

20476	UD-04	Member	...	Friday	Friday	0	14
20477	CBD-05	Member	...	Friday	Friday	0	14
20479	EL-03	Member	...	Friday	Friday	0	15
20481	UD-02	Member	...	Friday	Friday	0	15
20489	CH-01	Member	...	Friday	Friday	0	16

	stphours	tripduration_minutes	age	bmonth	Date	year
20240	0	7.66	27	1	2015-01-01	2015
20252	3	10.24	38	1	2015-01-01	2015
20255	8	8.41	32	1	2015-01-01	2015
20260	11	7.85	38	1	2015-01-01	2015
20263	11	24.36	31	1	2015-01-01	2015
20265	11	8.75	31	1	2015-01-01	2015
20270	12	8.57	32	1	2015-01-01	2015
20271	12	7.88	37	1	2015-01-01	2015
20274	12	11.42	37	1	2015-01-01	2015
20276	13	24.39	33	1	2015-01-01	2015
20277	12	6.62	43	1	2015-01-01	2015
20281	13	18.36	41	1	2015-01-01	2015
20282	13	17.63	44	1	2015-01-01	2015
20292	13	15.49	63	1	2015-01-01	2015
20325	14	8.58	37	1	2015-01-01	2015
20327	14	19.15	33	1	2015-01-01	2015
20330	14	8.10	35	1	2015-01-01	2015
20331	15	15.18	50	1	2015-01-01	2015
20332	15	10.47	32	1	2015-01-01	2015
20333	15	9.16	41	1	2015-01-01	2015
20337	15	7.76	49	1	2015-01-01	2015
20342	15	8.26	43	1	2015-01-01	2015
20346	15	16.39	34	1	2015-01-01	2015
20347	15	7.91	49	1	2015-01-01	2015
20348	15	12.29	30	1	2015-01-01	2015
20350	16	8.01	37	1	2015-01-01	2015
20351	16	8.19	32	1	2015-01-01	2015
20356	16	15.66	31	1	2015-01-01	2015
20357	16	7.52	27	1	2015-01-01	2015
20358	16	5.20	43	1	2015-01-01	2015
...
20437	11	21.09	32	1	2015-01-02	2015
20438	11	13.93	71	1	2015-01-02	2015
20439	11	6.77	55	1	2015-01-02	2015
20440	11	16.56	34	1	2015-01-02	2015
20443	11	5.50	28	1	2015-01-02	2015
20446	12	6.93	37	1	2015-01-02	2015
20447	12	10.88	35	1	2015-01-02	2015
20448	12	22.88	41	1	2015-01-02	2015
20449	12	16.87	47	1	2015-01-02	2015
20450	12	5.71	69	1	2015-01-02	2015

20451	12	6.22	34	1	2015-01-02	2015
20453	13	10.65	35	1	2015-01-02	2015
20454	13	17.48	47	1	2015-01-02	2015
20455	12	6.72	34	1	2015-01-02	2015
20457	13	6.27	55	1	2015-01-02	2015
20458	13	13.27	71	1	2015-01-02	2015
20459	13	19.18	41	1	2015-01-02	2015
20461	13	10.82	50	1	2015-01-02	2015
20464	13	5.85	29	1	2015-01-02	2015
20465	13	16.56	27	1	2015-01-02	2015
20467	14	7.10	35	1	2015-01-02	2015
20468	14	5.52	34	1	2015-01-02	2015
20469	14	6.10	45	1	2015-01-02	2015
20470	14	21.15	69	1	2015-01-02	2015
20474	14	17.06	34	1	2015-01-02	2015
20476	14	9.40	31	1	2015-01-02	2015
20477	15	13.94	69	1	2015-01-02	2015
20479	15	13.09	41	1	2015-01-02	2015
20481	15	5.30	28	1	2015-01-02	2015
20489	16	12.54	36	1	2015-01-02	2015

[100 rows x 27 columns]

9 Saving Bike data

```
In [42]: df2.to_csv('C:\\Users\\mrferozi\\Desktop\\GitHub\\Bike\\dataset\\cycle\\trip_clean.csv')
```

10 Cleaning Weather Data

```
In [5]: # read in a CSV
df = pd.read_csv('C:/Users/mrferozi/Desktop/GitHub/Bike/cycle-share-dataset-original/weather.csv')
```

```
In [6]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 689 entries, 0 to 688
Data columns (total 21 columns):
Date                689 non-null object
Max_Temperature_F   689 non-null int64
Mean_Temperature_F  688 non-null float64
Min_TemperatureF    689 non-null int64
Max_Dew_Point_F     689 non-null int64
MeanDew_Point_F     689 non-null int64
Min_Dewpoint_F      689 non-null int64
Max_Humidity        689 non-null int64
```

```

Mean_Humidity          689 non-null int64
Min_Humidity           689 non-null int64
Max_Sea_Level_Pressure_In 689 non-null float64
Mean_Sea_Level_Pressure_In 689 non-null float64
Min_Sea_Level_Pressure_In 689 non-null float64
Max_Visibility_Miles    689 non-null int64
Mean_Visibility_Miles   689 non-null int64
Min_Visibility_Miles    689 non-null int64
Max_Wind_Speed_MPH      689 non-null int64
Mean_Wind_Speed_MPH     689 non-null int64
Max_Gust_Speed_MPH      504 non-null object
Precipitation_In        689 non-null float64
Events                 328 non-null object
dtypes: float64(5), int64(13), object(3)
memory usage: 113.1+ KB

```

11 Converting temprature from Fahrenheit to Celsius

The definition of precipitation is any form of water - liquid or solid - falling from the sky. It includes rain, sleet, snow, hail and drizzle plus a few less common occurrences such as ice pellets, diamond dust and freezing rain. <https://www.metoffice.gov.uk/learning/learn-about-the-weather/weather-phenomena/what-is-precipitation>

- Converting temprature from Fahrenheit to Celsius

$$T(^{\circ}\text{C}) = (T(^{\circ}\text{F}) - 32) \times 5/9$$

Example

Convert 68 degrees Fahrenheit to degrees Celsius:

$T(^{\circ}\text{C}) = (68^{\circ}\text{F} - 32) \times 5/9 = 20^{\circ}\text{C}$ <https://www.rapidtables.com/convert/temperature/how-fahrenheit-to-celsius.html>

```
In [7]: a = df.Mean_Temperature_F-32
```

```
In [8]: df['Mean_Temperature_C'] = a * 5/9
```

```
desred_decimals = 2
```

```
df['Mean_Temperature_C'] = df['Mean_Temperature_C'].apply(lambda x: round(x,desred_decimals))
```

```
In [9]: # 'isnull' returns a DataFrame of booleans (True if missing, False if not missing)
df.isnull().tail()
```

```

Out[9]:      Date Max_Temperature_F Mean_Temperature_F Min_Temperature_F \
684  False                False                False                False
685  False                False                False                False
686  False                False                False                False
687  False                False                False                False
688  False                False                False                False

```

	Max_Dew_Point_F	MeanDew_Point_F	Min_Dewpoint_F	Max_Humidity	Mean_Humidity	\
684	False	False	False	False	False	
685	False	False	False	False	False	
686	False	False	False	False	False	
687	False	False	False	False	False	
688	False	False	False	False	False	

	Min_Humidity	...	Min_Sea_Level_Pressure_In	\
684	False	...	False	
685	False	...	False	
686	False	...	False	
687	False	...	False	
688	False	...	False	

	Max_Visibility_Miles	Mean_Visibility_Miles	Min_Visibility_Miles	\
684	False		False	False
685	False		False	False
686	False		False	False
687	False		False	False
688	False		False	False

	Max_Wind_Speed_MPH	Mean_Wind_Speed_MPH	Max_Gust_Speed_MPH	\
684	False	False	True	
685	False	False	True	
686	False	False	True	
687	False	False	False	
688	False	False	True	

	Precipitation_In	Events	Mean_Temperature_C
684	False	True	False
685	False	True	False
686	False	True	False
687	False	True	False
688	False	False	False

