```
In [1]: # Basic Libraries
        import numpy as np
        import pandas as pd
        # Feature Scaling
        from sklearn.preprocessing import RobustScaler
        # Visaulization
        import matplotlib
        import matplotlib.pyplot as plt
        import seaborn as sns
        # Classifier (machine learning algorithm)
        from sklearn.linear model import LogisticRegression
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.svm import SVC
        from sklearn.naive bayes import GaussianNB
        from sklearn.tree import DecisionTreeClassifier
        from sklearn.ensemble import RandomForestClassifier
        from sklearn import metrics
        from sklearn.metrics import classification report
        from sklearn.discriminant analysis import LinearDiscriminantAnalysis as LDA
        from sklearn.linear_model import Ridge, RidgeCV
        from sklearn.model selection import LeavePOut
        from sklearn import linear model
        from sklearn.ensemble import AdaBoostClassifier
        from sklearn.model selection import RandomizedSearchCV
        # Evaluation
        from sklearn.model_selection import cross_val_score, cross_val_predict
        # Parameter Tuning
        from sklearn.model selection import GridSearchCV
        # Settings
        pd.options.mode.chained assignment = None # Stop warning when use inplace=True
         of fillna
        train = pd.read csv("sfpd-dispatch/sfpd dispatch data subset.csv", index col=0
        #train.head()
        # test = pd.read_csv("test_final.csv", index_col=0)
```

In [2]: train.describe()

Out[2]:

	call_number	incident_number	zipcode_of_incident	station_area	kod
count	1.000000e+04	1.000000e+04	10000.000000	10000.000000	10000.000000
mean	1.801872e+08	1.800791e+07	94113.473000	18.066600	4019.476500
std	3.457123e+04	1.417024e+03	10.490342	14.350699	2327.782885
min	1.801300e+08	1.800543e+07	94102.000000	1.000000	131.000000
25%	1.801537e+08	1.800669e+07	94103.750000	5.000000	2117.000000
50%	1.801902e+08	1.800793e+07	94110.000000	14.000000	3426.000000
75%	1.802201e+08	1.800914e+07	94121.000000	32.000000	5667.000000
max	1.802442e+08	1.801035e+07	94158.000000	51.000000	8871.000000

```
In [3]: train.isnull().sum()
Out[3]: call_number
                                                 0
         unit id
                                                 0
         incident_number
                                                 0
         call type
                                                 0
         call date
                                                 0
         watch_date
                                                 0
         received_timestamp
                                                 0
         entry_timestamp
                                                 0
         dispatch_timestamp
                                                 0
         response_timestamp
                                               261
         on scene timestamp
                                              1915
         transport_timestamp
                                              7031
        hospital_timestamp
                                              7087
         call_final_disposition
                                                 0
         available_timestamp
                                                 0
         address
                                                 0
         city
                                                18
         zipcode_of_incident
                                                 0
         battalion
                                                 0
         station area
                                                 0
         box
                                                 0
         original_priority
                                                 0
         priority
                                                 0
         final_priority
                                                 0
         als_unit
                                                 0
         call_type_group
                                                 0
         number of alarms
         unit_sequence_in_call_dispatch
                                                 0
         fire prevention district
                                                 0
         supervisor_district
                                                 0
         neighborhood_district
                                             10000
         location
                                                 0
         row id
                                                 0
         latitude
                                                 0
         longitude
                                                 0
                                                 0
         unit_type
         dtype: int64
```

```
In [4]: # This fills all the nan spots with the text 'nan'
train = train.fillna('nan')
```

```
In [5]: # takes about 4 minutes
        #Dictionary the values for processing
        train['call type'] = train['call type'].map({'Medical Incident' : 0, 'Alarms'
        : 1, 'Structure Fire' : 2, 'Traffic Collision' : 3, 'Outside Fire' : 4, 'Othe
        r' : 5, 'Train / Rail Incident' : 6, 'Citizen Assist / Service Call' : 7, 'Ele
        ctrical Hazard' : 8, 'Elevator / Escalator Rescue' : 9, 'Fuel Spill' : 10, 'Ga
        s Leak (Natural and LP Gases)' : 11, 'Water Rescue' : 12, 'Vehicle Fire' : 13,
         'Smoke Investigation (Outside)' : 14, 'Odor (Strange / Unknown)' : 15, 'HazMa
        t' : 16 }).astype(int)
        train['call final disposition'] = train['call final disposition'].map({'Code 2
         Transport' : 0, 'Fire' : 1, 'Code 3 Transport' : 2, 'Patient Declined Transpo
        rt' : 3, 'No Merit' : 4, 'Medical Examiner' : 5, 'Against Medical Advice' : 6,
         'Cancelled' : 7, 'Unable to Locate' : 8, 'SFPD' : 9, 'Gone on Arrival' : 10,
        'Other' : 11}).astype(int)
        train['city'] = train['city'].map({'nan' : 0, 'San Francisco' : 1, 'Presidio'
        : 2, 'Treasure Isla' : 3, 'Yerba Buena' : 4, 'Hunters Point' : 5, 'Fort Mason'
         : 6 }).astype(int)
        train['battalion'] = train['battalion'].map({'B99': 0, 'B01': 1, 'B02': 2, 'B0
        3': 3, 'B04': 4, 'B05': 5, 'B06': 6, 'B07': 7, 'B08': 8, 'B09': 9, 'B10': 10
        }).astype(int)
