thesis_test

September 29, 2019

1 Imports

In [1]:
```
import pandas as pd
import psycopg2

%matplotlib inline

import matplotlib
import numpy as np
import matplotlib.pyplot as plt
from datetime import datetime
from sklearn import neighbors, datasets
from matplotlib.colors import ListedColormap
```

2 Conect to db

In [2]:
```
# Define our connection string
conn_string = "host='localhost' port='5432' dbname='GTA' user='jcostamagna'"

# print the connection string we will use to connect
print ("Connecting to database\n        ->")

# get a connection, if a connect cannot be made an exception will be raised here
conn = psycopg2.connect(conn_string)

# conn.cursor will return a cursor object, you can use this cursor to perform queries
cursor = conn.cursor()
print( "Connected!\n" )

#cursor.execute('SELECT * FROM  public."GTA_DataView";')
```

```
Connecting to database
        ->
Connected!
```

3 Create dataframes

In [3]:
```
cursor.execute('SELECT time, idsession, scenario,' +
                        ' MAX(CASE data_type_id WHEN 1 THEN value END) as posX,' +
                        'MAX(CASE data_type_id WHEN 2 THEN value END) as posY,' +
                        ' MAX(CASE data_type_id WHEN 10 THEN value END) as hadCollision,'+
                        ' MAX(CASE data_type_id WHEN 17 THEN value END) as pedId,'+
                        ' MAX(CASE data_type_id WHEN 15 THEN value END) as nextPedRunning,'+
                        ' MAX(CASE data_type_id WHEN 11 THEN value END) as distancePed,'+
                         ' MAX(CASE data_type_id WHEN 8 THEN value END) as speed,'+
                         ' MAX(CASE data_type_id WHEN 12 THEN value END) as posPedX,'+
                         ' MAX(CASE data_type_id WHEN 13 THEN value END) as posPedY,'+
                        ' MAX(CASE data_type_id WHEN 596 THEN value END) as currentDistance'+
                        ' FROM datagta gta WHERE gta.idsession >= 144 AND gta.data_type_id in (
                        ' AND gta.idsession not in (145,146,147,148,149,151,155,158,162,164)'+
                        ' GROUP BY time, idsession, scenario;')
rows = cursor.fetchall()
timestamps = [row[0] for row in rows]
if (len(timestamps) == 0):
    print ("Something went wrong")
firstTime = timestamps[:1][0]
lastTime = timestamps[-1:][0]
dfGTA = pd.DataFrame(rows, index=timestamps)
#dfDiffGTA = dfGTA.diff()
dfGTA.columns=['time','session_id','scenario', 'Pos_X', 'Pos_Y', 'hadCollision', 'pedI
```

In [4]:
```
cursor.execute('SELECT time, idsession, scenario,' +
                        ' MAX(CASE data_type_id WHEN 1 THEN value END) as Steering,' +
                        'MAX(CASE data_type_id WHEN 2 THEN value END) as Brake,' +
                        ' MAX(CASE data_type_id WHEN 3 THEN value END) as Throttle '+
                        ' FROM datasteering gta WHERE gta.idsession >= 144 AND gta.data_type_id
                        ' AND gta.idsession not in (145,146,147,148,149,151,155,158,162,164) '+
                        ' GROUP BY time, idsession, scenario;')
rows = cursor.fetchall()
timestamps = [row[0] for row in rows]
df_steer = pd.DataFrame(rows, index=timestamps)
df_steer.columns=['time2','session_id','scenario', 'steering', 'brake',  'throttle']
freq_resample = '50000U'
df_steer = df_steer.resample(freq_resample).ffill()
freq_resample = '50000U'
dfGTA = dfGTA.resample(freq_resample).ffill()
steer = df_steer.drop(df_steer.columns[[1, 2]], axis=1)
df_joined = pd.concat([dfGTA, steer], axis=1, join='inner')
```

In [5]:
```
#pedCross = dfGTA[(dfGTA.nextPedRunning == True) & (dfGTA.distancePed < 20)  & (dfGTA.
#df_joined['acceleration'] = df_joined['speed'].diff()
df_joined = df_joined[~df_joined.index.duplicated()]
df_joined.head()
```

Out[5]:

	time	session_id	scenario	Pos_X	Pos_Y	hadCollision	pedId	nextPedRunning	distancePed	speed	posPedX	posPedY	currentDistance	time2	steering	brake	throttle
2017-10-04 12:46:18.050	NaT	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		33096.0	65535.0	65535.0
2017-10-04 12:46:18.100	2017-10-04 12:46:18.071232	144.0	7.0	-975.7495	82.24962	0.0	17.0	0.0	70.352844	0.0	154.0415	-23.48739	1253.50293	2017-10-04 12:46:18.088780	33096.0	65535.0	65535.0
2017-10-04 12:46:18.150	2017-10-04 12:46:18.071232	144.0	7.0	-975.7495	82.24962	0.0	17.0	0.0	70.352844	0.0	154.0415	-23.48739	1253.50293	2017-10-04 12:46:18.148940	33096.0	65535.0	65535.0
2017-10-04 12:46:18.200	2017-10-04 12:46:18.071232	144.0	7.0	-975.7495	82.24962	0.0	17.0	0.0	70.352844	0.0	154.0415	-23.48739	1253.50293	2017-10-04 12:46:18.188544	33096.0	65535.0	65535.0
2017-10-04 12:46:18.250	2017-10-04 12:46:18.071232	144.0	7.0	-975.7495	82.24962	0.0	17.0	0.0	70.352844	0.0	154.0415	-23.48739	1253.50293	2017-10-04 12:46:18.248732	33096.0	65535.0	65535.0

4 Backup Csv

In [2]:
```
data = pd.read_csv("DataJoined.csv", index_col=0)
data = data.dropna()
```

In [3]:
```
df_joined = data.copy()
df_joined = df_joined[(df_joined.pedId != 16)  & (df_joined.pedId != 17)  & (df_joined
df_joined.head()
```

Out[3]:

	time	session_id	scenario
2017-10-04 12:50:28.400	2017-10-04 12:50:28.383484	144.0	1.0
2017-10-04 12:50:28.450	2017-10-04 12:50:28.437127	144.0	1.0
2017-10-04 12:50:28.500	2017-10-04 12:50:28.494279	144.0	1.0

```
2017-10-04 12:50:28.550  2017-10-04 12:50:28.520850       144.0       1.0
2017-10-04 12:50:28.600  2017-10-04 12:50:28.584519       144.0       1.0

                              Pos_X      Pos_Y  hadCollision  pedId  \
2017-10-04 12:50:28.400 -975.703700  82.260376           0.0    1.0
2017-10-04 12:50:28.450 -975.709534  82.251080           0.0    1.0
2017-10-04 12:50:28.500 -975.714233  82.246090           0.0    1.0
2017-10-04 12:50:28.550 -975.716000  82.244156           0.0    1.0
2017-10-04 12:50:28.600 -975.718100  82.241540           0.0    1.0

                           nextPedRunning  distancePed     speed   posPedX  \
2017-10-04 12:50:28.400             0.0    89.979675  0.013278 -885.6892
2017-10-04 12:50:28.450             0.0    89.984790  0.135856 -885.6892
2017-10-04 12:50:28.500             0.0    89.988880  0.125762 -885.6892
2017-10-04 12:50:28.550             0.0    89.990330  0.102796 -885.6892
2017-10-04 12:50:28.600             0.0    89.991715  0.047856 -885.6892

                            posPedY  currentDistance  \
2017-10-04 12:50:28.400  70.21315       1253.45900
2017-10-04 12:50:28.450  70.21315       1253.46400
2017-10-04 12:50:28.500  70.21315       1253.46814
2017-10-04 12:50:28.550  70.21315       1253.46960
2017-10-04 12:50:28.600  70.21315       1253.47168

                                              time.1  steering    brake  \
2017-10-04 12:50:28.400  2017-10-04 12:50:28.384988   32307.0  65535.0
2017-10-04 12:50:28.450  2017-10-04 12:50:28.444647   32307.0  65535.0
2017-10-04 12:50:28.500  2017-10-04 12:50:28.484754   32307.0  65535.0
2017-10-04 12:50:28.550  2017-10-04 12:50:28.544914   32307.0  65535.0
2017-10-04 12:50:28.600  2017-10-04 12:50:28.585020   32307.0  65535.0

                           throttle
2017-10-04 12:50:28.400   65535.0
2017-10-04 12:50:28.450   65535.0
2017-10-04 12:50:28.500   65535.0
2017-10-04 12:50:28.550   65535.0
2017-10-04 12:50:28.600   65535.0
```

4.1 Clean Data

```
In [4]: df_joined['steering'] = df_joined['steering'].divide(65535)
        df_joined['brake'] = df_joined['brake'].divide(65535)
        df_joined['throttle'] = df_joined['throttle'].divide(65535)
        df_joined['acceleration'] = df_joined['speed'].diff()
        df_joined = df_joined[(df_joined.pedId != 16)  & (df_joined.pedId != 17)  & (df_joined
```

5 Means fields

```
In [5]: means = df_joined.groupby(['session_id', 'scenario', 'pedId'], as_index=False).mean()
        #clean the ones in which didnt run the pedestrian
        means = means[means.nextPedRunning != 0]
        means_cleaned = means.drop(means.columns[[3,4,6,7,9,10,11]], axis=1)

        #means_cleaned = means.dropna(thresh=11)
        #df_joined = df_joined[df_joined.notnull()]
        #session_id        scenario        pedId        hadCollision speed                steee
        ##means_cleaned['hadCollision'] = np.where(means_cleaned['hadCollision'] == 0, False,

        #means_cleaned.drop(means_cleaned[means_cleaned.nextPedRunning == 0])
        means_cleaned.head()
```

Out[5]:

	session_id	scenario	pedId	hadCollision	speed	steering	brake	throttle	acceleration
0	144.0	1.0	1.0	0.000000	5.919151	0.503649	0.965743	0.820576	0.030731
1	144.0	1.0	3.0	0.000000	7.580378	0.499771	0.891302	0.878839	-0.026652
2	144.0	1.0	15.0	0.008366	9.474048	0.494557	0.952182	0.781126	0.006292
3	144.0	2.0	0.0	0.000000	3.398595	0.523305	0.960227	0.858795	0.027363
4	144.0	2.0	2.0	0.000000	11.669419	0.500661	0.891913	0.522365	0.008028

```
In [6]: means_cleaned.shape
```

```
Out[6]: (655, 9)
```

```
In [7]: #Rows in which the pedestrian didnt run
        means[means.nextPedRunning == 0]
```

```
Out[7]: Empty DataFrame
        Columns: [session_id, scenario, pedId, Pos_X, Pos_Y, hadCollision, nextPedRunning, dist
        Index: []
```

6 Variances

```
In [8]: variance = df_joined.groupby(['session_id', 'scenario', 'pedId'], as_index=False).var()
        variance_cleaned = variance.drop(variance.columns[[3,4,5,6,7,9,10,11]], axis=1)
        #variance_cleaned['hadCollision'] = np.where(variance_cleaned['hadCollision'] == 0, Fa
        variance_cleaned.head()
```

Out[8]:

	session_id	scenario	pedId	speed	steering	brake	throttle
0	144.0	1.0	1.0	13.796202	0.000655	0.014468	0.028719

```
1       144.0       1.0    3.0  31.451253  0.000345  0.058767  0.010391
2       144.0       1.0   15.0  53.873833  0.001231  0.022506  0.045416
3       144.0       2.0    0.0   5.036717  0.000101  0.018643  0.075860
4       144.0       2.0    2.0  47.209285  0.000396  0.055982  0.112551

   acceleration
0      0.039370
1      0.063480
2      0.106281
3      0.027735
4      0.159198
```

```
In [9]: variance_cleaned.shape
```

```
Out[9]: (671, 8)
```

7 Between start, run and crash

7.1 Total time

```
In [10]: first = df_joined.groupby(['session_id', 'scenario', 'pedId'], as_index=False).first()
         last = df_joined.groupby(['session_id', 'scenario', 'pedId'], as_index=False).last()
```

```
In [11]: total_time = first[['session_id', 'scenario', 'pedId']].copy()
         first['timeFinal'] = last['time']
         total_time['total_time'] = (pd.to_datetime(first['timeFinal']) - pd.to_datetime(first
         #plt.plot(first['total_time'])
         #s.dt.total_seconds()
         total_time.head()
```

```
Out[11]:    session_id  scenario  pedId  total_time
         0       144.0       1.0    1.0   15.405173
         1       144.0       1.0    3.0   11.412381
         2       144.0       1.0   15.0  102.356492
         3       144.0       2.0    0.0   14.315100
         4       144.0       2.0    2.0    7.505478
```

7.2 Initial distance

```
In [12]: initial_distance = first[['session_id', 'scenario', 'pedId', 'distancePed']].copy()
         initial_distance.head()
         ## desconfianza con el index 2 y su tiempo
         #144.0        1.0        15.0

         #testing_distance = df_joined[(df_joined.pedId == 15)  & (df_joined.scenario == 1)  &
         #plt.plot(testing_distance['distancePed'])
```

```
Out[12]:    session_id  scenario  pedId  distancePed
         0       144.0       1.0    1.0    89.979675
```

```
1       144.0       1.0    3.0    11.443494
2       144.0       1.0   15.0    10.211066
3       144.0       2.0    0.0    47.977978
4       144.0       2.0    2.0     3.084074
```

```
In [13]: max_values = df_joined.groupby(['session_id', 'scenario', 'pedId'], as_index=False).ma
         initial_distance = max_values[['session_id', 'scenario', 'pedId', 'distancePed']].copy
         initial_distance.head()
```

```
Out[13]:    session_id  scenario  pedId  distancePed
         0       144.0       1.0    1.0    89.992450
         1       144.0       1.0    3.0    85.063860
         2       144.0       1.0   15.0   789.212800
         3       144.0       2.0    0.0    47.977978
         4       144.0       2.0    2.0    88.011610
```