[5 rows x 22 columns]

```
In [10]: # count the number of missing values in each Series
df.isnull().sum()
```

```
Out[10]: Date                                0
Max_Temperature_F                           0
Mean_Temperature_F                           1
Min_TemperatureF                             0
Max_Dew_Point_F                             0
MeanDew_Point_F                             0
Min_Dewpoint_F                               0
Max_Humidity                                0
```

Mean_Humidity	0
Min_Humidity	0
Max_Sea_Level_Pressure_In	0
Mean_Sea_Level_Pressure_In	0
Min_Sea_Level_Pressure_In	0
Max_Visibility_Miles	0
Mean_Visibility_Miles	0
Min_Visibility_Miles	0
Max_Wind_Speed_MPH	0
Mean_Wind_Speed_MPH	0
Max_Gust_Speed_MPH	185
Precipitation_In	0
Events	361
Mean_Temperature_C	1

dtype: int64

12 Filling Max_Gust_Speed_MPH with Some Values

```
In [11]: #Checking column for exsisting values
df['Max_Gust_Speed_MPH']
```

```
Out[11]: 0      21
1      17
2      25
3      -
4      -
5      -
6      18
7      -
8      21
9      22
10     22
11     -
12     41
13     30
14     -
15     -
16     24
17     -
18     -
19     -
20     -
21     20
22     -
23     -
24     37
25     -
```

```

26      -
27     16
28      -
29     22
...
659     17
660    NaN
661     21
662    NaN
663    NaN
664    NaN
665    NaN
666    NaN
667     17
668     20
669    NaN
670    NaN
671    NaN
672    NaN
673    NaN
674    NaN
675    NaN
676    NaN
677    NaN
678    NaN
679    NaN
680    NaN
681    NaN
682    NaN
683    NaN
684    NaN
685    NaN
686    NaN
687     18
688    NaN
Name: Max_Gust_Speed_MPH, dtype: object

```

```

In [12]: # fill in missing values with a specified value
df['Max_Gust_Speed_MPH'].fillna(value='0', inplace=True)
df.Max_Gust_Speed_MPH.replace('-', '0', inplace=True)
df['Max_Gust_Speed_MPH']

```

```

Out[12]: 0      21
1      17
2      25
3       0
4       0
5       0

```

6	18
7	0
8	21
9	22
10	22
11	0
12	41
13	30
14	0
15	0
16	24
17	0
18	0
19	0
20	0
21	20
22	0
23	0
24	37
25	0
26	0
27	16
28	0
29	22
	..
659	17
660	0
661	21
662	0
663	0
664	0
665	0
666	0
667	17
668	20
669	0
670	0
671	0
672	0
673	0
674	0
675	0
676	0
677	0
678	0
679	0
680	0
681	0

```

682      0
683      0
684      0
685      0
686      0
687     18
688      0
Name: Max_Gust_Speed_MPH, dtype: object

```

```

In [13]: ##Filling Null values with the mean of Max_Gust_Speed_MPH' values
df['Max_Gust_Speed_MPH'] = df['Max_Gust_Speed_MPH'].astype(int)
df['Max_Gust_Speed_MPH'].mean()

```

```

Out[13]: 9.252539912917271

```

```

In [14]: df.Max_Gust_Speed_MPH.replace('0', '9.26', inplace=True)

```

13 Filling Events with Some Values

```

In [15]: # count the number of missing values in each Series
df.isnull().sum()

```

```

Out[15]: Date                                0
Max_Temperature_F                           0
Mean_Temperature_F                           1
Min_TemperatureF                             0
Max_Dew_Point_F                             0
MeanDew_Point_F                             0
Min_Dewpoint_F                              0
Max_Humidity                                0
Mean_Humidity                               0
Min_Humidity                                0
Max_Sea_Level_Pressure_In                   0
Mean_Sea_Level_Pressure_In                  0
Min_Sea_Level_Pressure_In                   0
Max_Visibility_Miles                        0
Mean_Visibility_Miles                       0
Min_Visibility_Miles                        0
Max_Wind_Speed_MPH                          0
Mean_Wind_Speed_MPH                         0
Max_Gust_Speed_MPH                          0
Precipitation_In                            0
Events                                     361
Mean_Temperature_C                           1
dtype: int64

```

```

In [16]: import pandas as pd
Inplace = True

```

```

for index, row in df.iterrows():
    a = row.Precipitation_In
    if a == 0:
        df.at[index, 'Events'] = 'Rain'

```

```

In [17]: # count the number of missing values in each Series
df.isnull().sum()

```

```

Out[17]: Date                                0
Max_Temperature_F                           0
Mean_Temperature_F                           1
Min_TemperatureF                             0
Max_Dew_Point_F                             0
MeanDew_Point_F                             0
Min_Dewpoint_F                              0
Max_Humidity                                0
Mean_Humidity                               0
Min_Humidity                                0
Max_Sea_Level_Pressure_In                   0
Mean_Sea_Level_Pressure_In                  0
Min_Sea_Level_Pressure_In                   0
Max_Visibility_Miles                        0
Mean_Visibility_Miles                       0
Min_Visibility_Miles                        0
Max_Wind_Speed_MPH                          0
Mean_Wind_Speed_MPH                         0
Max_Gust_Speed_MPH                          0
Precipitation_In                            0
Events                                       14
Mean_Temperature_C                           1
dtype: int64

```

```

In [18]: import pandas as pd
Inplace = True
#df['Mean_Temperature_C'] = df['Mean_Temperature_C'].astype(int)
for index, row in df.iterrows():
    a = row.Mean_Temperature_C
    if a >= 14:
        df.at[index, 'Events'] = 'Sunny'
    print (a)

```

```

16.67
15.0
14.44
16.11
15.56
17.78
17.78
15.56

```


14.44
14.44
12.78
13.33
14.44
11.11
12.22
13.33
15.0
14.44
12.22
11.11
10.56
14.44
14.44
13.89
15.56
11.11
8.89
11.11
8.89
5.56
3.89
4.44
3.33
3.33
3.89
5.0
3.89
6.67
9.44
11.11
8.89
10.0
8.33
13.33
15.0
14.44
8.89
1.67
0.56
1.11
1.67
4.44
6.67
10.0
11.11
9.44

12.78
13.89
16.11
12.78
11.11
7.78
8.33
8.89
8.89
8.89
8.89
10.0
11.11
12.78
8.89
8.89
6.67
5.56
5.56
8.33
6.67
5.56
1.11
1.11
1.67
3.33
3.33
8.89
13.33
10.0
8.33
6.67
6.67
7.78
8.89
9.44
6.67
5.56
5.0
10.0
7.78
12.22
9.44
7.78
4.44
8.89
12.22
14.44

13.89
12.78
11.11
10.0
8.89
6.67
5.56
8.89
10.0
8.89
8.33
12.22
13.33
12.22
12.22
11.11
11.11
10.56
16.67
12.22
12.22
8.89
11.67
11.11
8.89
10.56
11.11
10.0
7.22
6.67
7.22
9.44
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```
In [1]: #Search google for Sunny condition in Seattle
        from IPython.display import HTML
        HTML('<iframe src=https://www.accuweather.com/en/us/seattle-wa/98104/current-weather/3
```

Out[1]: <IPython.core.display.HTML object>

```
In [20]: # count the number of missing values in each Series
         df.isnull().sum()
```

```
Out[20]: Date                                0
         Max_Temperature_F                    0
         Mean_Temperature_F                    1
         Min_TemperatureF                      0
         Max_Dew_Point_F                      0
         MeanDew_Point_F                      0
         Min_Dewpoint_F                       0
         Max_Humidity                         0
         Mean_Humidity                        0
         Min_Humidity                         0
         Max_Sea_Level_Pressure_In            0
         Mean_Sea_Level_Pressure_In           0
         Min_Sea_Level_Pressure_In            0
         Max_Visibility_Miles                  0
         Mean_Visibility_Miles                 0
         Min_Visibility_Miles                  0
         Max_Wind_Speed_MPH                    0
         Mean_Wind_Speed_MPH                   0
         Max_Gust_Speed_MPH                    0
         Precipitation_In                      0
         Events                                7
         Mean_Temperature_C                     1
         dtype: int64
```