        train['original_priority'] = train['original_priority'].map({'I' : 1, '2' : 2,
         '3' : 3, 'A' : 4, 'B' : 5, 'C' : 6, 'E' : 7}).astype(int)
        train['priority'] = train['priority'].map({'I' : 1, '2' : 2, '3' : 3, 'E' : 4
        }).astype(int)
        train['als unit'] = train['als unit'].map({True : 1, False : 0}).astype(int)
        train['call type group'] = train['call type group'].map({'nan' : 0, 'Non Life-
        threatening' : 1, 'Potentially Life-Threatening' : 2, 'Alarm' : 3, 'Fire' : 4
        }).astype(int)
        #Create a travel duration column
        for k in range(0,train.shape[0]):
            train['call date'][k] = int(train['call date'][k][2:4])
            train['watch_date'][k] = int(train['watch_date'][k][2:4])
            timestamps = ['received_timestamp','entry_timestamp','dispatch_timestamp',
        'response_timestamp','on_scene_timestamp','transport_timestamp','hospital_time
        stamp','available_timestamp']
            for stamp in timestamps:
                 if train[stamp][k]=='nan':
                    train[stamp][k] = 0
                else:
                    train[stamp][k] = int(train[stamp][k][17:19]) + 60 * int(train[sta
        mp[[k][14:16]) + 3600 * int(train[stamp][k][11:13]) + 3600 * 24 * int(train[stamp][k][11:13])
        amp][k][8:10])
        train['unit_type'] = train['unit_type'].map({'ENGINE' : 0, 'MEDIC' : 1, 'PRIVA
        TE': 2, 'TRUCK': 3, 'CHIEF': 4, 'RESCUE CAPTAIN': 5, 'RESCUE SQUAD': 6,
        'SUPPORT' : 7, 'INVESTIGATION' : 8 }).astype(int)
```

```
In [6]: #remove noisy data such as neighbouhood_district and call_number for faster
    processing
    drop_elements=['call_number','incident_number','call_date','watch_date','nei
        ghborhood_district','zipcode_of_incident','station_area','address','city','l
        ocation','supervisor_district']
    train = train.drop(drop_elements, axis = 1)
```

```
In [7]: #encode missing times as 99999, will be used for checks
    train['travel_duration'] = train['on_scene_timestamp']-train['dispatch_timestamp']
    for k in range(0, train.shape[0]):
        temp = int(train['travel_duration'][k])
        if (temp > 1500):
            train['travel_duration'][k] = 99999
        if (temp <= 0):
            train['travel_duration'][k] = 99999</pre>
```

```
In [8]: #set bins for graph
    ranges = list(np.arange(0,2000,60))
    incidences_per_unit_type = pd.DataFrame(train.groupby(pd.cut(train.travel_duration, ranges)).count()["unit_type"])
    incidences_per_unit_type.columns = ["count"]

#set parameters for plot
    ax = incidences_per_unit_type.plot.bar(figsize=(15,12), fontsize = 12, legend = False, color = "palegreen")#counts per elevation
    ax.set_xlabel("Time to Arrive (s)", fontsize = 16)
    ax.set_ylabel("Count of Fire Department Calls", fontsize = 16)
    ax.set_title("Count of Fire Department Calls to Response Time", fontsize = 20)
    plt.savefig("bin_check.png")
```

```
In [9]: def get_cat(col, cat):
            for token in col:
                 if(token > 2000):
                     cat.append('Other')
                else:
                     chunks = []
                     temp = int(int(token)/60)
                     min = temp * 60
                     max = min + 60
                     chunks = ['(', str(min), ', ', str(max), ']']
                     category = ''.join(chunks)
                     cat.append(str(category))
        Response_time = []
        get_cat(train["travel_duration"].tolist(), Response_time)
        train["response_cat"] = Response_time
        train['response_cat']
```

```
Out[9]: temp
         MEDIC
                              (840, 900]
                              (780, 840]
         MEDIC
                              (180, 240]
         ENGINE
         ENGINE
                              (120, 180]
         CHIEF
                              (180, 240]
         ENGINE
                              (180, 240]
                              (480, 540]
         ENGINE
         MEDIC
                              (360, 420]
         ENGINE
                              (240, 300]
         MEDIC
                              (240, 300]
         MEDIC
                             (960, 1020]
         MEDIC
                              (360, 420]
                              (180, 240]
         ENGINE
         PRIVATE
                              (540, 600]
                              (240, 300]
         MEDIC
         RESCUE SQUAD
                              (300, 360]
                              (300, 360]
         ENGINE
         ENGINE
                              (300, 360]
                              (300, 360]
         MEDIC
         TRUCK
                                   0ther
         MEDIC
                              (300, 360]
         ENGINE