7.3 Max Speed

```
In [14]: max_speed = max_values[['session_id', 'scenario', 'pedId', 'speed']].copy()
         max_speed.head()
```

```
Out[14]:    session_id  scenario  pedId      speed
         0       144.0       1.0    1.0  11.669766
         1       144.0       1.0    3.0  13.499710
         2       144.0       1.0   15.0  25.851397
         3       144.0       2.0    0.0  10.266865
         4       144.0       2.0    2.0  20.055070
```

```
In [15]: max_speed.shape
```

```
Out[15]: (671, 4)
```

7.4 Before pedestrian crossed

```
In [16]: def delete_unused_columns(df):
             df.drop('Pos_X', axis=1, inplace=True, errors='ignore')
             df.drop('Pos_Y', axis=1, inplace=True, errors='ignore')
             df.drop('distancePed', axis=1, inplace=True, errors='ignore')
             df.drop('nextPedRunning', axis=1, inplace=True, errors='ignore')
             df.drop('posPedX', axis=1, inplace=True, errors='ignore')
             df.drop('posPedY', axis=1, inplace=True, errors='ignore')
             df.drop('currentDistance', axis=1, inplace=True, errors='ignore')
             df.drop('hadCollision', axis=1, inplace=True, errors='ignore')
```

```
In [17]: pedDidNotCross = df_joined[(df_joined.nextPedRunning == False)]
         firstRowPedDidNotCross = pedDidNotCross.groupby(['session_id', 'scenario', 'pedId'], a
         lastRowPedDidNotCross = pedDidNotCross.groupby(['session_id', 'scenario', 'pedId'], as
         meansPedDidNotCross = pedDidNotCross.groupby(['session_id', 'scenario', 'pedId'], as_i
         varPedDidNotCross = pedDidNotCross.groupby(['session_id', 'scenario', 'pedId'], as_ind
         delete_unused_columns(meansPedDidNotCross)
         delete_unused_columns(varPedDidNotCross)
```

```
In [18]: meansPedDidNotCross.shape

Out[18]: (652, 8)

In [19]: meansPedDidNotCross[meansPedDidNotCross.isnull().values == True]

Out[19]: Empty DataFrame
         Columns: [session_id, scenario, pedId, speed, steering, brake, throttle, acceleration]
         Index: []

In [20]: varPedDidNotCross = varPedDidNotCross.dropna()
         varPedDidNotCross.shape

Out[20]: (648, 8)

In [21]: varPedDidNotCross[varPedDidNotCross.isnull().values == True]

Out[21]: Empty DataFrame
         Columns: [session_id, scenario, pedId, speed, steering, brake, throttle, acceleration]
         Index: []
```

7.4.1 Total time before run

```
In [22]: beforePedCross = firstRowPedDidNotCross[['session_id', 'scenario', 'pedId']].copy()
         firstRowPedDidNotCross['timeFinal'] = lastRowPedDidNotCross['time']
         beforePedCross['total_time_before_run'] = (pd.to_datetime(firstRowPedDidNotCross['time
         beforePedCross.head()
         beforePedCross.shape

Out[22]: (652, 4)

In [23]: beforePedCrossMerge = pd.merge(varPedDidNotCross, meansPedDidNotCross , on=['session_i
         delete_unused_columns(beforePedCrossMerge)
         beforePedCrossMerge.shape

Out[23]: (648, 13)

In [24]: beforePedCrossTotalColumns = pd.merge(beforePedCrossMerge, beforePedCross , on=['sessi
         beforePedCrossTotalColumns.head()

Out[24]:    session_id  scenario  pedId  speed_before_var  steering_before_var  \
         0       144.0       1.0    1.0         13.894033             0.000338
         1       144.0       1.0    3.0          1.185339             0.000238
         2       144.0       1.0   15.0         54.290571             0.001118
         3       144.0       2.0    0.0          4.118861             0.000045
         4       144.0       2.0    2.0          6.272065             0.000233

            brake_before_var  throttle_before_var  acceleration_before_var  \
         0          0.000000             0.017802                 0.016136
         1          0.000000             0.013602                 0.004440
```

```
2          0.009412             0.045813                 0.032457
3          0.020263             0.014640                 0.019610
4          0.000000             0.094616                 0.011947

   speed_before_mean  steering_before_mean  brake_before_mean  \
0           6.614995              0.517642           1.000000
1          12.405217              0.481101           1.000000
2           9.719554              0.496611           0.969863
3           3.134071              0.525620           0.956432
4          15.771787              0.492397           1.000000

   throttle_before_mean  acceleration_before_mean  total_time_before_run
0              0.863178                  0.038096              10.091857
1              0.772151                  0.054448               3.673778
2              0.788247                  0.013562              94.469000
3              0.933556                  0.000010              13.052240
4              0.310690                  0.178377               2.488626
```

In [25]:
```
beforePedCrossTotalColumns.shape
```

Out[25]:
```
(648, 14)
```

7.5 When pedestrian is running

In [26]:
```
pedCross = df_joined[(df_joined.nextPedRunning == True)]
```

In [27]:
```
firstRowPedCross = pedCross.groupby(['session_id', 'scenario', 'pedId'], as_index=Fals
lastRowPedCross = pedCross.groupby(['session_id', 'scenario', 'pedId'], as_index=Fals
meansPedCross = pedCross.groupby(['session_id', 'scenario', 'pedId'], as_index=False)
varPedCross = pedCross.groupby(['session_id', 'scenario', 'pedId'], as_index=False).va
delete_unused_columns(meansPedCross)
delete_unused_columns(varPedCross)
```

7.5.1 Total time when running

In [28]:
```
whenPedCross = firstRowPedCross[['session_id', 'scenario', 'pedId']].copy()
firstRowPedCross['timeFinal'] = lastRowPedCross['time']
whenPedCross['total_time_when_running'] = (pd.to_datetime(firstRowPedCross['timeFinal
whenPedCross.head()
```

Out[28]:
```
   session_id  scenario  pedId  total_time_when_running
0       144.0       1.0    1.0                 5.239119
1       144.0       1.0    3.0                 7.705014
2       144.0       1.0   15.0                 7.848389
3       144.0       2.0    0.0                 1.202199
4       144.0       2.0    2.0                 4.975241
```

In [29]:
```
whenPedCrossMerge = pd.merge(varPedCross, meansPedCross , on=['session_id','scenario'
delete_unused_columns(whenPedCrossMerge)
```

```
whenPedCrossTotalColumns = pd.merge(whenPedCrossMerge, whenPedCross , on=['session_id
whenPedCrossTotalColumns.head()
```

Out[29]:

	session_id	scenario	pedId	speed_when_running_var	steering_when_running_var	brake_when_running_var	throttle_when_running_var	acceleration_when_running_var	speed_when_running_mean	steering_when_running_mean	brake_when_running_mean	throttle_when_running_mean	acceleration_when_running_mean	total_time_when_running
0	144.0	1.0	1.0	11.033679	0.000174	0.035741	0.039745	0.083699	4.593110	0.476982	0.900460	0.739390	0.016767	5.239119
1	144.0	1.0	3.0	29.475829	0.000151	0.078618	0.000856	0.087210	5.276907	0.508684	0.839408	0.929774	-0.065371	7.705014
2	144.0	1.0	15.0	42.756716	0.001915	0.121984	0.035585	0.869046	6.933534	0.473303	0.769213	0.707442	-0.068940	7.848389
3	144.0	2.0	0.0	6.459006	0.000048	0.000000	0.044336	0.023489	6.170808	0.499042	1.000000	0.075305	0.314027	1.202199
4	144.0	2.0	2.0	55.270252	0.000429	0.075733	0.087892	0.212602	9.577210	0.504876	0.836788	0.630319	-0.078850	4.975241

```
In [30]: whenPedCrossTotalColumns.shape
```

Out[30]: (655, 14)

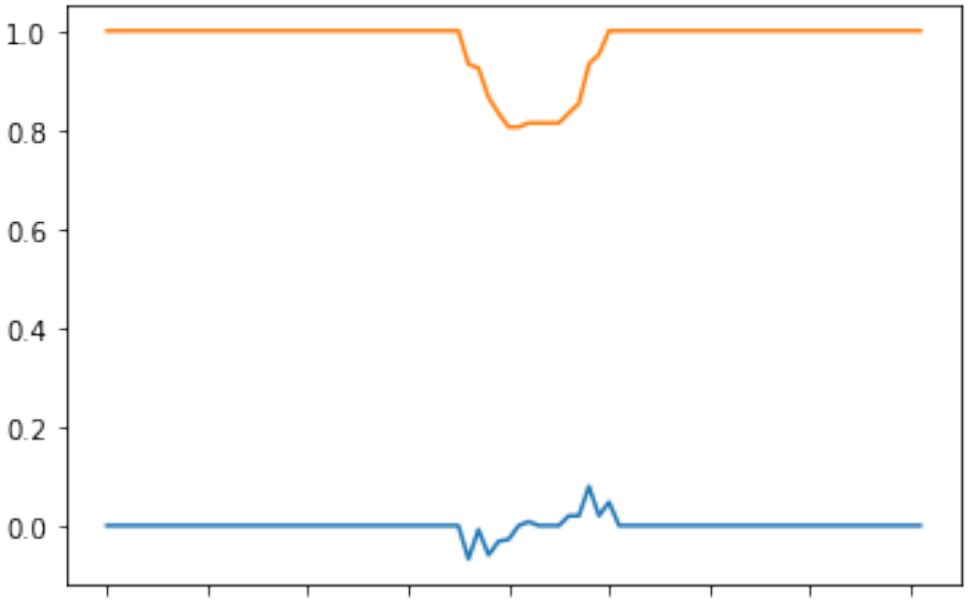
```
In [31]: whenPedCrossTotalColumns.isnull().values.any()
```

```
Out[31]: False
```

7.5.2 Break Reaction Time

```
In [81]: df_joined['brakeDiff'] = df_joined['brake'].diff()
         pedCross = df_joined[(df_joined.nextPedRunning == True)]
         dfNew = pedCross[(pedCross.session_id == 175.0) & (pedCross.scenario == 6.0) & (pedCr
         dfNew['brakeDiff'].plot()
         dfNew['brake'].plot()
```

```
Out[81]: <matplotlib.axes._subplots.AxesSubplot at 0x1112cabe0>
```




```
In [82]: startReactionBrake = pedCross
         finishReactionBrake = pedCross[(pedCross.brakeDiff < 0)]
         finishReactionBrake.shape
```

```
Out[82]: (6536, 19)
```

```
In [83]: firstRowPedCross = startReactionBrake.groupby(['session_id', 'scenario', 'pedId'], as
         firstRowPressBrake = finishReactionBrake.groupby(['session_id', 'scenario', 'pedId'],
         firstRowPressBrake.shape
         firstRowPedCross.shape
```

```
Out[83]: (655, 19)
```

```
In [84]: reactionBrake = pd.merge(firstRowPedCross, firstRowPressBrake, on=['session_id','scena
         reactionBrake.head()
```

Out[84]:

	session_id	scenario	pedId	time_init	Pos_X_init	Pos_Y_init	hadCollision_init	nextPedRunning_init	distancePed_init	speed_init	...	speed_fin	posPedX_fin	posPedY_fin	currentDistance_fin	time.1_fin	steering_fin	brake_fin	throttle_fin	acceleration_fin	brakeDiff_fin
0	144.0	1.0	1.0	2017-10-04 12:50:38.549538	-909.306458	80.473330	0.0	1.0	23.562530	7.754880	...	8.946986	-885.665649	72.027790	1178.73486	2017-10-04 12:50:39.586297	0.477104	0.980163	0.948409	0.072344	-0.019837
1	144.0	1.0	3.0	2017-10-04 12:50:47.530114	-839.965700	83.164460	0.0	1.0	39.895280	13.472353	...	13.034663	-801.989441	93.251990	1094.34827	2017-10-04 12:50:49.648762	0.533089	0.980163	0.948409	-0.009284	-0.019837
2	144.0	1.0	15.0	2017-10-04 12:52:29.782335	-62.199753	40.680523	0.0	1.0	49.374313	25.585112	...	25.720684	-20.775450	15.126913	355.46260	2017-10-04 12:52:29.893130	0.504952	0.900801	0.445304	0.069757	-0.099199
3	144.0	2.0	0.0	2017-10-04 12:57:42.500777	-933.749300	84.487495	0.0	1.0	7.474411	2.517837	...	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
4	144.0	2.0	2.0	2017-10-04 12:57:46.323453	-885.836365	77.199104	0.0	1.0	49.291360	19.412086	...	19.944923	-836.171631	80.316110	1141.20886	2017-10-04 12:57:47.485545	0.480339	0.964294	0.932540	-0.081133	-0.035706