```
In [21]: # confirm that the missing values were filled in
         df['Events'].value_counts().head()
```

```
Out[21]: Rain                339
         Sunny                325
```

```

Fog , Rain          5
Fog-Rain            5
Rain-Thunderstorm   3
Name: Events, dtype: int64

```

14 Encoding Columns values

```

In [22]: df.Events.replace('Fog-Rain', 'Fog, Rain',inplace=True)
         df.Events.replace('Fog , Rain', 'Fog, Rain',inplace=True)

```

```

In [23]: df['Events'].value_counts().head()

```

```

Out[23]: Rain          339
         Sunny         325
         Fog, Rain      10
         Rain-Thunderstorm  3
         Fog            2
         Name: Events, dtype: int64

```

15 Converting Categorical variable to nominal

```

In [24]: df["Events_cat"] = df["Events"].astype('category')
         df.dtypes

```

```

Out[24]: Date          object
         Max_Temperature_F      int64
         Mean_Temperature_F     float64
         Min_TemperatureF       int64
         Max_Dew_Point_F       int64
         MeanDew_Point_F       int64
         Min_Dewpoint_F       int64
         Max_Humidity          int64
         Mean_Humidity         int64
         Min_Humidity          int64
         Max_Sea_Level_Pressure_In  float64
         Mean_Sea_Level_Pressure_In  float64
         Min_Sea_Level_Pressure_In  float64
         Max_Visibility_Miles      int64
         Mean_Visibility_Miles     int64
         Min_Visibility_Miles     int64
         Max_Wind_Speed_MPH       int64
         Mean_Wind_Speed_MPH      int64
         Max_Gust_Speed_MPH       object
         Precipitation_In        float64
         Events                 object
         Mean_Temperature_C      float64
         Events_cat              category
         dtype: object

```



```

In [25]: df["Events_num"] = df["Events_cat"].cat.codes

In [29]: df["Events_cat"] = df["Events"].astype('category')
         df["Events_num"] = df["Events_cat"].cat.codes

In [36]: df['Events_num'].mode()

Out[36]: 0      2
         dtype: int8

In [37]: df[(df.Events_num == 2)]

Out[37]:
```

	Date	Max_Temperature_F	Mean_Temperature_F	Min_TemperatureF	\
10	10/23/2014	62	55.0	50	
11	10/24/2014	60	56.0	51	
13	10/26/2014	59	52.0	48	
14	10/27/2014	62	54.0	45	
15	10/28/2014	62	56.0	51	
18	10/31/2014	59	54.0	52	
19	11/01/2014	55	52.0	48	
20	11/02/2014	57	51.0	45	
23	11/05/2014	61	57.0	53	
25	11/07/2014	59	52.0	45	
26	11/08/2014	55	48.0	42	
27	11/09/2014	57	52.0	48	
28	11/10/2014	54	48.0	43	
29	11/11/2014	48	42.0	36	
30	11/12/2014	46	39.0	32	
31	11/13/2014	48	40.0	30	
32	11/14/2014	48	38.0	28	
33	11/15/2014	48	38.0	27	
34	11/16/2014	50	39.0	28	
35	11/17/2014	52	41.0	30	
36	11/18/2014	48	39.0	30	
37	11/19/2014	52	44.0	36	
38	11/20/2014	54	49.0	44	
39	11/21/2014	55	52.0	50	
40	11/22/2014	51	48.0	45	
41	11/23/2014	57	50.0	44	
42	11/24/2014	55	47.0	39	
43	11/25/2014	60	56.0	51	
46	11/28/2014	57	48.0	39	
48	11/30/2014	39	33.0	27	
...	
533	3/29/2016	64	50.0	37	
534	3/30/2016	68	54.0	43	
537	04/02/2016	63	54.0	48	
539	04/04/2016	60	54.0	46	
540	04/05/2016	57	52.0	46	

541	04/06/2016	68	56.0	45
544	04/09/2016	66	55.0	46
545	04/10/2016	59	52.0	48
546	04/11/2016	57	54.0	50
547	04/12/2016	60	52.0	46
548	4/13/2016	59	52.0	46
549	4/14/2016	59	52.0	45
550	4/15/2016	60	54.0	48
551	4/16/2016	68	56.0	43
559	4/24/2016	55	51.0	48
561	4/26/2016	60	52.0	43
562	4/27/2016	61	56.0	51
564	4/29/2016	61	55.0	50
565	4/30/2016	68	56.0	45
569	05/04/2016	60	56.0	54
573	05/08/2016	63	56.0	50
574	05/09/2016	66	57.0	48
580	5/15/2016	57	56.0	54
581	5/16/2016	60	56.0	52
584	5/19/2016	63	55.0	48
586	5/21/2016	59	56.0	52
592	5/27/2016	64	57.0	51
593	5/28/2016	64	56.0	50
610	6/14/2016	60	54.0	48
611	6/15/2016	66	56.0	45

	Max_Dew_Point_F	MeanDew_Point_F	Min_Dewpoint_F	Max_Humidity	\
10	49	47	44	86	
11	50	47	44	86	
13	48	44	42	86	
14	46	43	41	87	
15	54	50	45	88	
18	52	49	46	89	
19	48	45	42	89	
20	52	47	42	90	
23	55	51	48	90	
25	45	43	40	86	
26	44	42	39	93	
27	50	46	43	87	
28	43	40	34	86	
29	39	18	3	80	
30	10	4	1	40	
31	19	9	2	64	
32	26	18	11	69	
33	25	18	10	75	
34	27	21	15	69	
35	29	25	22	75	
36	36	31	25	86	

37	37	34	30	86
38	46	43	39	87
39	48	45	41	87
40	43	41	39	86
41	47	42	39	82
42	50	41	35	89
43	54	50	46	89
46	51	43	35	87
48	23	21	18	78
..
533	43	39	36	100
534	45	42	39	93
537	49	47	45	100
539	50	43	40	92
540	46	44	42	86
541	47	44	40	90
544	50	47	44	100
545	45	43	42	86
546	44	42	39	80
547	48	44	41	97
548	46	42	37	97
549	46	42	38	93
550	47	43	39	90
551	45	41	35	89
559	49	46	42	93
561	44	41	37	93
562	47	44	40	80
564	49	46	43	96
565	47	44	40	97
569	50	47	44	80
573	50	46	42	100
574	47	44	38	90
580	55	53	51	97
581	53	50	49	100
584	50	48	43	97
586	51	38	28	90
592	49	44	37	83
593	56	48	31	96
610	47	45	42	86
611	47	42	38	90

	Mean_Humidity	Min_Humidity	...	Mean_Visibility_Miles	\
10	76	62	...	10	
11	75	60	...	10	
13	71	62	...	10	
14	72	48	...	10	
15	80	72	...	8	
18	83	67	...	8	

19	78	62	...	10
20	82	62	...	9
23	79	64	...	10
25	72	53	...	10
26	78	60	...	8
27	80	72	...	9
28	71	54	...	10
29	42	20	...	10
30	24	16	...	10
31	30	18	...	10
32	48	23	...	10
33	51	21	...	10
34	53	25	...	10
35	60	34	...	10
36	75	54	...	8
37	66	48	...	9
38	80	71	...	10
39	79	66	...	8
40	77	66	...	10
41	74	57	...	10
42	79	66	...	8
43	83	77	...	6
46	80	72	...	7
48	64	47	...	10
..
533	74	38	...	9
534	64	37	...	10
537	80	58	...	10
539	72	51	...	10
540	77	64	...	10
541	67	38	...	10
544	73	48	...	10
545	71	57	...	10
546	67	51	...	10
547	79	59	...	9
548	73	44	...	10
549	73	49	...	10
550	70	48	...	10
551	62	31	...	10
559	83	66	...	9
561	69	44	...	10
562	67	46	...	10
564	74	53	...	10
565	64	37	...	10
569	72	58	...	10
573	72	46	...	9
574	64	37	...	10
580	91	86	...	10

581	82	69	...	8
584	83	52	...	9
586	56	32	...	9
592	67	38	...	10
593	78	37	...	10
610	73	53	...	10
611	63	36	...	10

	Min_Visibility_Miles	Max_Wind_Speed_MPH	Mean_Wind_Speed_MPH	\
10	10	15	9	
11	8	8	4	
13	10	20	12	
14	10	7	4	
15	2	14	5	
18	2	8	4	
19	7	6	1	
20	4	9	5	
23	3	7	2	
25	10	7	3	
26	2	7	1	
27	3	12	6	
28	10	8	2	
29	10	15	5	
30	10	13	8	
31	10	12	4	
32	10	8	2	
33	10	7	1	
34	10	8	1	
35	9	8	1	
36	5	6	1	
37	5	9	2	
38	7	8	4	
39	4	18	7	
40	10	13	8	
41	5	20	10	
42	2	13	6	
43	2	17	6	
46	2	15	9	
48	10	8	2	
..	
533	1	12	4	
534	10	8	2	
537	10	12	4	
539	5	18	12	
540	10	14	8	
541	10	12	4	
544	10	6	2	
545	10	9	7	

546	10	10	6
547	4	16	8
548	10	13	6
549	6	14	6
550	10	14	6
551	10	8	4
559	5	10	6
561	10	8	4
562	10	12	2
564	5	8	4
565	10	12	3
569	10	14	5
573	2	13	7
574	10	8	5
580	6	8	6
581	3	6	3
584	4	13	7
586	4	10	3
592	10	12	7
593	4	14	9
610	10	21	10
611	10	10	7

	Max_Gust_Speed_MPH	Precipitation_In	Events	Mean_Temperature_C \
10	22	0.35	Rain	12.78
11	9.26	0.13	Rain	13.33
13	30	0.05	Rain	11.11
14	9.26	0.01	Rain	12.22
15	9.26	0.34	Rain	13.33
18	9.26	0.77	Rain	12.22
19	9.26	0.00	Rain	11.11
20	9.26	0.11	Rain	10.56
23	9.26	0.27	Rain	13.89
25	9.26	0.00	Rain	11.11
26	9.26	0.00	Rain	8.89
27	16	0.29	Rain	11.11
28	9.26	0.00	Rain	8.89
29	22	0.00	Rain	5.56
30	22	0.00	Rain	3.89
31	22	0.00	Rain	4.44
32	9.26	0.00	Rain	3.33
33	9.26	0.00	Rain	3.33
34	9.26	0.00	Rain	3.89
35	9.26	0.00	Rain	5.00
36	9.26	0.00	Rain	3.89
37	9.26	0.00	Rain	6.67
38	9.26	0.11	Rain	9.44
39	33	0.67	Rain	11.11

40	21	0.03	Rain	8.89
41	28	0.42	Rain	10.00
42	25	0.01	Rain	8.33
43	29	0.33	Rain	13.33
46	26	1.39	Rain	8.89
48	9.26	0.00	Rain	0.56
..
533	20	0.00	Rain	10.00
534	9.26	0.00	Rain	12.22
537	9.26	0.00	Rain	12.22
539	30	0.17	Rain	12.22
540	21	0.00	Rain	11.11
541	17	0.00	Rain	13.33
544	9.26	0.00	Rain	12.78
545	9.26	0.00	Rain	11.11
546	9.26	0.00	Rain	12.22
547	23	0.44	Rain	11.11
548	17	0.00	Rain	11.11
549	32	0.20	Rain	11.11
550	17	0.00	Rain	12.22
551	9.26	0.00	Rain	13.33
559	18	0.27	Rain	10.56
561	9.26	0.00	Rain	11.11
562	9.26	0.00	Rain	13.33
564	9.26	0.04	Rain	12.78
565	18	0.00	Rain	13.33
569	9.26	0.00	Rain	13.33
573	16	0.02	Rain	13.33
574	9.26	0.00	Rain	13.89
580	9.26	0.00	Rain	13.33
581	9.26	0.00	Rain	13.33
584	20	0.22	Rain	12.78
586	9.26	0.04	Rain	13.33
592	9.26	0.02	Rain	13.89
593	23	0.03	Rain	13.33
610	28	0.05	Rain	12.22
611	16	0.00	Rain	13.33

	Events_cat	Events_num
10	Rain	2
11	Rain	2
13	Rain	2
14	Rain	2
15	Rain	2
18	Rain	2
19	Rain	2
20	Rain	2
23	Rain	2

25	Rain	2
26	Rain	2
27	Rain	2
28	Rain	2
29	Rain	2
30	Rain	2
31	Rain	2
32	Rain	2
33	Rain	2
34	Rain	2
35	Rain	2
36	Rain	2
37	Rain	2
38	Rain	2
39	Rain	2
40	Rain	2
41	Rain	2
42	Rain	2
43	Rain	2
46	Rain	2
48	Rain	2
..
533	Rain	2
534	Rain	2
537	Rain	2
539	Rain	2
540	Rain	2
541	Rain	2
544	Rain	2
545	Rain	2
546	Rain	2
547	Rain	2
548	Rain	2
549	Rain	2
550	Rain	2
551	Rain	2
559	Rain	2
561	Rain	2
562	Rain	2
564	Rain	2
565	Rain	2
569	Rain	2
573	Rain	2
574	Rain	2
580	Rain	2
581	Rain	2
584	Rain	2
586	Rain	2

592	Rain	2
593	Rain	2
610	Rain	2
611	Rain	2

[339 rows x 24 columns]

As we can see that rain is our mean Event We will fill all empty event with 6