                              (180, 240]
         ENGINE
                              (180, 240]
         ENGINE
                              (120, 180]
                              (360, 420]
         MEDIC
         MEDIC
                              (300, 360)
         MEDIC
                              (180, 240]
         MEDIC
                                   0ther
         MEDIC
                              (420, 480]
         ENGINE
                                 (0, 60]
                                . . .
                              (180, 240]
         ENGINE
                              (240, 300]
         ENGINE
                              (480, 540]
         MEDIC
         TRUCK
                                   0ther
         RESCUE CAPTAIN
                               (60, 120]
                              (480, 540]
         PRIVATE
         ENGINE
                              (240, 300]
         PRIVATE
                              (660, 720]
                              (240, 300]
         ENGINE
         ENGINE
                              (180, 240]
                              (180, 240]
         ENGINE
         MEDIC
                              (300, 360]
                                   Other
         PRIVATE
         PRIVATE
                              (240, 300]
         MEDIC
                              (480, 540]
         RESCUE CAPTAIN
                              (240, 300]
         ENGINE
                                   0ther
         ENGINE
                              (180, 240]
         MEDIC
                              (480, 540]
         MEDIC
                              (300, 360]
         ENGINE
                                   0ther
         MEDIC
                              (240, 300]
         ENGINE
                                   0ther
                               (60, 120]
         MEDIC
         ENGINE
                                   0ther
```

```
ENGINE (120, 180]

ENGINE (180, 240]

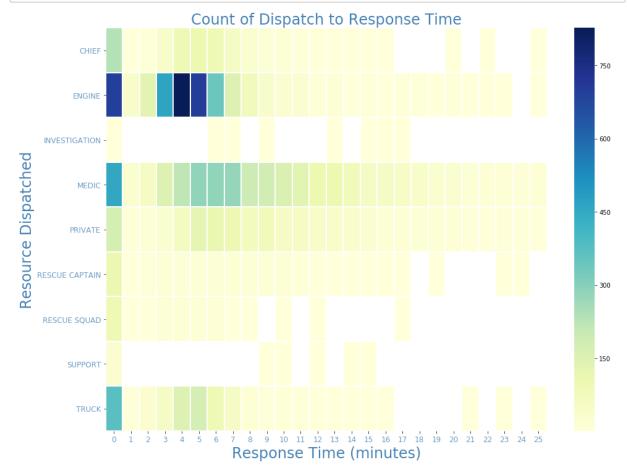
ENGINE Other

SUPPORT Other

MEDIC (420, 480]

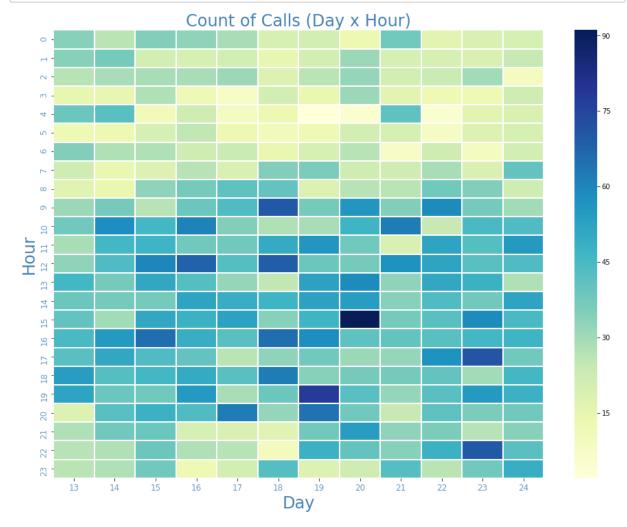
Name: response_cat, Length: 10000, dtype: object
```

```
In [11]:
         %matplotlib inline
         plt.figure(figsize=(16,12))
         incidence_count_matrix_long = pd.DataFrame({'count' : train.groupby( [ "temp",
         "response cat"] ).size()}).reset index()
         incidence count matrix pivot = incidence count matrix long.pivot("temp","respo
         nse_cat","count")
         ax = sns.heatmap(incidence count matrix pivot, annot=False, fmt="d", linewidth
         s=1, square = False, cmap="YlGnBu")
         ax = plt.xticks(fontsize = 12,color="steelblue", alpha=0.8)
         ax = plt.yticks(fontsize = 12,color="steelblue", alpha=0.8)
         ax = plt.xlabel("Response Time (minutes)", fontsize = 24, color="steelblue")
         ax = plt.ylabel("Resource Dispatched", fontsize = 24, color="steelblue")
         ax = plt.title("Count of Dispatch to Response Time", fontsize = 24, color="ste
         elblue")
         plt.savefig('Dispatch response.png')
```



```
In [13]: def get_day_time(col, days, hours, months, years):
              for token in col:
                   day = int(token.split()[0].split("-")[2])
                  month = int(token.split()[0].split("-")[1])
year = int(token.split()[0].split("-")[0])
                   hour = int(token.split()[1].split(":")[0])
                   days.append(day)
                   months.append(month)
                   years.append(year)
                   hours.append(hour)
          hours = []
          days = []
          months = []
          years = []
          get_day_time(train_original["received_timestamp"].tolist(), days, hours, mon
          ths, years)
          train_original["received_hour"] = hours
          train original["received day"] = days
          train original["received month"] = months
          train_original["received_year"] = years
```

```
In [14]: %matplotlib inline
   plt.figure(figsize=(16,12))
   incidence_count_matrix_long = pd.DataFrame({'count' : train_original.groupby(
        [ "received_hour","received_day"] ).size()}).reset_index()
   incidence_count_matrix_pivot = incidence_count_matrix_long.pivot("received_hou
        r","received_day","count")
   ax = sns.heatmap(incidence_count_matrix_pivot, annot=False, fmt="d", linewidth
        s=1, square = False, cmap="YlGnBu")
   ax = plt.xticks(fontsize = 12,color="steelblue", alpha=0.8)
   ax = plt.yticks(fontsize = 12,color="steelblue", alpha=0.8)
   ax = plt.xlabel("Day", fontsize = 24, color="steelblue")
   ax = plt.ylabel("Hour", fontsize = 24, color="steelblue")
   ax = plt.title("Count of Calls (Day x Hour)", fontsize = 24, color="steelblue")
   plt.savefig("call_day.png")
```