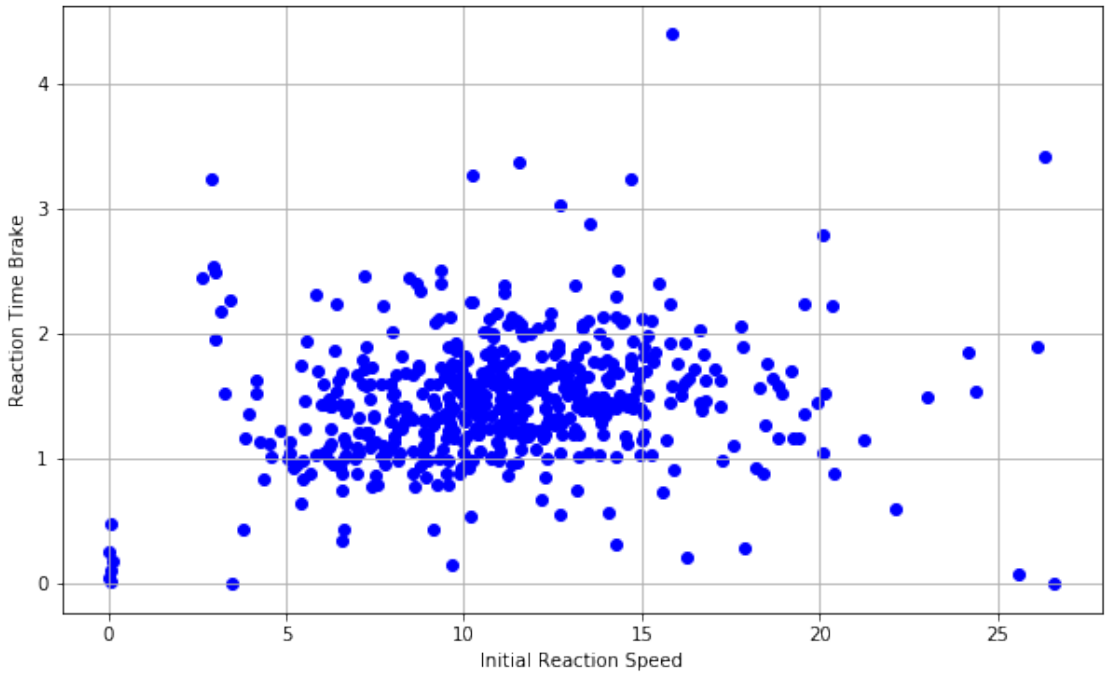
[5 rows x 35 columns]

In [85]:
```
reactionPressBrake = firstRowPedCross[['session_id', 'scenario', 'pedId', 'speed']].co
reactionPressBrake['reaction_time'] = (pd.to_datetime(reactionBrake['time_fin']) - pd
reactionPressBrake = reactionPressBrake.dropna()
reactionPressBrake.head()
```

Out[85]:

	session_id	scenario	pedId	speed	reaction_time
0	144.0	1.0	1.0	7.754880	1.048791
1	144.0	1.0	3.0	13.472353	2.106615
2	144.0	1.0	15.0	25.585112	0.079211
4	144.0	2.0	2.0	19.412086	1.161592
5	144.0	2.0	4.0	18.461056	1.275896

```
In [86]: reactionPressBrake.shape

Out[86]: (573, 5)

In [87]: reactionPressBrake.isnull().values.any()

Out[87]: False

In [88]: plt.figure(figsize=(10, 6))
         plt.xlabel("Initial Reaction Speed")
         plt.ylabel("Reaction Time Brake")
         reactionPressBrake = reactionPressBrake[reactionPressBrake.reaction_time < 50]

         noHits = plt.scatter(reactionPressBrake['speed'], reactionPressBrake['reaction_time']

         plt.grid(True)

         plt.show()
```

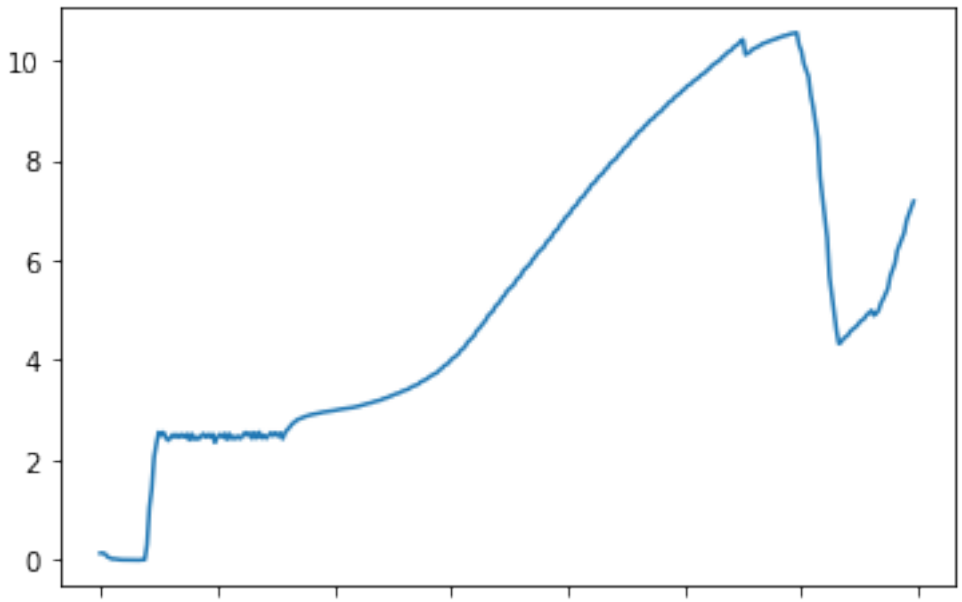



```
In [89]: reactionPressBrake.shape

Out[89]: (559, 5)
```

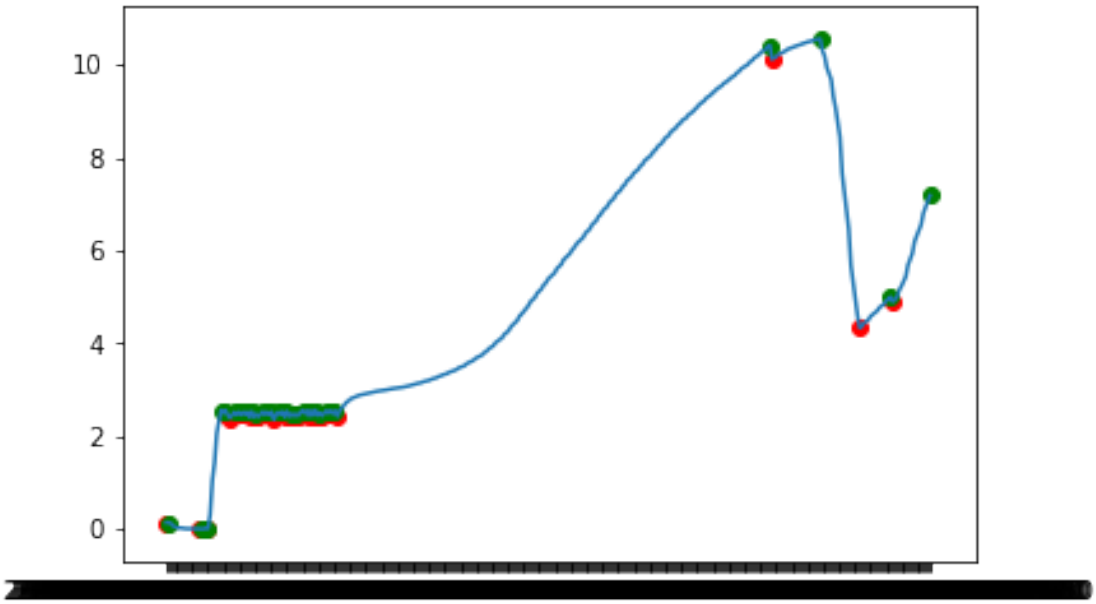
7.6 Agressive driving measure: PKE

In [38]:
```
dfPKE = df_joined.copy()
dfNew = dfPKE[(dfPKE.session_id == 175.0) & (dfPKE.scenario == 6.0) & (dfPKE.pedId ==
dfNew['speed'].plot()
```

Out[38]: <matplotlib.axes._subplots.AxesSubplot at 0x10cdaf2e8>



In [39]:
```
from scipy.signal import argrelextrema

df = dfNew.copy()
n=1 # number of points to be checked before and after
# Find local peaks
df['min'] = df.iloc[argrelextrema(df.speed.values, np.less_equal, order=n)[0]]['speed
df['max'] = df.iloc[argrelextrema(df.speed.values, np.greater_equal, order=n)[0]]['spe

# Plot results
plt.scatter(df.index, df['min'], c='r')
plt.scatter(df.index, df['max'], c='g')
plt.plot(df.index, df['speed'])
plt.show()
```




```
In [40]: df.head()
```

Out[40]:

	time	session_id	scenario	Pos_X	Pos_Y	hadCollision	pedId	nextPedRunning	distancePed	speed	...	posPedY	currentDistance
2017-10-10 12:26:12.200	2017-10-10 12:26:12.192493	175.0	6.0	-975.707031	82.254616	0.0	1.0	0.0	89.97974	0.137321	...	70.21315	1253.46167
2017-10-10 12:26:12.250	2017-10-10 12:26:12.234605	175.0	6.0	-975.710000	82.250580	0.0	1.0	0.0	89.98381	0.140411	...	70.21315	1253.46448
2017-10-10 12:26:12.300	2017-10-10 12:26:12.292259	175.0	6.0	-975.714844	82.245700	0.0	1.0	0.0	89.98826	0.123498	...		
2017-10-10 12:26:12.350	2017-10-10 12:26:12.346905	175.0	6.0	-975.717900	82.242210	0.0	1.0	0.0	89.99129	0.068495	...		
2017-10-10 12:26:12.400	2017-10-10 12:26:12.374478	175.0	6.0	-975.718500	82.241270	0.0	1.0	0.0	89.99179	0.044158	...		

```
2017-10-10 12:26:12.300  70.21315       1253.46863
2017-10-10 12:26:12.350  70.21315       1253.47144
2017-10-10 12:26:12.400  70.21315       1253.47192

                                              time.1  steering  brake  \
2017-10-10 12:26:12.200  2017-10-10 12:26:12.195000  0.499016    1.0
2017-10-10 12:26:12.250  2017-10-10 12:26:12.235107  0.499016    1.0
2017-10-10 12:26:12.300  2017-10-10 12:26:12.295267  0.499016    1.0
2017-10-10 12:26:12.350  2017-10-10 12:26:12.334873  0.499016    1.0
2017-10-10 12:26:12.400  2017-10-10 12:26:12.395032  0.499016    1.0

                         throttle  acceleration  brakeDiff       min       max
2017-10-10 12:26:12.200       1.0     -5.898169        0.0  0.137321       NaN
2017-10-10 12:26:12.250       1.0      0.003090        0.0       NaN  0.140411
2017-10-10 12:26:12.300       1.0     -0.016913        0.0       NaN       NaN
2017-10-10 12:26:12.350       1.0     -0.055003        0.0       NaN       NaN
2017-10-10 12:26:12.400       1.0     -0.024337        0.0       NaN       NaN

[5 rows x 21 columns]
```

In [41]:
```
import math

def aggressiveDriverPKE(df):
    vi = None
    acumulative = 0
    distance = df.currentDistance.max() - df.currentDistance.min()
    df = df[(df['min'].notnull() | df['max'].notnull())]
    for index, row in df.iterrows():
        if (not math.isnan(row['min'])):
            vi = row['min']
        else:
            if (vi is None):
                continue
            else:
                if (not math.isnan(row['max'])):
                    vf = row['max']
                    acumulative += vf*vf - vi * vi
                    vi = None
                    vf = None

    return acumulative/distance
```

In [42]:
```
n=1 # number of points to be checked before and after
# Find local peaks
dfPKE = df_joined.copy()

dfPKE['min'] = dfPKE.iloc[argrelextrema(dfPKE.speed.values, np.less_equal, order=n)[0]
dfPKE['max'] = dfPKE.iloc[argrelextrema(dfPKE.speed.values, np.greater_equal, order=n)
```

```
#print(aggressiveDriverPKE( dfPKE[(dfPKE.session_id == 175.0) & (dfPKE.scenario == 6.
```

In [43]:
```
#dfPKE = dfPKE[['session_id', 'scenario', 'pedId', 'currentDistance', 'min','max']]
#dfPKE = dfPKE[(dfPKE['min'].notnull() | dfPKE['max'].notnull())]
#dfPKE.head()
```

In [44]:
```
dfPKE = dfPKE[['session_id', 'scenario', 'pedId', 'currentDistance', 'min','max']]
total_PKE = dfPKE.groupby(['session_id', 'scenario', 'pedId'], as_index=False).apply(a

total_PKE.head()
```

Out[44]:

	session_id	scenario	pedId	PKE
0	144.0	1.0	1.0	1.932290
1	144.0	1.0	3.0	0.878493
2	144.0	1.0	15.0	2.857169
3	144.0	2.0	0.0	3.659420
4	144.0	2.0	2.0	2.969647

In [45]:
```
total_PKE.shape
```

Out[45]:
```
(671, 4)
```

7.6.1 Steering PKE

In [46]:
```
dfSteeringPKE = df_joined.copy()
dfSteeringPKE['min'] = dfSteeringPKE.iloc[argrelextrema(dfSteeringPKE.speed.values, np
dfSteeringPKE['max'] = dfSteeringPKE.iloc[argrelextrema(dfSteeringPKE.speed.values, np
```

In [47]:
```
dfSteeringPKE = dfSteeringPKE[['session_id', 'scenario', 'pedId', 'currentDistance',
total_steering_PKE = dfSteeringPKE.groupby(['session_id', 'scenario', 'pedId'], as_inc

total_steering_PKE.head()
```

Out[47]:

	session_id	scenario	pedId	PKE_Steering
0	144.0	1.0	1.0	-0.000150
1	144.0	1.0	3.0	0.000274
2	144.0	1.0	15.0	0.000108
3	144.0	2.0	0.0	-0.000627
4	144.0	2.0	2.0	-0.000258