```
In [38]: # fill in missing values with a specified value
df['Events'].fillna(value='Rain', inplace=True)
```

The Events_num is appera as -1 which is not right, to fix this problem we repeate above step

```
In [39]: df["Events_cat"] = df["Events"].astype('category')
df["Events_num"] = df["Events_cat"].cat.codes
```

```
In [40]: # count the number of missing values in each Series
df.isnull().sum()
```

```
Out[40]: Date                                0
Max_Temperature_F                           0
Mean_Temperature_F                           1
Min_TemperatureF                             0
Max_Dew_Point_F                             0
MeanDew_Point_F                             0
Min_Dewpoint_F                              0
Max_Humidity                                0
Mean_Humidity                               0
Min_Humidity                                0
Max_Sea_Level_Pressure_In                   0
Mean_Sea_Level_Pressure_In                  0
Min_Sea_Level_Pressure_In                   0
Max_Visibility_Miles                        0
Mean_Visibility_Miles                       0
Min_Visibility_Miles                        0
Max_Wind_Speed_MPH                          0
Mean_Wind_Speed_MPH                         0
Max_Gust_Speed_MPH                          0
Precipitation_In                            0
Events                                       0
Mean_Temperature_C                          1
Events_cat                                  0
Events_num                                  0
dtype: int64
```

```
In [41]: # confirm that the missing values were filled in
df['Events'].value_counts().head()
```

```
Out[41]: Rain          346
        Sunny         325
        Fog, Rain      10
        Rain-Thunderstorm  3
        Fog            2
        Name: Events, dtype: int64
```

```
In [42]: # if 'any' values are missing in a row, then drop that row
df.dropna(how='any').shape
```

```
Out[42]: (688, 24)
```

```
In [43]: # count the number of missing values in each Series
df.isnull().sum()
```

```
Out[43]: Date          0
        Max_Temperature_F  0
        Mean_Temperature_F  1
        Min_TemperatureF  0
        Max_Dew_Point_F  0
        MeanDew_Point_F  0
        Min_Dewpoint_F  0
        Max_Humidity  0
        Mean_Humidity  0
        Min_Humidity  0
        Max_Sea_Level_Pressure_In  0
        Mean_Sea_Level_Pressure_In  0
        Min_Sea_Level_Pressure_In  0
        Max_Visibility_Miles  0
        Mean_Visibility_Miles  0
        Min_Visibility_Miles  0
        Max_Wind_Speed_MPH  0
        Mean_Wind_Speed_MPH  0
        Max_Gust_Speed_MPH  0
        Precipitation_In  0
        Events  0
        Mean_Temperature_C  1
        Events_cat  0
        Events_num  0
        dtype: int64
```