In [15]: train.head()

Out[15]:

	unit_id	call_type	received_timestamp	entry_timestamp	dispatch_timestamp
temp					
MEDIC	84	0	2136976	2137101	2137181
MEDIC	61	0	2099105	2099105	2099131
ENGINE	E22	0	2081061	2081137	2081164
ENGINE	E03	1	2145855	2145912	2145924
CHIEF	B03	1	2145788	2145936	2145957

5 rows × 27 columns

```
In [42]: def get_hour(col, hours):
    for token in col:
        hour = int(token.split()[1].split(":")[0])
        hours.append(hour)

#knn_data_subset.csv is a cleaned up data set only containing unit_type, latit
    ude, longitude, and received_timestamp
    hours = []
    train_knn = pd.read_csv("sfpd-dispatch/knn_data_subset.csv", index_col=0)
    get_hour(train_knn["received_timestamp"].tolist(), hours)
    train_knn["hour"] = hours
```

```
In [43]: y = train_knn.unit_type
    train_knn = train_knn.drop("received_timestamp", axis = 1)
    train_knn = train_knn.drop("unit_type", axis = 1)
    X = train_knn
    X.head()
```

Out[43]:

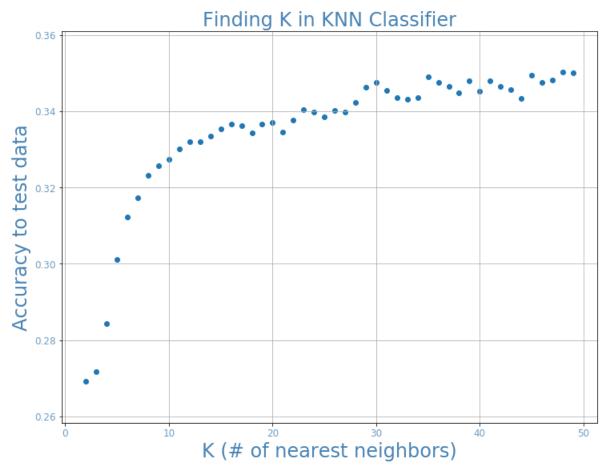
	longitude	hour
latitude		
37.774442	-122.504679	17
37.774095	-122.420001	7
37.755218	-122.475540	2
37.790319	-122.423163	20
37.777328	-122.393089	20