In [48]:
```
total_steering_PKE.shape
```

Out[48]:
```
(671, 4)
```

7.7 Backup PKE's

In [50]:
```
#total_PKE.to_csv('pke.csv', index=False)
#total_steering_PKE.to_csv('steering_pke.csv', index=False)

total_PKE = pd.read_csv("pke.csv")
total_steering_PKE = pd.read_csv("steering_pke.csv")
```

8 Merge dataframes

In [90]:
```
df_suffix = pd.merge(means_cleaned, variance_cleaned, on=['session_id','scenario','pe
df_suffix = pd.merge(df_suffix, total_time, on=['session_id','scenario','pedId'],how=
df_suffix = pd.merge(df_suffix, initial_distance, on=['session_id','scenario','pedId'
df_suffix = pd.merge(df_suffix, max_speed, on=['session_id','scenario','pedId'],how='
df_suffix = pd.merge(df_suffix, total_PKE, on=['session_id','scenario','pedId'],how='
df_suffix = pd.merge(df_suffix, total_steering_PKE, on=['session_id','scenario','pedI
df_suffix = pd.merge(df_suffix, reactionPressBrake, on=['session_id','scenario','pedI
df_suffix.head()
```

Out[90]:

	session_id	scenario	pedId	hadCollision	speed_total_mean	steering_total_mean	brake_total_mean	throttle_total_mean	acceleration_total_mean	speed_total_var	...	brake_total_var	throttle_total_var	acceleration_total_var	total_time	distancePed	speed	PKE	PKE_Steering	speed_react	reaction_time
0	144.0	1.0	1.0	0.000000	5.919151	0.503649	0.965743	0.820576	0.030731	13.796202	...	0.014468	0.028719	0.039370	15.405173	89.992450	11.669766	1.932290	-0.000150	7.754880	1.048791
1	144.0	1.0	3.0	0.000000	7.580378	0.499771	0.891302	0.878839	-0.026652	31.451253	...	0.058767	0.010391	0.063480	11.412381	85.063860	13.499710	0.878493	0.000274	13.472353	2.106615
2	144.0	1.0	15.0	0.008366	9.474048	0.494557	0.952182	0.781126	0.006292	53.873833	...	0.022506	0.045416	0.106281	102.356492	789.212800	25.851397	2.857169	0.000108	25.585112	0.079211
3	144.0	2.0	2.0	0.000000	11.669419	0.500661	0.891913	0.522365	0.008028	47.209285	...	0.055982	0.112551	0.159198	7.505478	88.011610	20.055070	2.969647	-0.000258	19.412086	1.161592
4	144.0	2.0	4.0	0.000000	12.187044	0.499769	0.861132	0.558120	0.001881	42.031423	...	0.102442	0.079023	0.158822	8.681609	105.973686	19.697004	4.033468	0.000066	18.461056	1.275896

[5 rows x 21 columns]

In [91]:
```
def delete_common_columns(df):
```

```
            df.drop('session_id', axis=1, inplace=True, errors='ignore')
            df.drop('pedId', axis=1, inplace=True, errors='ignore')

        delete_common_columns(df_suffix)
        df_scenario_cluster = df_suffix.copy()
        df_suffix.drop('scenario', axis=1, inplace=True, errors='ignore')
        df_suffix.rename(columns={'speed':'max_speed'}, inplace=True)
```

```
In [92]: df_suffix.isnull().values.any()
```

```
Out[92]: False
```

```
In [93]: df_suffix.shape
```

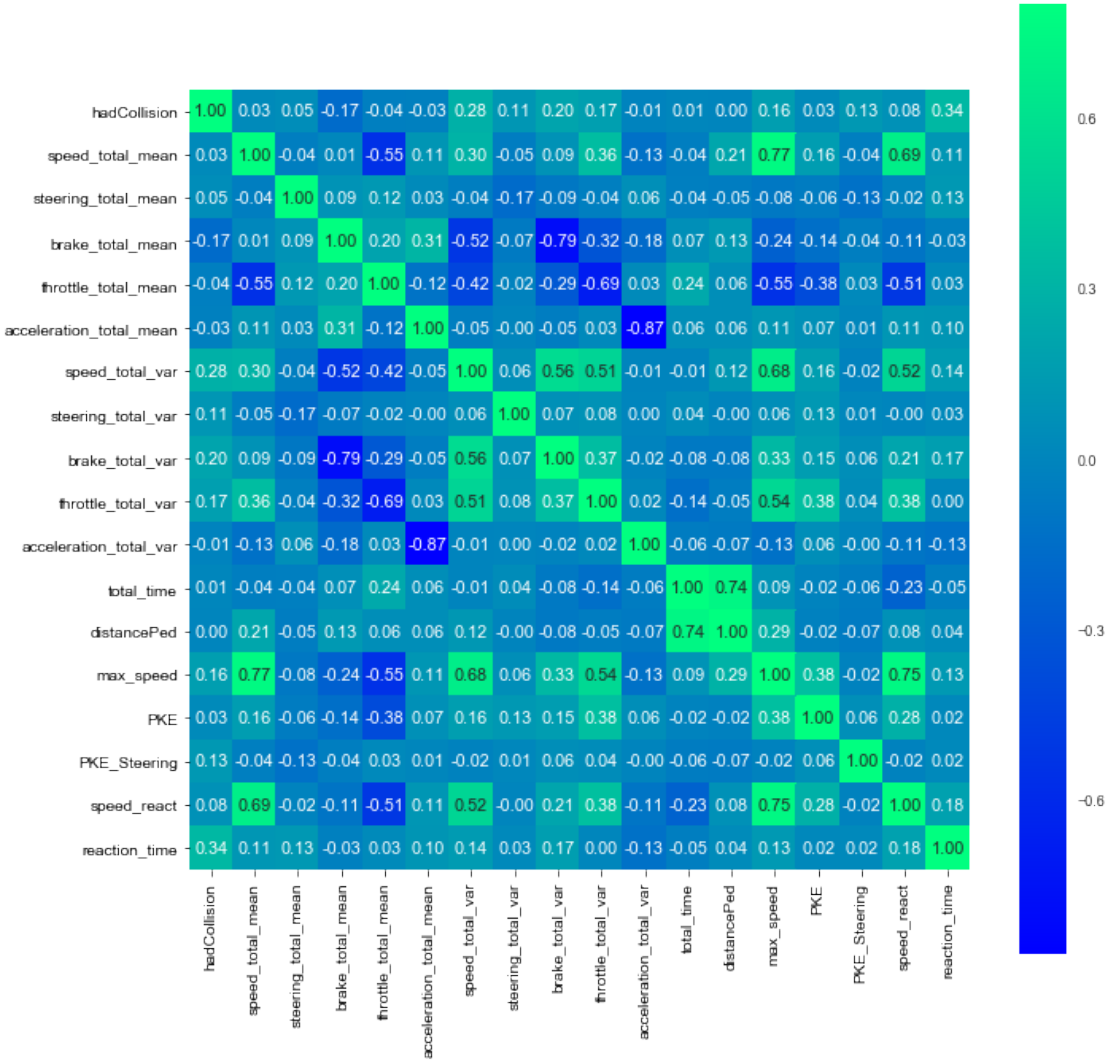
```
Out[93]: (559, 18)
```

```
In [94]: reactionPressBrake.shape
```

```
Out[94]: (559, 5)
```

```
In [95]: def plot_corr(df, size=11):
             """
             Function plots a graphical correlation matrix for each pair of columns in the dat

             Input:
                 df: pandas DataFrame
                 size: vertical and horizontal size of the plot

             Displays:
                 matrix of correlation between columns.  Blue-cyan-yellow-red-darkred => less
                                                         0 ------------------>  1
                                                         Expect a darkred line running from to
             """

             corr = df.corr()    # data frame correlation function
             fig, ax = plt.subplots(figsize=(size, size))
             ax.matshow(corr)   # color code the rectangles by correlation value
             plt.xticks(range(len(corr.columns)), corr.columns)  # draw x tick marks
             plt.yticks(range(len(corr.columns)), corr.columns)  # draw y tick marks
```

```
In [96]: import seaborn as sns

         corrmat = df_suffix.corr()
         f, ax = plt.subplots(figsize=(11,11))
         sns.set(font_scale=0.9)
         sns.heatmap(corrmat, vmax=.8, square=True, annot=True, fmt='.2f', cmap="winter")
         plt.show()
```



In [82]: `df_suffix.corr()`

Out[82]:

	hadCollision	speed_total_mean	steering_total_mean	\
hadCollision	1.000000	0.003023	0.037162	
speed_total_mean	0.003023	1.000000	-0.073901	
steering_total_mean	0.037162	-0.073901	1.000000	
brake_total_mean	-0.103735	0.057452	0.073249	
throttle_total_mean	0.000472	-0.691672	0.129062	
acceleration_total_mean	0.011156	0.320063	0.052155	
speed_total_var	0.212230	0.236173	-0.036002	
steering_total_var	0.147238	0.001821	-0.178203	
brake_total_var	0.143651	0.038165	-0.075838	
throttle_total_var	0.113680	0.382434	-0.047182	
acceleration_total_var	-0.025400	-0.258142	-0.022899	
total_time	0.031473	-0.056937	-0.022556	

distancePed	0.032134	0.163800	-0.071290
max_speed	0.097091	0.821473	-0.080720
PKE	0.027530	0.153890	-0.038472
PKE_Steering	0.107471	0.007824	-0.206095
speed_react	0.077155	0.690015	-0.016004
reaction_time	0.335029	0.113742	0.126902

	brake_total_mean	throttle_total_mean	\
hadCollision	-0.103735	0.000472	
speed_total_mean	0.057452	-0.691672	
steering_total_mean	0.073249	0.129062	
brake_total_mean	1.000000	0.083151	
throttle_total_mean	0.083151	1.000000	
acceleration_total_mean	0.111709	-0.262172	
speed_total_var	-0.464997	-0.339819	
steering_total_var	-0.039894	-0.039581	
brake_total_var	-0.731555	-0.197457	
throttle_total_var	-0.239204	-0.620655	
acceleration_total_var	-0.085719	0.160891	
total_time	0.064027	0.222875	
distancePed	0.092040	0.040787	
max_speed	-0.148294	-0.652242	
PKE	-0.144054	-0.354183	
PKE_Steering	-0.034307	-0.026244	
speed_react	-0.113997	-0.509799	
reaction_time	-0.032702	0.034894	

	acceleration_total_mean	speed_total_var	\
hadCollision	0.011156	0.212230	
speed_total_mean	0.320063	0.236173	
steering_total_mean	0.052155	-0.036002	
brake_total_mean	0.111709	-0.464997	
throttle_total_mean	-0.262172	-0.339819	
acceleration_total_mean	1.000000	0.116663	
speed_total_var	0.116663	1.000000	
steering_total_var	0.044029	0.088946	
brake_total_var	0.047181	0.556514	
throttle_total_var	0.128485	0.490132	
acceleration_total_var	-0.932188	-0.118702	
total_time	0.084147	0.012359	
distancePed	0.101619	0.165060	
max_speed	0.404565	0.602346	
PKE	0.135396	0.215235	
PKE_Steering	0.006508	-0.006044	
speed_react	0.113250	0.523054	
reaction_time	0.101548	0.143505	

	steering_total_var	brake_total_var	\

hadCollision	0.147238	0.143651
speed_total_mean	0.001821	0.038165
steering_total_mean	-0.178203	-0.075838
brake_total_mean	-0.039894	-0.731555
throttle_total_mean	-0.039581	-0.197457
acceleration_total_mean	0.044029	0.047181
speed_total_var	0.088946	0.556514
steering_total_var	1.000000	0.045281
brake_total_var	0.045281	1.000000
throttle_total_var	0.105340	0.318190
acceleration_total_var	-0.036567	-0.072101
total_time	0.053497	-0.075732
distancePed	0.071798	-0.049071
max_speed	0.097545	0.267796
PKE	0.133867	0.171541
PKE_Steering	-0.011529	0.050653
speed_react	-0.003533	0.207386
reaction_time	0.025953	0.165380

	throttle_total_var	acceleration_total_var	\
hadCollision	0.113680	-0.025400	
speed_total_mean	0.382434	-0.258142	
steering_total_mean	-0.047182	-0.022899	
brake_total_mean	-0.239204	-0.085719	
throttle_total_mean	-0.620655	0.160891	
acceleration_total_mean	0.128485	-0.932188	
speed_total_var	0.490132	-0.118702	
steering_total_var	0.105340	-0.036567	
brake_total_var	0.318190	-0.072101	
throttle_total_var	1.000000	-0.082509	
acceleration_total_var	-0.082509	1.000000	
total_time	-0.115168	-0.089473	
distancePed	-0.022477	-0.101559	
max_speed	0.533691	-0.339401	
PKE	0.374110	-0.055015	
PKE_Steering	0.051558	-0.002641	
speed_react	0.384503	-0.105586	
reaction_time	0.004183	-0.130727	

	total_time	distancePed	max_speed	PKE	\
hadCollision	0.031473	0.032134	0.097091	0.027530	
speed_total_mean	-0.056937	0.163800	0.821473	0.153890	
steering_total_mean	-0.022556	-0.071290	-0.080720	-0.038472	
brake_total_mean	0.064027	0.092040	-0.148294	-0.144054	
throttle_total_mean	0.222875	0.040787	-0.652242	-0.354183	
acceleration_total_mean	0.084147	0.101619	0.404565	0.135396	
speed_total_var	0.012359	0.165060	0.602346	0.215235	
steering_total_var	0.053497	0.071798	0.097545	0.133867	

```
brake_total_var           -0.075732    -0.049071    0.267796  0.171541
throttle_total_var        -0.115168    -0.022477    0.533691  0.374110
acceleration_total_var    -0.089473    -0.101559   -0.339401 -0.055015
total_time                 1.000000     0.491870    0.063446 -0.118974
distancePed                0.491870     1.000000    0.277116  0.019337
max_speed                  0.063446     0.277116    1.000000  0.363375
PKE                       -0.118974     0.019337    0.363375  1.000000
PKE_Steering              -0.037848    -0.049021    0.001949  0.079838
speed_react               -0.229716     0.075097    0.749418  0.280204
reaction_time             -0.047035     0.038478    0.128913  0.019924

                          PKE_Steering  speed_react  reaction_time
hadCollision                  0.107471     0.077155       0.335029
speed_total_mean              0.007824     0.690015       0.113742
steering_total_mean          -0.206095    -0.016004       0.126902
brake_total_mean             -0.034307    -0.113997      -0.032702
throttle_total_mean          -0.026244    -0.509799       0.034894
acceleration_total_mean       0.006508     0.113250       0.101548
speed_total_var              -0.006044     0.523054       0.143505
steering_total_var           -0.011529    -0.003533       0.025953
brake_total_var               0.050653     0.207386       0.165380
throttle_total_var            0.051558     0.384503       0.004183
acceleration_total_var       -0.002641    -0.105586      -0.130727
total_time                   -0.037848    -0.229716      -0.047035
distancePed                  -0.049021     0.075097       0.038478
max_speed                     0.001949     0.749418       0.128913
PKE                           0.079838     0.280204       0.019924
PKE_Steering                  1.000000    -0.021700       0.024101
speed_react                  -0.021700     1.000000       0.183851
reaction_time                 0.024101     0.183851       1.000000
```