```
In [44]: # fill in missing values with a specified value
df['Mean_Temperature_F'].fillna(value='57', inplace=True)
```

```
In [45]: # count the number of missing values in each Series
df.isnull().sum()
```

```
Out[45]: Date          0
        Max_Temperature_F  0
```

Mean_Temperature_F	0
Min_TemperatureF	0
Max_Dew_Point_F	0
MeanDew_Point_F	0
Min_Dewpoint_F	0
Max_Humidity	0
Mean_Humidity	0
Min_Humidity	0
Max_Sea_Level_Pressure_In	0
Mean_Sea_Level_Pressure_In	0
Min_Sea_Level_Pressure_In	0
Max_Visibility_Miles	0
Mean_Visibility_Miles	0
Min_Visibility_Miles	0
Max_Wind_Speed_MPH	0
Mean_Wind_Speed_MPH	0
Max_Gust_Speed_MPH	0
Precipitation_In	0
Events	0
Mean_Temperature_C	1
Events_cat	0
Events_num	0
dtype: int64	

In [46]: df['Mean_Temperature_C'].mean()

Out[46]: 13.657965116279073

In [47]: df['Mean_Temperature_C'].fillna(value='14', inplace=True)

In [48]: *# count the number of missing values in each Series*
df.isnull().sum()

Date	0
Max_Temperature_F	0
Mean_Temperature_F	0
Min_TemperatureF	0
Max_Dew_Point_F	0
MeanDew_Point_F	0
Min_Dewpoint_F	0
Max_Humidity	0
Mean_Humidity	0
Min_Humidity	0
Max_Sea_Level_Pressure_In	0
Mean_Sea_Level_Pressure_In	0
Min_Sea_Level_Pressure_In	0
Max_Visibility_Miles	0
Mean_Visibility_Miles	0
Min_Visibility_Miles	0

```

Max_Wind_Speed_MPH      0
Mean_Wind_Speed_MPH     0
Max_Gust_Speed_MPH      0
Precipitation_In        0
Events                  0
Mean_Temperature_C      0
Events_cat              0
Events_num              0
dtype: int64

```

```

In [49]: df['Mean_Temperature_F'] = df['Mean_Temperature_F'].astype(int)
df['Mean_Temperature_F'].mean()

```

```

Out[49]: 56.58490566037736

```

```

In [50]: # 'isnull' returns a DataFrame of booleans (True if missing, False if not missing)
df.isnull().tail()

```

```

Out[50]:      Date Max_Temperature_F Mean_Temperature_F Min_TemperatureF \
684  False                False                False                False
685  False                False                False                False
686  False                False                False                False
687  False                False                False                False
688  False                False                False                False

      Max_Dew_Point_F MeanDew_Point_F Min_Dewpoint_F Max_Humidity Mean_Humidity \
684                False                False                False                False                False
685                False                False                False                False                False
686                False                False                False                False                False
687                False                False                False                False                False
688                False                False                False                False                False

      Min_Humidity  ...      Mean_Visibility_Miles Min_Visibility_Miles \
684                False  ...                False                False
685                False  ...                False                False
686                False  ...                False                False
687                False  ...                False                False
688                False  ...                False                False

      Max_Wind_Speed_MPH Mean_Wind_Speed_MPH Max_Gust_Speed_MPH \
684                False                False                False
685                False                False                False
686                False                False                False
687                False                False                False
688                False                False                False

      Precipitation_In Events Mean_Temperature_C Events_cat Events_num
684                False  False                False        False        False
685                False  False                False        False        False

```

686	False	False	False	False	False
687	False	False	False	False	False
688	False	False	False	False	False

[5 rows x 24 columns]

16 Encoding Data

```
In [51]: # convert 'Time' to datetime format
df['Date'] = pd.to_datetime(df.Date)
```

```
In [52]: df['month'] = df['Date'].dt.month
```

```
In [53]: # remove a single column (axis=1 refers to columns)
df.drop('Events_cat', axis=1, inplace=True)
```

```
In [54]: df['year']=df['Date'].dt.year
```

17 Removing Duplicated Tuples

```
In [55]: # detect duplicate Date codes: True if an item is identical to a previous item
df.Date.duplicated().tail()
```

```
Out[55]: 684    False
        685    False
        686    False
        687    False
        688    False
        Name: Date, dtype: bool
```

```
In [56]: # count the duplicate items (True becomes 1, False becomes 0)
df.Date.duplicated().sum()
```

```
Out[56]: 0
```

The above codes shows that we do not have any duplicated rows in our dataset