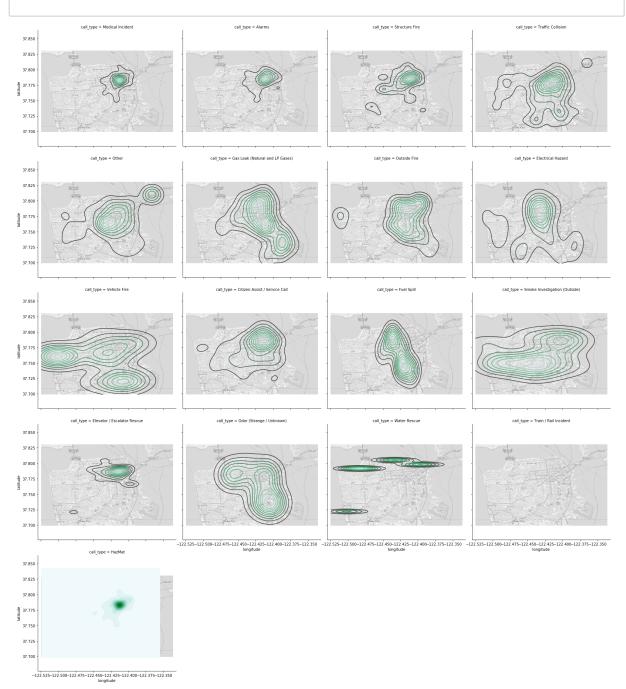
```
In [163]: from sklearn.model selection import train test split
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.8)
In [164]:
          print ("\nX_train:\n")
          print(X train.head())
          print (X_train.shape)
          print ("\nX_test:\n")
          print(X test.head())
          print (X_test.shape)
          X_train:
                      longitude hour
          latitude
          37.781217 -122.457001
                                    15
          37.782724 -122.396168
                                    10
          37.797865 -122.396701
                                    19
          37.751994 -122.409123
                                    11
          37.709148 -122.419780
                                    13
          (2000, 2)
          X test:
                      longitude hour
          latitude
          37.752373 -122.416266
                                    14
          37.777624 -122.399981
                                    12
                                    19
          37.764138 -122.402572
          37.788945 -122.426252
                                    16
          37.734027 -122.432838
                                    11
          (8000, 2)
In [170]:
          #odd number, cho
          from sklearn.neighbors import KNeighborsClassifier
          from sklearn.metrics import accuracy_score
          def get_k(X_train, y_train, X_test, k, acc):
              knn = KNeighborsClassifier(n_neighbors=k)
              knn.fit(X train, y train)
              predictions = knn.predict(X_test)
```

acc.append(accuracy_score(y_test, predictions))

[2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 41, 4 2, 43, 44, 45, 46, 47, 48, 49] [0.269125, 0.271749999999999, 0.284374999999999, 0.3011249999999998, 0.3 1225000000000003, 0.31737500000000002, 0.323249999999998, 0.325625, 0.32737 50000000003, 0.330125, 0.332000000000002, 0.332000000000000, 0.333500000 00000002, 0.3353749999999999, 0.3365000000000002, 0.3361250000000001, 0.33 424999999999, 0.33650000000000000, 0.3371250000000001, 0.334500000000000 2, 0.337749999999999, 0.3403749999999998, 0.33975, 0.33850000000000000, 0. 34012500000000001, 0.33975, 0.3423749999999998, 0.34625, 0.3476250000000000 2, 0.345499999999997, 0.3436250000000001, 0.34312500000000001, 0.343500000 00000003, 0.3489999999999999, 0.34762500000000002, 0.3463749999999999, 0.34 4874999999999, 0.3478749999999999, 0.34525, 0.347874999999999, 0.3463749 999999999, 0.34562500000000002, 0.343374999999999, 0.34949999999998, 0. 347499999999998, 0.34812500000000002, 0.3502500000000001, 0.3499999999999 998]

```
In [178]: plt.figure(figsize=(12,9))
    plt.scatter(k,acc)
    ax = plt.xticks(fontsize = 12,color="steelblue", alpha=0.8)
    ax = plt.yticks(fontsize = 12,color="steelblue", alpha=0.8)
    ax = plt.xlabel('K (# of nearest neighbors)', fontsize = 24, color="steelblue"
    )
    ax = plt.ylabel('Accuracy to test data', fontsize = 24, color="steelblue")
    ax = plt.title('Finding K in KNN Classifier', fontsize = 24, color="steelblue"
    )
    plt.grid(True)
    plt.savefig("knn.png")
```





avg response time is 381.4806784848871seconds

```
In [28]: sum = 0
    count = 0
    trainc = train[train.unit_type == 0]
    for i in trainc['travel_duration'].tolist():
        if(i > 1500):
            continue
        else:
            sum += i
            count += 1
        print("avg engine response time is " + str(sum/count) + "seconds")
```

avg engine response time is 262.61057360055287seconds

```
In [29]: sum = 0
    count = 0
    trainc = train[train.unit_type == 1]
    for i in trainc['travel_duration'].tolist():
        if(i > 1500):
            continue
    else:
        sum += i
        count += 1
    print("avg medic response time is " + str(sum/count) + "seconds")
```

avg medic response time is 496.4316877152698seconds