8.1 Mold Data

8.1.1 Data Types

Inspect data types to see if there are any issues. Data should be numeric.

```
In [97]: df = df_suffix
         #hadCollision_map = {True : 1, False : 0}
         df['hadCollision'] = np.where(df['hadCollision'] == 0, 0, 1)

In [98]: cols_at_end = ['hadCollision']
         df = df[[c for c in df if c not in cols_at_end]
                 + [c for c in cols_at_end if c in df]]
         df.head(5)

Out[98]:    speed_total_mean  steering_total_mean  brake_total_mean  \
         0          5.919151             0.503649          0.965743
         1          7.580378             0.499771          0.891302
```

```
2          9.474048             0.494557          0.952182
3         11.669419             0.500661          0.891913
4         12.187044             0.499769          0.861132

   throttle_total_mean  acceleration_total_mean  speed_total_var  \
0             0.820576                 0.030731        13.796202
1             0.878839                -0.026652        31.451253
2             0.781126                 0.006292        53.873833
3             0.522365                 0.008028        47.209285
4             0.558120                 0.001881        42.031423

   steering_total_var  brake_total_var  throttle_total_var  \
0            0.000655         0.014468            0.028719
1            0.000345         0.058767            0.010391
2            0.001231         0.022506            0.045416
3            0.000396         0.055982            0.112551
4            0.000430         0.102442            0.079023

   acceleration_total_var  total_time  distancePed  max_speed       PKE  \
0                0.039370   15.405173    89.992450  11.669766  1.932290
1                0.063480   11.412381    85.063860  13.499710  0.878493
2                0.106281  102.356492   789.212800  25.851397  2.857169
3                0.159198    7.505478    88.011610  20.055070  2.969647
4                0.158822    8.681609   105.973686  19.697004  4.033468

   PKE_Steering  speed_react  reaction_time  hadCollision
0     -0.000150     7.754880       1.048791             0
1      0.000274    13.472353       2.106615             0
2      0.000108    25.585112       0.079211             1
3     -0.000258    19.412086       1.161592             0
4      0.000066    18.461056       1.275896             0
```

8.1.2 Backup merge data

```
In [99]: #df.to_csv('mergeData_v5.csv', index=False)
```

8.1.3 Distribution

```
In [86]: num_obs = len(df)
         num_true = len(df.loc[df['hadCollision'] == 1])
         num_false = len(df.loc[df['hadCollision'] == 0])
         print("Number of True cases:  {0} ({1:2.2f}%)".format(num_true, (num_true/num_obs) * 1
         print("Number of False cases: {0} ({1:2.2f}%)".format(num_false, (num_false/num_obs) *

Number of True cases:  61 (9.31%)
Number of False cases: 594 (90.69%)
```

8.1.4 Spliting the data

70% for training, 30% for testing

```
In [87]: from sklearn.model_selection import train_test_split

         predicted_column = ['hadCollision']

         feature_col_names = [c for c in df if c not in predicted_column]
         predicted_class_names = predicted_column

         X = df[feature_col_names].values     # predictor feature columns (10 X m)
         y = df[predicted_class_names].values # predicted class (1=true, 0=false) column (1 X
         split_test_size = 0.30

         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=split_test_size,
                                     # test_size = 0.3 is 30%, 42 is the answer to everything
```

We check to ensure we have the the desired 70% train, 30% test split of the data

```
In [88]: print("{0:0.2f}% in training set".format((len(X_train)/len(df.index)) * 100))
         print("{0:0.2f}% in test set".format((len(X_test)/len(df.index)) * 100))
```

```
69.92% in training set
30.08% in test set
```

Verifying predicted value was split correctly

```
In [89]: print("Original True  : {0} ({1:0.2f}%)".format(len(df.loc[df['hadCollision'] == 1]),
         print("Original False : {0} ({1:0.2f}%)".format(len(df.loc[df['hadCollision'] == 0]),
         print("")
         print("Training True  : {0} ({1:0.2f}%)".format(len(y_train[y_train[:] == 1]), (len(y
         print("Training False : {0} ({1:0.2f}%)".format(len(y_train[y_train[:] == 0]), (len(y
         print("")
         print("Test True      : {0} ({1:0.2f}%)".format(len(y_test[y_test[:] == 1]), (len(y_te
         print("Test False     : {0} ({1:0.2f}%)".format(len(y_test[y_test[:] == 0]), (len(y_te
```

```
Original True  : 61 (9.31%)
Original False : 594 (90.69%)

Training True  : 48 (10.48%)
Training False : 410 (89.52%)

Test True      : 13 (6.60%)
Test False     : 184 (93.40%)
```

8.1.5 Post-split Data Preparation

```
In [90]: df.head()
```

Out[90]:

	speed_total_mean	steering_total_mean	brake_total_mean	throttle_total_mean	acceleration_total_mean	speed_total_var	steering_total_var	brake_total_var	throttle_total_var	acceleration_total_var	total_time	distancePed	max_speed	PKE	PKE_Steering	speed_react	reaction_time	hadCollision
0	5.919151	0.503649	0.965743	0.820576	0.030731	13.796202	0.000655	0.014468	0.028719	0.039370	15.405173	89.992450	11.669766	1.932290	-0.000150	7.754880	1.048791	0
1	7.580378	0.499771	0.891302	0.878839	-0.026652	31.451253	0.000345	0.058767	0.010391	0.063480	11.412381	85.063860	13.499710	0.878493	0.000274	13.472353	2.106615	0
2	9.474048	0.494557	0.952182	0.781126	0.006292	53.873833	0.001231	0.022506	0.045416	0.106281	102.356492	789.212800	25.851397	2.857169	0.000108	25.585112	0.079211	1
3	3.398595	0.523305	0.960227	0.858795	0.027363	5.036717	0.000101	0.018643	0.075860	0.027735	14.315100	47.977978	10.266865	3.659420	-0.000627	NaN	NaN	0
4	11.669419	0.500661	0.891913	0.522365	0.008028	47.209285	0.000396	0.055982	0.112551	0.159198	7.505478	88.011610	20.055070	2.969647	-0.000258	19.412086	1.161592	0

9 Training Initial Algorithm

9.1 Naive Bayes

```
In [91]: from sklearn.naive_bayes import GaussianNB

         # create Gaussian Naive Bayes model object and train it with the data
         nb_model = GaussianNB()
```

```
 nb_model.fit(X_train, y_train.ravel())
```

```
---------------------------------------------------------------------------

ValueError                                Traceback (most recent call last)

<ipython-input-91-9ac5589a2207> in <module>()
  4 nb_model = GaussianNB()
  5
----> 6 nb_model.fit(X_train, y_train.ravel())


/anaconda3/lib/python3.6/site-packages/sklearn/naive_bayes.py in fit(self, X, y, sample
181             Returns self.
182         """
--> 183         X, y = check_X_y(X, y)
184         return self._partial_fit(X, y, np.unique(y), _refit=True,
185                                  sample_weight=sample_weight)


/anaconda3/lib/python3.6/site-packages/sklearn/utils/validation.py in check_X_y(X, y, a
571     X = check_array(X, accept_sparse, dtype, order, copy, force_all_finite,
572                     ensure_2d, allow_nd, ensure_min_samples,
--> 573                     ensure_min_features, warn_on_dtype, estimator)
574     if multi_output:
575         y = check_array(y, 'csr', force_all_finite=True, ensure_2d=False,


/anaconda3/lib/python3.6/site-packages/sklearn/utils/validation.py in check_array(array
451                              % (array.ndim, estimator_name))
452         if force_all_finite:
--> 453             _assert_all_finite(array)
454
455     shape_repr = _shape_repr(array.shape)


/anaconda3/lib/python3.6/site-packages/sklearn/utils/validation.py in _assert_all_finit
 42             and not np.isfinite(X).all()):
 43         raise ValueError("Input contains NaN, infinity"
---> 44                          " or a value too large for %r." % X.dtype)
 45
 46


ValueError: Input contains NaN, infinity or a value too large for dtype('float64').
```

9.1.1 Performance on Training Data

```
In [150]: # predict values using the training data
          nb_predict_train = nb_model.predict(X_train)

          # import the performance metrics library
          from sklearn import metrics

          # Accuracy
          print("Accuracy: {0:.4f}".format(metrics.accuracy_score(y_train, nb_predict_train)))
          print()
```

```
Accuracy: 0.6004
```

9.1.2 Performance on Testing Data

```
In [151]: # predict values using the testing data
          nb_predict_test = nb_model.predict(X_test)

          from sklearn import metrics

          # training metrics
          print("nb_predict_test", nb_predict_test)
          #print ("y_test", y_test)
          print("Accuracy: {0:.4f}".format(metrics.accuracy_score(y_test, nb_predict_test)))
```

```
nb_predict_test [0 0 0 0 1 0 1 1 0 0 1 1 0 1 1 0 0 1 0 0 1 1 0 1 0 0 1 0 0 0 0 1 0 0 0 0 0
 0 1 0 0 0 0 0 0 1 1 0 0 1 0 0 1 1 0 1 0 0 1 0 0 0 0 1 0 0 1 0 1 1 0 0 0 1
 0 0 1 1 1 0 1 0 0 1 0 0 0 0 0 1 0 1 0 1 0 1 0 0 1 0 0 0 1 0 0 0 0 0 0 1 1
 1 1 0 1 1 1 0 1 1 1 1 0 0 0 0 0 1 0 0 0 0 0 1 0 1 1 0 0 0 1 1 1 0 0 0 0 0
 1 1 0 0 1 0 0 1 1 0 1 0 0 1 1 1 0 1 1 0 1 1 1 0 1 1 1 1 1 0 0 1 0 0 0 1 1
 0 0 0 1 0 1 0 0 0 0 0 1]
Accuracy: 0.6091
```

Metrics

```
In [152]: print("Confusion Matrix")
          print("{0}".format(metrics.confusion_matrix(y_test, nb_predict_test)))
          print("")

          print("Classification Report")
          print(metrics.classification_report(y_test, nb_predict_test))
```

```
Confusion Matrix
[[112  72]
 [  5   8]]
```

```
Classification Report
             precision    recall  f1-score   support

          0       0.96      0.61      0.74       184
          1       0.10      0.62      0.17        13

avg / total       0.90      0.61      0.71       197
```

9.2 Random Forest

```
In [153]: from sklearn.ensemble import RandomForestClassifier
          rf_model = RandomForestClassifier(random_state=42, n_estimators=10)      # Create ra
          rf_model.fit(X_train, y_train.ravel())
```

```
Out[153]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                      max_depth=None, max_features='auto', max_leaf_nodes=None,
                      min_impurity_decrease=0.0, min_impurity_split=None,
                      min_samples_leaf=1, min_samples_split=2,
                      min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=1,
                      oob_score=False, random_state=42, verbose=0, warm_start=False)
```

9.2.1 Predict Training Data

```
In [154]: rf_predict_train = rf_model.predict(X_train)
          # training metrics
          print("Accuracy: {0:.4f}".format(metrics.accuracy_score(y_train, rf_predict_train)))
```

```
Accuracy: 0.9869
```

9.2.2 Predict Test Data

```
In [155]: rf_predict_test = rf_model.predict(X_test)

          # training metrics
          print("Accuracy: {0:.4f}".format(metrics.accuracy_score(y_test, rf_predict_test)))
```

```
Accuracy: 0.9289
```

```
In [156]: print(metrics.confusion_matrix(y_test, rf_predict_test) )
          print("")
          print("Classification Report")
          print(metrics.classification_report(y_test, rf_predict_test))

          ## TN  FP
          ## FN  TP
```

```
[[183   1]
 [ 13   0]]

Classification Report
             precision    recall  f1-score   support

          0       0.93      0.99      0.96       184
          1       0.00      0.00      0.00        13

avg / total       0.87      0.93      0.90       197
```