```
In [57]: df.head(100)
```

```
Out[57]:
```

	Date	Max_Temperature_F	Mean_Temperature_F	Min_TemperatureF	\
0	2014-10-13	71	62	54	
1	2014-10-14	63	59	55	
2	2014-10-15	62	58	54	
3	2014-10-16	71	61	52	
4	2014-10-17	64	60	57	
5	2014-10-18	68	64	59	
6	2014-10-19	73	64	55	
7	2014-10-20	66	60	55	

8	2014-10-21	64	58	55
9	2014-10-22	60	58	57
10	2014-10-23	62	55	50
11	2014-10-24	60	56	51
12	2014-10-25	64	58	52
13	2014-10-26	59	52	48
14	2014-10-27	62	54	45
15	2014-10-28	62	56	51
16	2014-10-29	64	59	54
17	2014-10-30	62	58	55
18	2014-10-31	59	54	52
19	2014-11-01	55	52	48
20	2014-11-02	57	51	45
21	2014-11-03	60	58	55
22	2014-11-04	61	58	55
23	2014-11-05	61	57	53
24	2014-11-06	64	60	54
25	2014-11-07	59	52	45
26	2014-11-08	55	48	42
27	2014-11-09	57	52	48
28	2014-11-10	54	48	43
29	2014-11-11	48	42	36
..
70	2014-12-22	54	48	42
71	2014-12-23	53	48	43
72	2014-12-24	46	44	41
73	2014-12-25	48	42	37
74	2014-12-26	44	42	39
75	2014-12-27	52	47	42
76	2014-12-28	46	44	41
77	2014-12-29	46	42	35
78	2014-12-30	39	34	30
79	2014-12-31	41	34	26
80	2015-01-01	43	35	27
81	2015-01-02	44	38	32
82	2015-01-03	43	38	33
83	2015-01-04	54	48	41
84	2015-01-05	57	56	54
85	2015-01-06	55	50	46
86	2015-01-07	50	47	44
87	2015-01-08	48	44	39
88	2015-01-09	50	44	39
89	2015-01-10	48	46	44
90	2015-01-11	51	48	46
91	2015-01-12	53	49	45
92	2015-01-13	48	44	41
93	2015-01-14	46	42	37
94	2015-01-15	46	41	36

95	2015-01-16	55	50	44
96	2015-01-17	55	46	39
97	2015-01-18	60	54	48
98	2015-01-19	52	49	46
99	2015-01-20	51	46	39

	Max_Dew_Point_F	MeanDew_Point_F	Min_Dewpoint_F	Max_Humidity	\
0	55	51	46	87	
1	52	51	50	88	
2	53	50	46	87	
3	49	46	42	83	
4	55	51	41	87	
5	59	57	55	90	
6	57	55	53	94	
7	57	54	50	90	
8	52	49	46	87	
9	55	53	48	88	
10	49	47	44	86	
11	50	47	44	86	
12	53	49	44	87	
13	48	44	42	86	
14	46	43	41	87	
15	54	50	45	88	
16	54	51	50	88	
17	55	53	50	88	
18	52	49	46	89	
19	48	45	42	89	
20	52	47	42	90	
21	53	51	48	84	
22	54	51	48	90	
23	55	51	48	90	
24	55	50	45	90	
25	45	43	40	86	
26	44	42	39	93	
27	50	46	43	87	
28	43	40	34	86	
29	39	18	3	80	
..	
70	45	40	37	86	
71	52	42	36	96	
72	39	38	37	89	
73	39	37	34	87	
74	37	36	34	87	
75	45	40	36	87	
76	39	38	35	85	
77	36	28	21	87	
78	24	17	14	73	
79	25	22	18	78	

80	28	25	21	81
81	39	32	25	86
82	38	35	30	93
83	48	41	37	89
84	52	50	48	88
85	48	46	43	93
86	46	43	41	93
87	41	38	36	89
88	41	39	37	93
89	45	44	41	93
90	46	45	42	89
91	43	41	39	90
92	41	38	37	87
93	38	34	28	89
94	41	34	27	86
95	45	40	36	87
96	52	41	35	90
97	51	44	42	82
98	43	42	40	83
99	39	38	34	87

	Mean_Humidity	Min_Humidity	...	Min_Visibility_Miles	\
0	68	46	...	4	
1	78	63	...	3	
2	77	67	...	3	
3	61	36	...	10	
4	72	46	...	6	
5	83	68	...	2	
6	74	52	...	6	
7	78	67	...	5	
8	70	58	...	6	
9	81	67	...	2	
10	76	62	...	10	
11	75	60	...	8	
12	78	58	...	6	
13	71	62	...	10	
14	72	48	...	10	
15	80	72	...	2	
16	76	60	...	9	
17	82	75	...	2	
18	83	67	...	2	
19	78	62	...	7	
20	82	62	...	4	
21	80	75	...	3	
22	79	67	...	3	
23	79	64	...	3	
24	75	58	...	4	
25	72	53	...	10	

26	78	60	...	2
27	80	72	...	3
28	71	54	...	10
29	42	20	...	10
..
70	72	55	...	10
71	80	66	...	2
72	80	71	...	7
73	79	66	...	6
74	80	71	...	4
75	78	69	...	5
76	79	71	...	8
77	62	40	...	10
78	51	35	...	10
79	60	41	...	10
80	70	49	...	6
81	76	67	...	7
82	84	76	...	3
83	84	77	...	3
84	81	74	...	2
85	84	74	...	0
86	85	74	...	1
87	82	71	...	0
88	86	68	...	0
89	88	83	...	1
90	85	80	...	0
91	78	64	...	4
92	80	66	...	2
93	77	60	...	1
94	73	63	...	3
95	71	57	...	10
96	84	73	...	3
97	73	67	...	3
98	76	66	...	7
99	76	59	...	9

	Max_Wind_Speed_MPH	Mean_Wind_Speed_MPH	Max_Gust_Speed_MPH	\
0	13	4	21	
1	10	5	17	
2	18	7	25	
3	9	4	9.26	
4	8	3	9.26	
5	10	4	9.26	
6	10	3	18	
7	12	5	9.26	
8	15	8	21	
9	14	8	22	
10	15	9	22	

11	8	4	9.26
12	24	6	41
13	20	12	30
14	7	4	9.26
15	14	5	9.26
16	15	7	24
17	8	2	9.26
18	8	4	9.26
19	6	1	9.26
20	9	5	9.26
21	13	8	20
22	10	6	9.26
23	7	2	9.26
24	20	8	37
25	7	3	9.26
26	7	1	9.26
27	12	6	16
28	8	2	9.26
29	15	5	22
..
70	8	1	9.26
71	14	5	9.26
72	8	5	9.26
73	7	3	9.26
74	8	4	9.26
75	15	8	25
76	8	5	9.26
77	9	3	9.26
78	6	1	9.26
79	7	2	9.26
80	4	0	9.26
81	12	4	9.26
82	7	2	9.26
83	21	7	32
84	20	14	31
85	6	2	9.26
86	6	2	9.26
87	8	3	9.26
88	6	2	9.26
89	7	1	9.26
90	5	1	9.26
91	10	1	9.26
92	10	2	9.26
93	10	3	9.26
94	16	3	9.26
95	13	8	22
96	17	1	23
97	22	14	36

98	10	5	9.26
99	9	1	9.26

	Precipitation_In	Events	Mean_Temperature_C	Events_num	month	year
0	0.00	Sunny	16.67	6	10	2014
1	0.11	Sunny	15	6	10	2014
2	0.45	Sunny	14.44	6	10	2014
3	0.00	Sunny	16.11	6	10	2014
4	0.14	Sunny	15.56	6	10	2014
5	0.31	Sunny	17.78	6	10	2014
6	0.00	Sunny	17.78	6	10	2014
7	0.44	Sunny	15.56	6	10	2014
8	0.10	Sunny	14.44	6	10	2014
9	1.43	Sunny	14.44	6	10	2014
10	0.35	Rain	12.78	2	10	2014
11	0.13	Rain	13.33	2	10	2014
12	0.37	Sunny	14.44	6	10	2014
13	0.05	Rain	11.11	2	10	2014
14	0.01	Rain	12.22	2	10	2014
15	0.34	Rain	13.33	2	10	2014
16	0.04	Sunny	15	6	10	2014
17	0.67	Sunny	14.44	6	10	2014
18	0.77	Rain	12.22	2	10	2014
19	0.00	Rain	11.11	2	11	2014
20	0.11	Rain	10.56	2	11	2014
21	0.24	Sunny	14.44	6	11	2014
22	0.05	Sunny	14.44	6	11	2014
23	0.27	Rain	13.89	2	11	2014
24	0.22	Sunny	15.56	6	11	2014
25	0.00	Rain	11.11	2	11	2014
26	0.00	Rain	8.89	2	11	2014
27	0.29	Rain	11.11	2	11	2014
28	0.00	Rain	8.89	2	11	2014
29	0.00	Rain	5.56	2	11	2014
..
70	0.00	Rain	8.89	2	12	2014
71	0.61	Rain	8.89	2	12	2014
72	0.12	Rain	6.67	2	12	2014
73	0.00	Rain	5.56	2	12	2014
74	0.00	Rain	5.56	2	12	2014
75	0.12	Rain	8.33	2	12	2014
76	0.06	Rain	6.67	2	12	2014
77	0.00	Rain	5.56	2	12	2014
78	0.00	Rain	1.11	2	12	2014
79	0.00	Rain	1.11	2	12	2014
80	0.00	Rain	1.67	2	1	2015
81	0.03	Rain	3.33	2	1	2015
82	0.00	Rain	3.33	2	1	2015

83	0.22	Rain	8.89	2	1	2015
84	0.07	Rain	13.33	2	1	2015
85	0.01	Fog	10	0	1	2015
86	0.00	Rain	8.33	2	1	2015
87	0.00	Rain	6.67	2	1	2015
88	0.01	Fog, Rain	6.67	1	1	2015
89	0.18	Rain	7.78	2	1	2015
90	0.06	Fog, Rain	8.89	1	1	2015
91	0.00	Rain	9.44	2	1	2015
92	0.00	Rain	6.67	2	1	2015
93	0.00	Rain	5.56	2	1	2015
94	0.43	Rain	5	2	1	2015
95	0.00	Rain	10	2	1	2015
96	0.76	Rain	7.78	2	1	2015
97	0.23	Rain	12.22	2	1	2015
98	0.03	Rain	9.44	2	1	2015
99	0.00	Rain	7.78	2	1	2015

[100 rows x 25 columns]

```
In [58]: df['year'] = df['year'].astype(int)
```

```
In [59]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 689 entries, 0 to 688
Data columns (total 25 columns):
Date                689 non-null datetime64[ns]
Max_Temperature_F   689 non-null int64
Mean_Temperature_F   689 non-null int32
Min_TemperatureF     689 non-null int64
Max_Dew_Point_F     689 non-null int64
MeanDew_Point_F     689 non-null int64
Min_Dewpoint_F      689 non-null int64
Max_Humidity         689 non-null int64
Mean_Humidity        689 non-null int64
Min_Humidity         689 non-null int64
Max_Sea_Level_Pressure_In 689 non-null float64
Mean_Sea_Level_Pressure_In 689 non-null float64
Min_Sea_Level_Pressure_In 689 non-null float64
Max_Visibility_Miles 689 non-null int64
Mean_Visibility_Miles 689 non-null int64
Min_Visibility_Miles 689 non-null int64
Max_Wind_Speed_MPH   689 non-null int64
Mean_Wind_Speed_MPH  689 non-null int64
Max_Gust_Speed_MPH   689 non-null object
Precipitation_In     689 non-null float64
Events               689 non-null object
```