9.3 Logistic Regression

```
In [157]: from sklearn.linear_model import LogisticRegression

          #lr_model =LogisticRegression(C=0.7, random_state=42, solver='liblinear', max_iter=1
          lr_model =LogisticRegression(C=0.7, random_state=42)
          lr_model.fit(X_train, y_train.ravel())
          lr_predict_test = lr_model.predict(X_test)

          # training metrics
          print("Accuracy: {0:.4f}".format(metrics.accuracy_score(y_test, lr_predict_test)))
          print(metrics.confusion_matrix(y_test, lr_predict_test) )
          print("")
          print("Classification Report")
          print(metrics.classification_report(y_test, lr_predict_test))
```

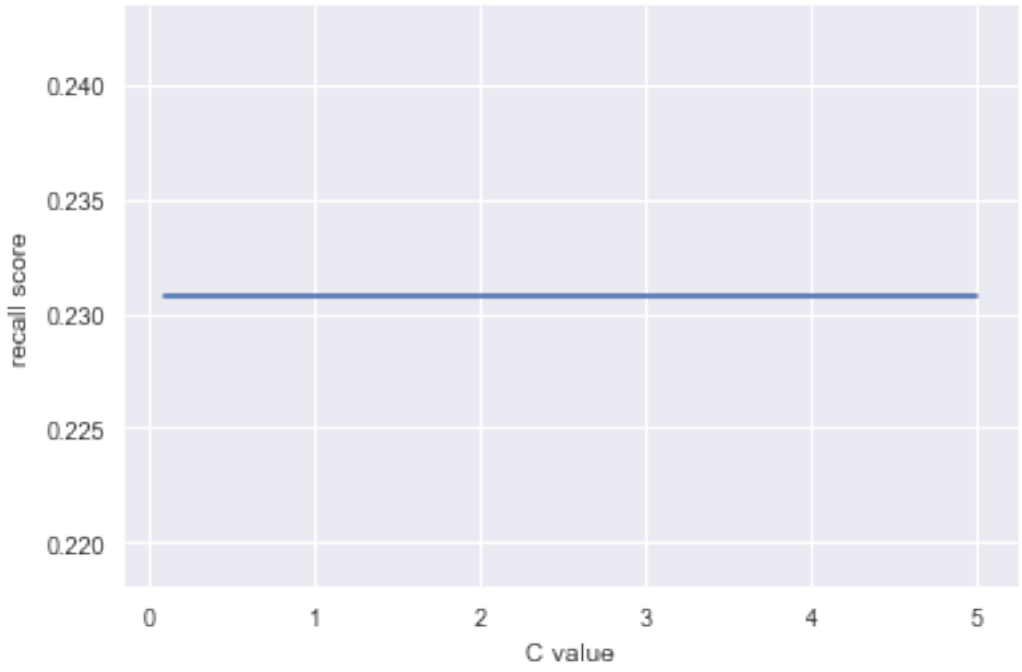
```
Accuracy: 0.9442
[[183   1]
 [ 10   3]]

Classification Report
             precision    recall  f1-score   support

          0       0.95      0.99      0.97       184
          1       0.75      0.23      0.35        13

avg / total       0.94      0.94      0.93       197
```

```
In [158]: C_start = 0.1
          C_end = 5
          C_inc = 0.1

          C_values, recall_scores = [], []
```

```
C_val = C_start
best_recall_score = 0
while (C_val < C_end):
    C_values.append(C_val)
    lr_model_loop = LogisticRegression(C=C_val, random_state=42, solver='liblinear')
    lr_model_loop.fit(X_train, y_train.ravel())
    lr_predict_loop_test = lr_model_loop.predict(X_test)
    recall_score = metrics.recall_score(y_test, lr_predict_loop_test)
    recall_scores.append(recall_score)
    if (recall_score > best_recall_score):
        best_recall_score = recall_score
        best_lr_predict_test = lr_predict_loop_test

    C_val = C_val + C_inc

best_score_C_val = C_values[recall_scores.index(best_recall_score)]
print("1st max value of {0:.3f} occured at C={1:.3f}".format(best_recall_score, best_

%matplotlib inline
plt.plot(C_values, recall_scores, "-")
plt.xlabel("C value")
plt.ylabel("recall score")
```

```
1st max value of 0.231 occured at C=0.100
```

Out[158]: Text(0,0.5,'recall score')

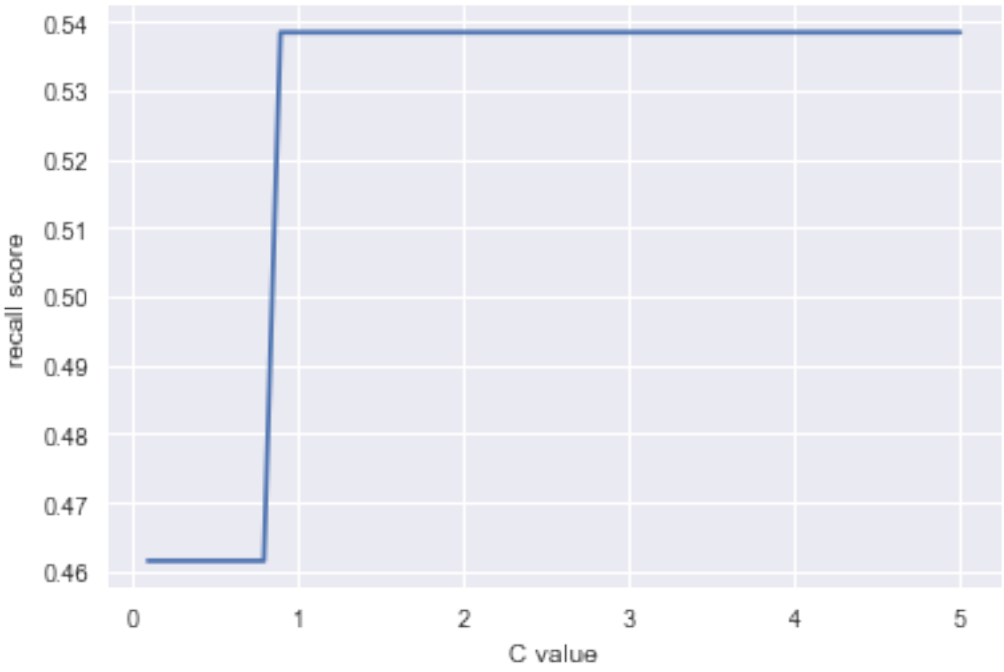


9.3.1 Logisitic regression with class_weight='balanced'

```
In [159]: C_start = 0.1
          C_end = 5
          C_inc = 0.1

          C_values, recall_scores = [], []

          C_val = C_start
          best_recall_score = 0
          while (C_val < C_end):
              C_values.append(C_val)
              lr_model_loop = LogisticRegression(C=C_val, class_weight="balanced", random_state
              lr_model_loop.fit(X_train, y_train.ravel())
              lr_predict_loop_test = lr_model_loop.predict(X_test)
              recall_score = metrics.recall_score(y_test, lr_predict_loop_test)
              recall_scores.append(recall_score)
              if (recall_score > best_recall_score):
                  best_recall_score = recall_score
                  best_lr_predict_test = lr_predict_loop_test

              C_val = C_val + C_inc

          best_score_C_val = C_values[recall_scores.index(best_recall_score)]
          print("1st max value of {0:.3f} occured at C={1:.3f}".format(best_recall_score, best_

          %matplotlib inline
          plt.plot(C_values, recall_scores, "-")
          plt.xlabel("C value")
          plt.ylabel("recall score")
```

```
1st max value of 0.538 occured at C=0.900
```

```
Out[159]: Text(0,0.5,'recall score')
```




```
In [160]: from sklearn.linear_model import LogisticRegression
          lr_model =LogisticRegression( class_weight="balanced", C=best_score_C_val, random_sta
          lr_model.fit(X_train, y_train.ravel())
          lr_predict_test = lr_model.predict(X_test)

          # training metrics
          print("Accuracy: {0:.4f}".format(metrics.accuracy_score(y_test, lr_predict_test)))
          print(metrics.confusion_matrix(y_test, lr_predict_test) )
          print("")
          print("Classification Report")
          print(metrics.classification_report(y_test, lr_predict_test))
          print(metrics.recall_score(y_test, lr_predict_test))
```

```
Accuracy: 0.6853
[[128  56]
 [  6   7]]

Classification Report
             precision    recall  f1-score   support

          0       0.96      0.70      0.81       184
          1       0.11      0.54      0.18        13

avg / total       0.90      0.69      0.76       197
```

```
0.5384615384615384
```

9.3.2 LogisticRegressionCV

```
In [161]: from sklearn.linear_model import LogisticRegressionCV
          lr_cv_model = LogisticRegressionCV(n_jobs=-1, random_state=42, Cs=3, cv=10, refit=Fal
          lr_cv_model.fit(X_train, y_train.ravel())
```

```
Out[161]: LogisticRegressionCV(Cs=3, class_weight='balanced', cv=10, dual=False,
                    fit_intercept=True, intercept_scaling=1.0, max_iter=500,
                    multi_class='ovr', n_jobs=-1, penalty='l2', random_state=42,
                    refit=False, scoring=None, solver='lbfgs', tol=0.0001,
                    verbose=0)
```

9.3.3 Predict on Test data

```
In [162]: lr_cv_predict_test = lr_cv_model.predict(X_test)

          # training metrics
          print("Accuracy: {0:.4f}".format(metrics.accuracy_score(y_test, lr_cv_predict_test)))
          print(metrics.confusion_matrix(y_test, lr_cv_predict_test) )
          print("")
          print("Classification Report")
          print(metrics.classification_report(y_test, lr_cv_predict_test))
```

```
Accuracy: 0.8071
[[152  32]
 [  6   7]]

Classification Report
             precision    recall  f1-score   support

          0       0.96      0.83      0.89       184
          1       0.18      0.54      0.27        13

avg / total       0.91      0.81      0.85       197
```

9.3.4 Linear SVC

```
In [163]: from sklearn.model_selection import train_test_split

          data = df.copy()
          X = data.drop('hadCollision', axis=1)
          Y = data['hadCollision']

          from sklearn import preprocessing
```

```
X = preprocessing.scale(X)


X_train, x_test, Y_train, y_test = train_test_split(X, Y, test_size=0.2, random_state
```

In [164]:
```
from sklearn.svm import LinearSVC

clf_svc = LinearSVC(penalty='l1', dual=False, tol=1e-3)
clf_svc.fit(X_train, Y_train)
```

Out[164]:
```
LinearSVC(C=1.0, class_weight=None, dual=False, fit_intercept=True,
     intercept_scaling=1, loss='squared_hinge', max_iter=1000,
     multi_class='ovr', penalty='l1', random_state=None, tol=0.001,
     verbose=0)
```

In [165]:
```
print(clf_svc.score(x_test, y_test))
```

```
0.916030534351145
```

In [166]:
```
from sklearn.decomposition import PCA

pca = PCA(n_components=5, whiten=True)
X_reduced = pca.fit_transform(X)
```

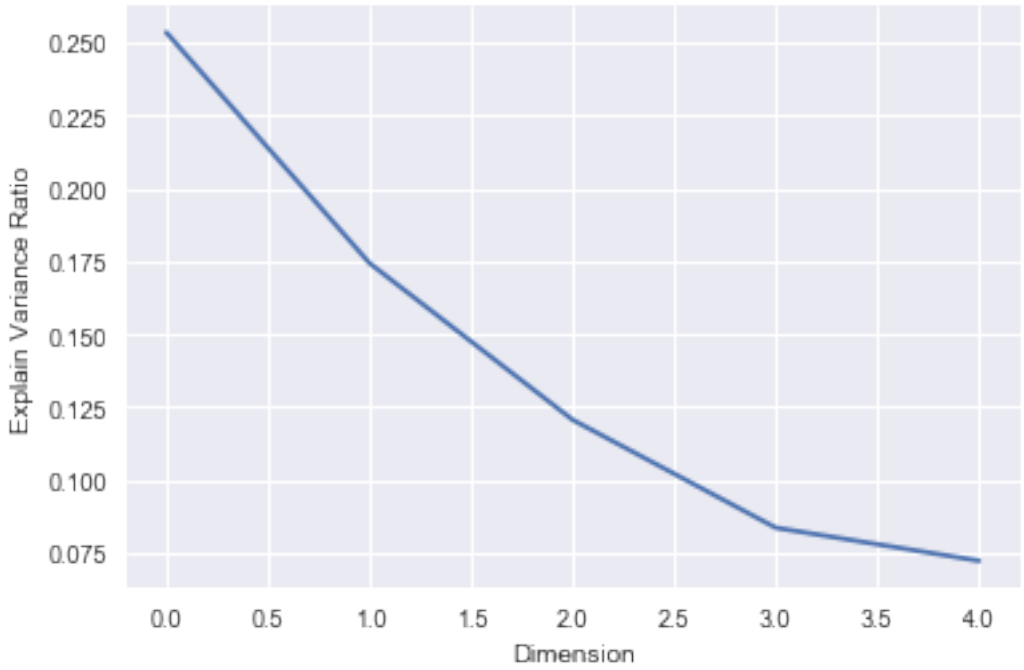
In [167]:
```
pca.explained_variance_
```

Out[167]:
```
array([6.09179704, 4.19158511, 2.90060492, 2.01183001, 1.73962064])
```

In [168]:
```
pca.explained_variance_ratio_
```

Out[168]:
```
array([0.25343736, 0.17438274, 0.12067402, 0.08369827, 0.07237353])
```