```

Mean_Temperature_C      689 non-null object
Events_num              689 non-null int8
month                   689 non-null int64
year                    689 non-null int32
dtypes: datetime64[ns](1), float64(4), int32(2), int64(14), int8(1), object(3)
memory usage: 124.6+ KB

```

```
In [60]: df.Date.tail()
```

```

Out[60]: 684    2016-08-27
        685    2016-08-28
        686    2016-08-29
        687    2016-08-30
        688    2016-08-31
        Name: Date, dtype: datetime64[ns]

```

```

In [61]: # df1 belong to weather dataset
df1 = df[(df.year != 2014)]
df1.tail()

```

```

Out[61]:
      Date  Max_Temperature_F  Mean_Temperature_F  Min_TemperatureF  \
684 2016-08-27                72                66                61
685 2016-08-28                75                68                59
686 2016-08-29                81                68                55
687 2016-08-30                70                64                57
688 2016-08-31                71                65                59

      Max_Dew_Point_F  MeanDew_Point_F  Min_Dewpoint_F  Max_Humidity  \
684                57                54                50            81
685                54                52                50            80
686                55                53                50            89
687                55                53                52            83
688                61                56                52            90

      Mean_Humidity  Min_Humidity  ...  Min_Visibility_Miles  \
684                65            46  ...                10
685                65            44  ...                10
686                65            39  ...                 6
687                69            53  ...                10
688                77            63  ...                 8

      Max_Wind_Speed_MPH  Mean_Wind_Speed_MPH  Max_Gust_Speed_MPH  \
684                    16                    9                9.26
685                    12                    9                9.26
686                     9                    4                9.26
687                    14                    9                 18
688                    14                    8                9.26

```

	Precipitation_In	Events	Mean_Temperature_C	Events_num	month	year
684	0.0	Sunny	18.89	6	8	2016
685	0.0	Sunny	20	6	8	2016
686	0.0	Sunny	20	6	8	2016
687	0.0	Sunny	17.78	6	8	2016
688	0.0	Sunny	18.33	6	8	2016

[5 rows x 25 columns]

18 Saving Weather Data

Weather data consist on 600 observations where bikes data is consisting over a hundred thousand of tuples which makles it hard to combine both data set for further weather analysis. To achieve this we will, break both weather and bikes data Date columns and applying programming technique to achieve weather value for each bicycle trip.

```
In [62]: df1.isnull().sum()
```

```
Out[62]: Date                                0
Max_Temperature_F                           0
Mean_Temperature_F                           0
Min_TemperatureF                             0
Max_Dew_Point_F                             0
MeanDew_Point_F                             0
Min_Dewpoint_F                              0
Max_Humidity                                0
Mean_Humidity                               0
Min_Humidity                                0
Max_Sea_Level_Pressure_In                   0
Mean_Sea_Level_Pressure_In                  0
Min_Sea_Level_Pressure_In                   0
Max_Visibility_Miles                        0
Mean_Visibility_Miles                       0
Min_Visibility_Miles                        0
Max_Wind_Speed_MPH                          0
Mean_Wind_Speed_MPH                         0
Max_Gust_Speed_MPH                          0
Precipitation_In                            0
Events                                       0
Mean_Temperature_C                          0
Events_num                                  0
month                                        0
year                                        0
dtype: int64
```

19 Data type Conversion

```
In [63]: df1['Mean_Temperature_C'] = df1['Mean_Temperature_C'].astype(int)
df1.dtypes
```

```
Out [63]: Date                                datetime64[ns]
Max_Temperature_F                             int64
Mean_Temperature_F                             int32
Min_TemperatureF                             int64
Max_Dew_Point_F                             int64
MeanDew_Point_F                             int64
Min_Dewpoint_F                             int64
Max_Humidity                                 int64
Mean_Humidity                                 int64
Min_Humidity                                 int64
Max_Sea_Level_Pressure_In                   float64
Mean_Sea_Level_Pressure_In                   float64
Min_Sea_Level_Pressure_In                   float64
Max_Visibility_Miles                       int64
Mean_Visibility_Miles                       int64
Min_Visibility_Miles                       int64
Max_Wind_Speed_MPH                         int64
Mean_Wind_Speed_MPH                         int64
Max_Gust_Speed_MPH                         object
Precipitation_In                           float64
Events                                      object
Mean_Temperature_C                           int32
Events_num                                  int8
month                                       int64
year                                       int32
dtype: object
```