In [169]:
```
plt.plot(pca.explained_variance_ratio_)
plt.xlabel('Dimension')
plt.ylabel('Explain Variance Ratio')
plt.show()
#Scree plot
```




```
In [170]: X_train, x_test, Y_train, y_test = train_test_split(X_reduced, Y, test_size=0.2, rand
          clf_svc = LinearSVC(penalty='l1', dual=False, tol=1e-3)
          clf_svc.fit(X_train, Y_train)

Out[170]: LinearSVC(C=1.0, class_weight=None, dual=False, fit_intercept=True,
               intercept_scaling=1, loss='squared_hinge', max_iter=1000,
               multi_class='ovr', penalty='l1', random_state=None, tol=0.001,
               verbose=0)

In [171]: print(clf_svc.score(x_test, y_test))

0.9083969465648855
```

Cluster from scenarios

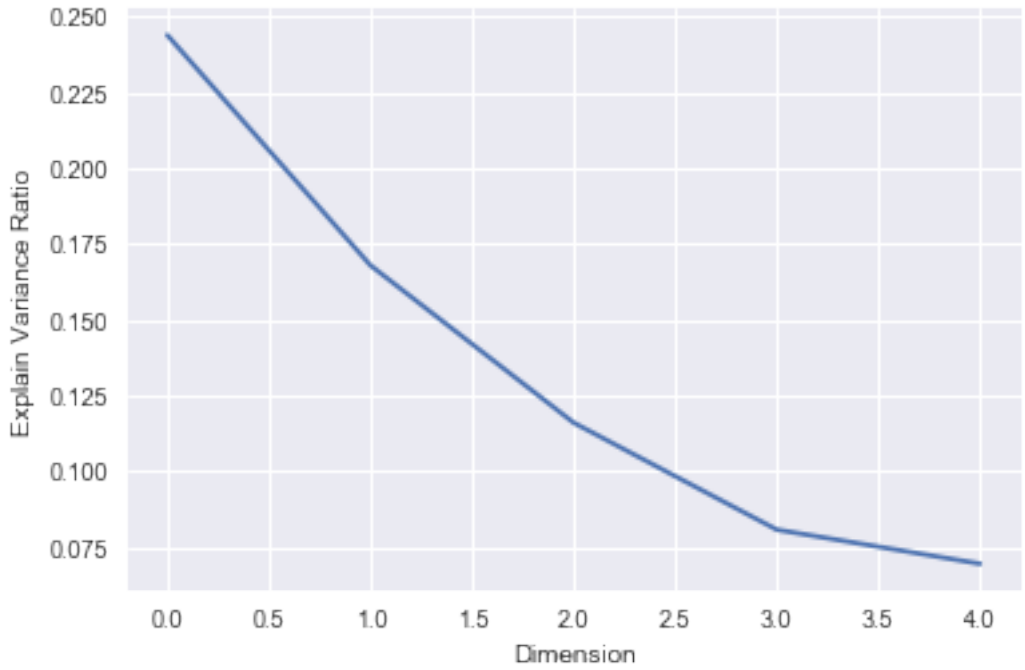
```
In [172]: df_scenario_cluster.head()

Out[172]:    scenario  hadCollision  speed_total_mean  steering_total_mean  \
          0       1.0      0.000000          5.919151             0.503649
          1       1.0      0.000000          7.580378             0.499771
          2       1.0      0.008366          9.474048             0.494557
          3       2.0      0.000000          3.398595             0.523305
          4       2.0      0.000000         11.669419             0.500661
```

```
   brake_total_mean  throttle_total_mean  acceleration_total_mean  \
0          0.965743             0.820576                 0.030731
1          0.891302             0.878839                -0.026652
2          0.952182             0.781126                 0.006292
3          0.960227             0.858795                 0.027363
4          0.891913             0.522365                 0.008028

   speed_total_var  steering_total_var  brake_total_var  \
0        13.796202            0.000655         0.014468
1        31.451253            0.000345         0.058767
2        53.873833            0.001231         0.022506
3         5.036717            0.000101         0.018643
4        47.209285            0.000396         0.055982

            ...             steering_when_running_var  brake_when_running_var  \
0           ...                              0.000174                0.035741
1           ...                              0.000151                0.078618
2           ...                              0.001915                0.121984
3           ...                              0.000048                0.000000
4           ...                              0.000429                0.075733

   throttle_when_running_var  acceleration_when_running_var  \
0                   0.039745                       0.083699
1                   0.000856                       0.087210
2                   0.035585                       0.869046
3                   0.044336                       0.023489
4                   0.087892                       0.212602

   speed_when_running_mean  steering_when_running_mean  \
0                 4.593110                    0.476982
1                 5.276907                    0.508684
2                 6.933534                    0.473303
3                 6.170808                    0.499042
4                 9.577210                    0.504876

   brake_when_running_mean  throttle_when_running_mean  \
0                 0.900460                    0.739390
1                 0.839408                    0.929774
2                 0.769213                    0.707442
3                 1.000000                    0.075305
4                 0.836788                    0.630319

   acceleration_when_running_mean  total_time_when_running
0                        0.016767                 5.239119
1                       -0.065371                 7.705014
2                       -0.068940                 7.848389
3                        0.314027                 1.202199
4                       -0.078850                 4.975241
```

```
          [5 rows x 26 columns]
```

```
In [173]: from sklearn.model_selection import train_test_split

          data = df_scenario_cluster.copy()
          X = data.drop('scenario', axis=1)
          Y = data['scenario']

          from sklearn import preprocessing
          X = preprocessing.scale(X)


          X_train, x_test, Y_train, y_test = train_test_split(X, Y, test_size=0.2, random_state
```

```
In [174]: from sklearn.svm import LinearSVC

          clf_svc = LinearSVC(penalty='l1', dual=False, tol=1e-3)
          clf_svc.fit(X_train, Y_train)

Out[174]: LinearSVC(C=1.0, class_weight=None, dual=False, fit_intercept=True,
               intercept_scaling=1, loss='squared_hinge', max_iter=1000,
               multi_class='ovr', penalty='l1', random_state=None, tol=0.001,
               verbose=0)

In [175]: print(clf_svc.score(x_test, y_test))

0.4351145038167939


In [176]: from sklearn.decomposition import PCA

          pca = PCA(n_components=5, whiten=True)
          X_reduced = pca.fit_transform(X)

In [177]: pca.explained_variance_

Out[177]: array([6.10683356, 4.20528569, 2.90596682, 2.02242817, 1.74126031])

In [178]: pca.explained_variance_ratio_

Out[178]: array([0.24390041, 0.16795462, 0.11606121, 0.08077362, 0.06954408])

In [179]: plt.plot(pca.explained_variance_ratio_)
          plt.xlabel('Dimension')
          plt.ylabel('Explain Variance Ratio')
          plt.show()
```




```
In [180]: X_train, x_test, Y_train, y_test = train_test_split(X_reduced, Y, test_size=0.2, ran
          clf_svc = LinearSVC(penalty='l1', dual=False, tol=1e-3)
          clf_svc.fit(X_train, Y_train)
```

```
Out[180]: LinearSVC(C=1.0, class_weight=None, dual=False, fit_intercept=True,
               intercept_scaling=1, loss='squared_hinge', max_iter=1000,
               multi_class='ovr', penalty='l1', random_state=None, tol=0.001,
               verbose=0)
```

```
In [181]: print(clf_svc.score(x_test, y_test))
```

```
0.29770992366412213
```

9.3.5 MeanShift

```
In [182]: from sklearn.cluster import MeanShift
          dataFrame = df.copy()

          analyzer = MeanShift(bandwidth=70)
          analyzer.fit(dataFrame)
```

```
Out[182]: MeanShift(bandwidth=70, bin_seeding=False, cluster_all=True, min_bin_freq=1,
               n_jobs=1, seeds=None)
```

```
In [183]: from sklearn.cluster import estimate_bandwidth

          estimate_bandwidth(dataFrame)

Out[183]: 80.26694790703168

In [184]: labels = analyzer.labels_

In [185]: np.unique(labels)

Out[185]: array([0, 1, 2, 3, 4, 5, 6])

In [186]: dataFrame['cluster_group'] = np.nan
          data_length = len(dataFrame)
          for i in range(data_length):
              dataFrame.iloc[i, dataFrame.columns.get_loc('cluster_group')] = labels[i]

In [187]: dataFrame.head()

Out[187]:    speed_total_mean  steering_total_mean  brake_total_mean  \
          0          5.919151             0.503649          0.965743
          1          7.580378             0.499771          0.891302
          2          9.474048             0.494557          0.952182
          3          3.398595             0.523305          0.960227
          4         11.669419             0.500661          0.891913

             throttle_total_mean  acceleration_total_mean  speed_total_var  \
          0             0.820576                 0.030731        13.796202
          1             0.878839                -0.026652        31.451253
          2             0.781126                 0.006292        53.873833
          3             0.858795                 0.027363         5.036717
          4             0.522365                 0.008028        47.209285

             steering_total_var  brake_total_var  throttle_total_var  \
          0            0.000655         0.014468            0.028719
          1            0.000345         0.058767            0.010391
          2            0.001231         0.022506            0.045416
          3            0.000101         0.018643            0.075860
          4            0.000396         0.055982            0.112551

             acceleration_total_var      ...        throttle_when_running_var  \
          0                0.039370      ...                          0.039745
          1                0.063480      ...                          0.000856
          2                0.106281      ...                          0.035585
          3                0.027735      ...                          0.044336
          4                0.159198      ...                          0.087892

             acceleration_when_running_var  speed_when_running_mean  \
          0                       0.083699                 4.593110
```

```
1                       0.087210                 5.276907
2                       0.869046                 6.933534
3                       0.023489                 6.170808
4                       0.212602                 9.577210

   steering_when_running_mean  brake_when_running_mean  \
0                    0.476982                 0.900460
1                    0.508684                 0.839408
2                    0.473303                 0.769213
3                    0.499042                 1.000000
4                    0.504876                 0.836788

   throttle_when_running_mean  acceleration_when_running_mean  \
0                    0.739390                        0.016767
1                    0.929774                       -0.065371
2                    0.707442                       -0.068940
3                    0.075305                        0.314027
4                    0.630319                       -0.078850

   total_time_when_running  hadCollision  cluster_group
0                 5.239119             0            0.0
1                 7.705014             0            0.0
2                 7.848389             1            2.0
3                 1.202199             0            0.0
4                 4.975241             0            0.0

[5 rows x 26 columns]
```

```
In [188]: dataFrame.describe()
```

```
Out[188]:        speed_total_mean  steering_total_mean  brake_total_mean  \
count        655.000000           655.000000        655.000000
mean           7.987191             0.502204          0.959427
std            3.740045             0.010688          0.052920
min            0.015117             0.418696          0.441398
25%            5.757110             0.497913          0.944744
50%            7.474806             0.501098          0.975665
75%            9.599790             0.505841          0.991609
max           35.358340             0.561057          1.000000

       throttle_total_mean  acceleration_total_mean  speed_total_var  \
count           655.000000               655.000000       655.000000
mean              0.812115                -0.020020        15.369249
std               0.126159                 0.153846        13.677285
min               0.199416                -1.779510         0.000251
25%               0.771546                -0.008199         7.271839
50%               0.831697                 0.000143        12.569265
75%               0.883929                 0.008954        19.953399
```

```
max               1.000000                 0.171440       124.624108

       steering_total_var  brake_total_var  throttle_total_var  \
count          655.000000       655.000000          655.000000
mean             0.000927         0.018102            0.032908
std              0.002787         0.027826            0.037133
min              0.000000         0.000000            0.000000
25%              0.000078         0.000966            0.011291
50%              0.000304         0.005730            0.020177
75%              0.000662         0.024807            0.039857
max              0.037609         0.200328            0.226159

       acceleration_total_var      ...        throttle_when_running_var  \
count              655.000000      ...                       655.000000
mean                 0.273928      ...                         0.035769
std                  1.551995      ...                         0.041068
min                  0.000007      ...                         0.000000
25%                  0.021014      ...                         0.009976
50%                  0.037177      ...                         0.021740
75%                  0.074261      ...                         0.044298
max                 19.011509      ...                         0.232309

       acceleration_when_running_var  speed_when_running_mean  \
count                   6.550000e+02               655.000000
mean                    2.562516e-01                 7.291854
std                     1.531710e+00                 4.108181
min                     1.663293e-07                 0.015117
25%                     2.505745e-02                 4.883573
50%                     5.581903e-02                 6.459509
75%                     1.002668e-01                 8.786491
max                     1.901151e+01                36.319434

       steering_when_running_mean  brake_when_running_mean  \
count                  655.000000               655.000000
mean                     0.501457                 0.924892
std                      0.015147                 0.082771
min                      0.385388                 0.441398
25%                      0.496080                 0.887307
50%                      0.500045                 0.953593
75%                      0.505914                 0.985116
max                      0.589599                 1.000000

       throttle_when_running_mean  acceleration_when_running_mean  \
count                  655.000000                      655.000000
mean                     0.808358                       -0.044074
std                      0.143912                        0.155841
min                      0.075305                       -1.779510
25%                      0.757800                       -0.052260
```

```
50%                       0.835322                      -0.025605
75%                       0.896792                      -0.002892
max                       1.000000                       0.314027

       total_time_when_running  hadCollision  cluster_group
count               655.000000    655.000000     655.000000
mean                 10.767179      0.093130       0.247328
std                  31.374280      0.290836       0.748018
min                   0.150901      0.000000       0.000000
25%                   4.152824      0.000000       0.000000
50%                   5.155710      0.000000       0.000000
75%                   6.358851      0.000000       0.000000
max                 344.235341      1.000000       6.000000

[8 rows x 26 columns]
```

In [189]:
```
dataFrameCluster = dataFrame.groupby(['cluster_group']).mean()
dataFrameCluster
```

Out[189]:
```
               speed_total_mean  steering_total_mean  brake_total_mean  \
cluster_group
0.0                    7.887470             0.502157          0.958623
1.0                    8.825458             0.503529          0.963621
2.0                    9.232445             0.499030          0.976546
3.0                    7.572369             0.502950          0.974084
4.0                    9.322995             0.508132          0.909123
5.0                    2.641436             0.499082          0.933273
6.0                    3.072536             0.497784          0.992410

               throttle_total_mean  acceleration_total_mean  speed_total_var  \
cluster_group
0.0                       0.806904                -0.022963        14.705734
1.0                       0.818967                 0.003037        21.534679
2.0                       0.837564                 0.001770        22.508241
3.0                       0.910601                -0.011341        12.586728
4.0                       0.895185                -0.014790        16.368998
5.0                       0.974126                -0.002999         6.484751
6.0                       0.969904                -0.003625         2.967740

               steering_total_var  brake_total_var  throttle_total_var  \
cluster_group
0.0                      0.000879         0.018488            0.033597
1.0                      0.001373         0.019832            0.030851
2.0                      0.001884         0.010716            0.028988
3.0                      0.000204         0.007492            0.024185
4.0                      0.000609         0.021577            0.025884
5.0                      0.000026         0.019491            0.002310
6.0                      0.000033         0.000902            0.004328
```

cluster_group	acceleration_total_var	...	brake_when_running_var
0.0	0.307975	...	0.030969
1.0	0.032106	...	0.047379
2.0	0.027321	...	0.049719
3.0	0.095436	...	0.009590
4.0	0.175735	...	0.023818
5.0	0.150991	...	0.021568
6.0	0.033197	...	0.000006

cluster_group	throttle_when_running_var	acceleration_when_running_var
0.0	0.035905	0.280664
1.0	0.034138	0.080932
2.0	0.048340	0.123751
3.0	0.025079	0.083352
4.0	0.023126	0.171621
5.0	0.003742	0.012650
6.0	0.000267	0.017874

cluster_group	speed_when_running_mean	steering_when_running_mean
0.0	7.277498	0.501614
1.0	7.647502	0.500095
2.0	7.410295	0.495831
3.0	6.738427	0.504081
4.0	8.424250	0.512584
5.0	2.570975	0.501220
6.0	2.312557	0.495784

cluster_group	brake_when_running_mean	throttle_when_running_mean
0.0	0.926740	0.809789
1.0	0.905508	0.767292
2.0	0.897335	0.732501
3.0	0.966015	0.935004
4.0	0.893861	0.919901
5.0	0.910892	0.968243
6.0	0.999638	0.995387

cluster_group	acceleration_when_running_mean	total_time_when_running
0.0	-0.049503	5.630962
1.0	-0.008089	6.032106
2.0	-0.000749	6.589248
3.0	-0.021705	201.165259
4.0	-0.014631	79.213483

```
5.0                                -0.004881                 9.074812
6.0                                -0.003943               344.235341

               hadCollision
cluster_group
0.0                0.077465
1.0                0.181818
2.0                0.272727
3.0                0.000000
4.0                0.166667
5.0                1.000000
6.0                1.000000

[7 rows x 25 columns]
```

In [190]:
```
dataFrameCluster['Counts'] = pd.Series(dataFrame.groupby(['cluster_group']).size())
dataFrameCluster
```

Out[190]:

cluster_group	speed_total_mean	steering_total_mean	brake_total_mean	\
0.0	7.887470	0.502157	0.958623	
1.0	8.825458	0.503529	0.963621	
2.0	9.232445	0.499030	0.976546	
3.0	7.572369	0.502950	0.974084	
4.0	9.322995	0.508132	0.909123	
5.0	2.641436	0.499082	0.933273	
6.0	3.072536	0.497784	0.992410	

cluster_group	throttle_total_mean	acceleration_total_mean	speed_total_var	\
0.0	0.806904	-0.022963	14.705734	
1.0	0.818967	0.003037	21.534679	
2.0	0.837564	0.001770	22.508241	
3.0	0.910601	-0.011341	12.586728	
4.0	0.895185	-0.014790	16.368998	
5.0	0.974126	-0.002999	6.484751	
6.0	0.969904	-0.003625	2.967740	

cluster_group	steering_total_var	brake_total_var	throttle_total_var	\
0.0	0.000879	0.018488	0.033597	
1.0	0.001373	0.019832	0.030851	
2.0	0.001884	0.010716	0.028988	
3.0	0.000204	0.007492	0.024185	
4.0	0.000609	0.021577	0.025884	
5.0	0.000026	0.019491	0.002310	
6.0	0.000033	0.000902	0.004328	

cluster_group	acceleration_total_var	...	throttle_when_running_var
0.0	0.307975	...	0.035905
1.0	0.032106	...	0.034138
2.0	0.027321	...	0.048340
3.0	0.095436	...	0.025079
4.0	0.175735	...	0.023126
5.0	0.150991	...	0.003742
6.0	0.033197	...	0.000267

cluster_group	acceleration_when_running_var	speed_when_running_mean
0.0	0.280664	7.277498
1.0	0.080932	7.647502
2.0	0.123751	7.410295
3.0	0.083352	6.738427
4.0	0.171621	8.424250
5.0	0.012650	2.570975
6.0	0.017874	2.312557

cluster_group	steering_when_running_mean	brake_when_running_mean
0.0	0.501614	0.926740
1.0	0.500095	0.905508
2.0	0.495831	0.897335
3.0	0.504081	0.966015
4.0	0.512584	0.893861
5.0	0.501220	0.910892
6.0	0.495784	0.999638

cluster_group	throttle_when_running_mean	acceleration_when_running_mean
0.0	0.809789	-0.049503
1.0	0.767292	-0.008089
2.0	0.732501	-0.000749
3.0	0.935004	-0.021705
4.0	0.919901	-0.014631
5.0	0.968243	-0.004881
6.0	0.995387	-0.003943

cluster_group	total_time_when_running	hadCollision	Counts
0.0	5.630962	0.077465	568
1.0	6.032106	0.181818	44
2.0	6.589248	0.272727	22
3.0	201.165259	0.000000	13
4.0	79.213483	0.166667	6
5.0	9.074812	1.000000	1

```
6.0                          344.235341      1.000000       1

[7 rows x 26 columns]
```

```
In [191]: dataFrame[dataFrame['cluster_group'] == 2.0].describe()
```

Out[191]:

	speed_total_mean	steering_total_mean	brake_total_mean
count	22.000000	22.000000	22.000000
mean	9.232445	0.499030	0.976546
std	1.608238	0.008093	0.013095
min	5.653445	0.463986	0.948506
25%	8.272409	0.499774	0.967847
50%	9.500332	0.500607	0.978298
75%	10.183151	0.501842	0.984882
max	12.279223	0.504581	0.996079

	throttle_total_mean	acceleration_total_mean	speed_total_var
count	22.000000	22.000000	22.000000
mean	0.837564	0.001770	22.508241
std	0.048628	0.001965	10.582811
min	0.737526	-0.001478	6.450247
25%	0.794283	0.000632	16.599492
50%	0.851024	0.001406	21.486849
75%	0.867522	0.002863	26.074040
max	0.913743	0.006292	53.873833

	steering_total_var	brake_total_var	throttle_total_var
count	22.000000	22.000000	22.000000
mean	0.001884	0.010716	0.028988
std	0.005150	0.011309	0.020958
min	0.000217	0.000506	0.005739
25%	0.000340	0.002784	0.012855
50%	0.000544	0.006052	0.021298
75%	0.001216	0.017754	0.043858
max	0.024657	0.037204	0.083085

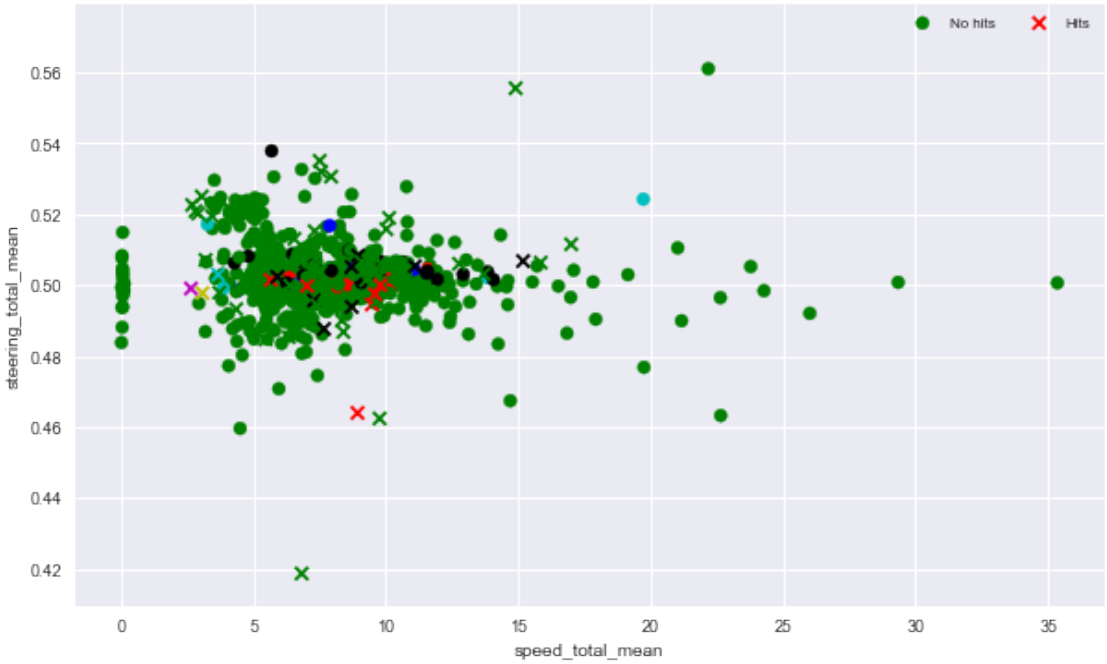
	acceleration_total_var	...	throttle_when_running_var
count	22.000000	...	22.000000
mean	0.027321	...	0.048340
std	0.020993	...	0.038391
min	0.007523	...	0.002528
25%	0.015021	...	0.028253
50%	0.021042	...	0.036624
75%	0.031946	...	0.064602
max	0.106281	...	0.169520

	acceleration_when_running_var	speed_when_running_mean
count	22.000000	22.000000

```
mean                        0.123751                 7.410295
std                         0.176535                 2.723078
min                         0.011597                 4.257427
25%                         0.044624                 5.495379
50%                         0.081811                 6.714656
75%                         0.117960                 8.176628
max                         0.869046                14.278999

       steering_when_running_mean  brake_when_running_mean  \
count                   22.000000                22.000000
mean                     0.495831                 0.897335
std                      0.008894                 0.099058
min                      0.470238                 0.611971
25%                      0.493801                 0.890907
50%                      0.497187                 0.926809
75%                      0.501390                 0.960155
max                      0.505552                 1.000000

       throttle_when_running_mean  acceleration_when_running_mean  \
count                   22.000000                       22.000000
mean                     0.732501                       -0.000749
std                      0.157627                        0.055754
min                      0.279519                       -0.068940
25%                      0.651544                       -0.024866
50%                      0.769994                       -0.015546
75%                      0.830914                        0.003633
max                      0.951579                        0.208821

       total_time_when_running  hadCollision  cluster_group
count                22.000000     22.000000           22.0
mean                  6.589248      0.272727            2.0
std                   4.290100      0.455842            0.0
min                   1.889065      0.000000            2.0
25%                   4.458859      0.000000            2.0
50%                   5.486243      0.000000            2.0
75%                   6.962391      0.750000            2.0
max                  21.506701      1.000000            2.0

[8 rows x 26 columns]
```

In [230]:
```
plt.figure(figsize=(10, 6))

#dataFrame = reactionPressBrake[reactionPressBrake.reaction_time.dt.total_seconds()

LABEL_COLOR_MAP = {0 : 'g',
                   1 : 'k',
                   2 : 'r',
                   3 : 'b',
```

```
                    4 : 'c',
                    5 : 'm',
                    6 : 'y'
                    }

label_color = [LABEL_COLOR_MAP[l] for l in dataFrame['cluster_group']]

markerMap = np.where(dataFrame['hadCollision'] == 0, 'o', 'x')
dataFrameNoHits = dataFrame[dataFrame.hadCollision == 0]
label_color_no_hit = [LABEL_COLOR_MAP[l] for l in dataFrameNoHits['cluster_group']]
dataFrameHits = dataFrame[dataFrame.hadCollision == 1]
label_color_hit = [LABEL_COLOR_MAP[l] for l in dataFrameHits['cluster_group']]


x = 'speed_total_mean'
y = 'steering_total_mean'
c = 'cluster_group'

noHits = plt.scatter(dataFrameNoHits[x], dataFrameNoHits[y], c=label_color_no_hit, ma
Hits = plt.scatter(dataFrameHits[x], dataFrameHits[y], c=label_color_hit, marker='x')

plt.xlabel(x)
plt.ylabel(y)

plt.legend(( noHits,  Hits),
           ("No hits","Hits"),
           scatterpoints=1,
           ncol=3,
           fontsize=8)

plt.grid(True)

plt.show()
```



In [ ]: *## Como tomar los datos de session a partir de acá*