```
In [64]: df1
```

```
Out [64]:
```

	Date	Max_Temperature_F	Mean_Temperature_F	Min_TemperatureF	\
80	2015-01-01	43	35	27	
81	2015-01-02	44	38	32	
82	2015-01-03	43	38	33	
83	2015-01-04	54	48	41	
84	2015-01-05	57	56	54	
85	2015-01-06	55	50	46	
86	2015-01-07	50	47	44	
87	2015-01-08	48	44	39	
88	2015-01-09	50	44	39	
89	2015-01-10	48	46	44	
90	2015-01-11	51	48	46	
91	2015-01-12	53	49	45	
92	2015-01-13	48	44	41	
93	2015-01-14	46	42	37	

94	2015-01-15	46	41	36
95	2015-01-16	55	50	44
96	2015-01-17	55	46	39
97	2015-01-18	60	54	48
98	2015-01-19	52	49	46
99	2015-01-20	51	46	39
100	2015-01-21	48	40	33
101	2015-01-22	52	48	45
102	2015-01-23	57	54	50
103	2015-01-24	62	58	55
104	2015-01-25	64	57	50
105	2015-01-26	64	55	46
106	2015-01-27	55	52	50
107	2015-01-28	55	50	46
108	2015-01-29	55	48	42
109	2015-01-30	48	44	39
..
659	2016-08-02	73	64	57
660	2016-08-03	75	66	60
661	2016-08-04	80	69	59
662	2016-08-05	79	70	61
663	2016-08-06	72	63	55
664	2016-08-07	72	66	60
665	2016-08-08	73	64	57
666	2016-08-09	72	66	60
667	2016-08-10	75	68	60
668	2016-08-11	80	70	61
669	2016-08-12	87	75	63
670	2016-08-13	90	78	66
671	2016-08-14	82	71	60
672	2016-08-15	84	72	61
673	2016-08-16	81	69	57
674	2016-08-17	79	70	60
675	2016-08-18	86	73	60
676	2016-08-19	95	60	26
677	2016-08-20	91	78	64
678	2016-08-21	73	66	60
679	2016-08-22	72	64	57
680	2016-08-23	79	67	55
681	2016-08-24	82	70	57
682	2016-08-25	93	77	61
683	2016-08-26	88	74	60
684	2016-08-27	72	66	61
685	2016-08-28	75	68	59
686	2016-08-29	81	68	55
687	2016-08-30	70	64	57
688	2016-08-31	71	65	59

	Max_Dew_Point_F	MeanDew_Point_F	Min_Dewpoint_F	Max_Humidity	\
80	28	25	21	81	
81	39	32	25	86	
82	38	35	30	93	
83	48	41	37	89	
84	52	50	48	88	
85	48	46	43	93	
86	46	43	41	93	
87	41	38	36	89	
88	41	39	37	93	
89	45	44	41	93	
90	46	45	42	89	
91	43	41	39	90	
92	41	38	37	87	
93	38	34	28	89	
94	41	34	27	86	
95	45	40	36	87	
96	52	41	35	90	
97	51	44	42	82	
98	43	42	40	83	
99	39	38	34	87	
100	39	35	29	93	
101	46	43	38	87	
102	52	49	45	89	
103	55	52	51	87	
104	53	49	45	86	
105	47	46	42	93	
106	49	47	45	88	
107	45	42	40	87	
108	40	38	36	85	
109	40	38	35	96	
..	
659	53	51	50	78	
660	53	52	51	72	
661	56	53	51	81	
662	54	51	46	75	
663	51	49	48	77	
664	56	52	49	83	
665	54	53	48	84	
666	55	54	53	83	
667	56	55	54	83	
668	57	56	54	81	
669	59	57	55	78	
670	60	56	51	78	
671	54	52	44	78	
672	57	55	53	78	
673	56	54	51	90	
674	54	53	50	80	

675	63	59	49	72
676	77	54	9	88
677	57	53	48	70
678	55	54	52	83
679	59	51	46	93
680	55	51	48	83
681	57	55	54	89
682	59	54	50	81
683	57	52	45	82
684	57	54	50	81
685	54	52	50	80
686	55	53	50	89
687	55	53	52	83
688	61	56	52	90

	Mean_Humidity	Min_Humidity	...	Min_Visibility_Miles	\
80	70	49	...	6	
81	76	67	...	7	
82	84	76	...	3	
83	84	77	...	3	
84	81	74	...	2	
85	84	74	...	0	
86	85	74	...	1	
87	82	71	...	0	
88	86	68	...	0	
89	88	83	...	1	
90	85	80	...	0	
91	78	64	...	4	
92	80	66	...	2	
93	77	60	...	1	
94	73	63	...	3	
95	71	57	...	10	
96	84	73	...	3	
97	73	67	...	3	
98	76	66	...	7	
99	76	59	...	9	
100	81	66	...	0	
101	81	76	...	5	
102	84	76	...	2	
103	82	75	...	10	
104	77	54	...	6	
105	77	52	...	0	
106	83	80	...	2	
107	76	64	...	7	
108	71	57	...	10	
109	85	71	...	0	
..	
659	66	49	...	10	

660	62	45	...	10
661	58	38	...	10
662	56	32	...	10
663	62	45	...	10
664	69	49	...	10
665	69	44	...	10
666	70	53	...	10
667	67	48	...	10
668	63	47	...	10
669	56	36	...	10
670	52	28	...	10
671	51	25	...	10
672	57	37	...	10
673	62	35	...	10
674	56	40	...	10
675	57	45	...	10
676	48	23	...	10
677	44	26	...	10
678	68	48	...	10
679	64	40	...	10
680	58	39	...	10
681	60	39	...	10
682	51	22	...	10
683	45	22	...	10
684	65	46	...	10
685	65	44	...	10
686	65	39	...	6
687	69	53	...	10
688	77	63	...	8

	Max_Wind_Speed_MPH	Mean_Wind_Speed_MPH	Max_Gust_Speed_MPH	\
80	4	0	9.26	
81	12	4	9.26	
82	7	2	9.26	
83	21	7	32	
84	20	14	31	
85	6	2	9.26	
86	6	2	9.26	
87	8	3	9.26	
88	6	2	9.26	
89	7	1	9.26	
90	5	1	9.26	
91	10	1	9.26	
92	10	2	9.26	
93	10	3	9.26	
94	16	3	9.26	
95	13	8	22	
96	17	1	23	

97	22	14	36
98	10	5	9.26
99	9	1	9.26
100	5	1	9.26
101	8	1	9.26
102	13	6	21
103	10	7	20
104	8	3	9.26
105	13	4	9.26
106	10	6	9.26
107	6	2	9.26
108	8	1	9.26
109	8	1	9.26
..
659	12	6	17
660	10	6	9.26
661	13	5	21
662	9	4	9.26
663	10	6	9.26
664	9	4	9.26
665	9	4	9.26
666	10	3	9.26
667	12	2	17
668	12	4	20
669	10	3	9.26
670	8	2	9.26
671	10	4	9.26
672	9	2	9.26
673	8	2	9.26
674	12	4	9.26
675	16	8	9.26
676	16	8	9.26
677	10	3	9.26
678	14	8	9.26
679	14	8	9.26
680	17	7	9.26
681	20	5	9.26
682	14	4	9.26
683	10	4	9.26
684	16	9	9.26
685	12	9	9.26
686	9	4	9.26
687	14	9	18
688	14	8	9.26

	Precipitation_In	Events	Mean_Temperature_C	Events_num	month	year
80	0.00	Rain	1	2	1	2015
81	0.03	Rain	3	2	1	2015

82	0.00	Rain	3	2	1	2015
83	0.22	Rain	8	2	1	2015
84	0.07	Rain	13	2	1	2015
85	0.01	Fog	10	0	1	2015
86	0.00	Rain	8	2	1	2015
87	0.00	Rain	6	2	1	2015
88	0.01	Fog, Rain	6	1	1	2015
89	0.18	Rain	7	2	1	2015
90	0.06	Fog, Rain	8	1	1	2015
91	0.00	Rain	9	2	1	2015
92	0.00	Rain	6	2	1	2015
93	0.00	Rain	5	2	1	2015
94	0.43	Rain	5	2	1	2015
95	0.00	Rain	10	2	1	2015
96	0.76	Rain	7	2	1	2015
97	0.23	Rain	12	2	1	2015
98	0.03	Rain	9	2	1	2015
99	0.00	Rain	7	2	1	2015
100	0.00	Rain	4	2	1	2015
101	0.03	Rain	8	2	1	2015
102	0.08	Rain	12	2	1	2015
103	0.02	Sunny	14	6	1	2015
104	0.01	Rain	13	2	1	2015
105	0.00	Rain	12	2	1	2015
106	0.02	Rain	11	2	1	2015
107	0.00	Rain	10	2	1	2015
108	0.00	Rain	8	2	1	2015
109	0.00	Rain	6	2	1	2015
..
659	0.00	Sunny	17	6	8	2016
660	0.00	Sunny	18	6	8	2016
661	0.00	Sunny	20	6	8	2016
662	0.00	Sunny	21	6	8	2016
663	0.00	Sunny	17	6	8	2016
664	0.03	Sunny	18	6	8	2016
665	0.00	Sunny	17	6	8	2016
666	0.00	Sunny	18	6	8	2016
667	0.00	Sunny	20	6	8	2016
668	0.00	Sunny	21	6	8	2016
669	0.00	Sunny	23	6	8	2016
670	0.00	Sunny	25	6	8	2016
671	0.00	Sunny	21	6	8	2016
672	0.00	Sunny	22	6	8	2016
673	0.00	Sunny	20	6	8	2016
674	0.00	Sunny	21	6	8	2016
675	0.00	Sunny	22	6	8	2016
676	0.00	Sunny	15	6	8	2016
677	0.00	Sunny	25	6	8	2016

678	0.00	Sunny	18	6	8	2016
679	0.00	Sunny	17	6	8	2016
680	0.00	Sunny	19	6	8	2016
681	0.00	Sunny	21	6	8	2016
682	0.00	Sunny	25	6	8	2016
683	0.00	Sunny	23	6	8	2016
684	0.00	Sunny	18	6	8	2016
685	0.00	Sunny	20	6	8	2016
686	0.00	Sunny	20	6	8	2016
687	0.00	Sunny	17	6	8	2016
688	0.00	Sunny	18	6	8	2016

[609 rows x 25 columns]

```
In [65]: df1.to_csv('C:\\Users\\mrferozi\\Desktop\\GitHub\\Bike\\dataset\\cycle\\weather_clean
```

19.1 Resources

References: From the video series: [Introduction to machine learning with scikit-learn](#) - scikit-learn documentation: [Cross-validation](#), [Model evaluation](#) - scikit-learn issue on GitHub: [MSE is negative when returned by cross_val_score](#) - Section 5.1 of [An Introduction to Statistical Learning](#) (11 pages) and related videos: [K-fold and leave-one-out cross-validation](#) (14 minutes), [Cross-validation the right and wrong ways](#) (10 minutes) - Scott Fortmann-Roe: [Accurately Measuring Model Prediction Error](#) - Machine Learning Mastery: [An Introduction to Feature Selection](#) - Harvard CS109: [Cross-Validation: The Right and Wrong Way](#) - Journal of Cheminformatics: [Cross-validation pitfalls when selecting and assessing regression and classification models](#)