SPORT ANALYSIS WITH PYTHON





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Football analysis

- 1. Match (score, time, ...)
- 2. Players (performance, goal, award, match, time...)
- 3. Fan (number, loyalty, scandal...)

. . .

Match analysis (real-time match tracking)



OUTLINE

1 Data analysis with Python

2 Data visualization with Tableau

DATA



- 1. Timestamp
- 2. Devices_id
- 3. Positions (x_pos, y_pos)

Calculate

> Real Time

> Distance

> Total distance

- 1. Velocity
- 2. Turning points
- 3. Zone

Load Data

```
# # IMPORT LIBRARIES
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import random as rd
import os, time, math, datetime
from math import sqrt
# # Load Data
def loadCSV(csvName, sep):
   data = pd.read csv(csvName, sep= sep)
   return data
print os.getcwd()
/home/mapr/notebookHome/THUYLE TEST
```

```
path = "/home/mapr/notebookHome/THUYLE_TEST"

os.chdir(path)
sep = "|"
df = loadCSV("ThuyLe_P1_Demo_TotalDistance.csv", sep)
```

Load Data

df.head(15)

	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	distance	totalDistance
0	1.237747e+12	9	50.875526	25.792036	1	1	0.0000	0.0000
1	1.237747e+12	8	28.525500	41.909308	1	1	0.0000	0.0000
2	1.237747e+12	2	21.108100	12.481253	1	1	0.0000	0.0000
3	1.237747e+12	6	45.247400	26.818432	1	1	0.0000	0.0000
4	1.237747e+12	7	32.231844	12.636959	1	1	0.0000	0.0000
5	1.237747e+12	5	28.107198	40.277207	1	1	0.0000	0.0000
6	1.237747e+12	11	41.571300	61.727806	1	1	0.0000	0.0000
7	1.237747e+12	1	4.295718	45.317342	1	1	0.0000	0.0000
8	1.237747e+12	4	35.724664	29.717300	1	1	0.0000	0.0000
9	1.237747e+12	3	26.620200	56.002000	1	1	0.0000	0.0000
10	1.237747e+12	10	51.444286	49.769524	1	1	0.0000	0.0000
11	1.237747e+12	9	50.907026	25.734036	1	1	0.0660	0.0660
12	1.237747e+12	8	28.521800	41.885908	1	1	0.0237	0.0237
13	1.237747e+12	2	21.154600	12.468753	1	1	0.0482	0.0482
14	1.237747e+12	6	45.247400	26.818432	1	1	0.0000	0.0000

Data analysis

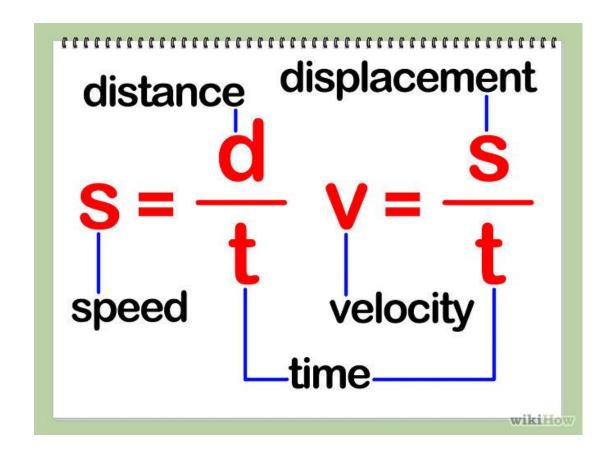
```
# # Columns of Data
df.columns
Index([u'timestamp', u'device_id', u'x_pos', u'y_pos', u'matchTime_minute',
      u'matchTime_second', u'distance', u'totalDistance'],
     dtype='object')
                 # # Unique value of column "device id"
                 devices = df['device_id'].unique()
                 print devices
                 [9 8 2 6 7 5 11 1 4 3 10]
```

Describe Data

#======== # # Describe Data #====== df.describe()

	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	totalDistance
count	5.940110e+05	594011.00000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000
mean	1.237748e+12	6.00000	40.995397	43.856025	23.000426	30.499454	3062.034685
std	7.794380e+05	3.16228	23.624175	19.094993	12.987441	17.318422	1757.739790
min	1.237747e+12	1.00000	-3.851690	-3.709834	1.000000	1.000000	0.000000
25%	1.237747e+12	3.00000	22.013650	29.903948	12.000000	15.000000	1598.513550
50%	1.237748e+12	6.00000	42.212144	43.642300	23.000000	30.000000	3025.199000
75%	1.237749e+12	9.00000	57.166875	57.709330	34.000000	45.000000	4474.828350
max	1.237749e+12	11.00000	104.214000	91.831700	46.000000	60.000000	7938.709100





```
# # Calculate the difference time
def difTime(_df, _devices):
   df['difTime'] = 0
   for d in devices:
       idd = _df[_df['device_id'] == d].index
       lend = len(idd)
       for i in range (1, lend):
           df.loc[idd[i], 'difTime'] = float(( df.loc[idd[i], 'timestamp'] \
                                                - _df.loc[idd[i-1], 'timestamp'])/1000)
   return df
```

```
df = difTime(df, devices)
df
```

	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	distance	totalDistance	difTime
0	1.237747e+12	9	50.875526	25.792036	1	1	0	0.0000	0.00
1	1.237747e+12	8	28.525500	41.909308	1	1	0	0.0000	0.00
2	1.237747e+12	2	21.108100	12.481253	1	1	0	0.0000	0.00
3	1.237747e+12	6	45.247400	26.818432	1	1	0	0.0000	0.00
4	1.237747e+12	7	32.231844	12.636959	1	1	0	0.0000	0.00
5	1.237747e+12	5	28.107198	40.277207	1	1	0	0.0000	0.00
6	1.237747e+12	11	41.571300	61.727806	1	1	0	0.0000	0.00
7	1.237747e+12	1	4.295718	45.317342	1	1	0	0.0000	0.00
8	1.237747e+12	4	35.724664	29.717300	1	1	0	0.0000	0.00
9	1.237747e+12	3	26.620200	56.002000	1	1	0	0.0000	0.00
10	1.237747e+12	10	51.444286	49.769524	1	1	0	0.0000	0.00
11	1.237747e+12	9	50.907026	25.734036	1	1	0.0660	0.0660	0.05
12	1.237747e+12	8	28.521800	41.885908	1	1	0.0237	0.0237	0.05
13	1.237747e+12	2	21.154600	12.468753	1	1	0.0482	0.0482	0.05

```
df['difTime'].value_counts(dropna=False)
```

```
0.05 594000
0.00 11
```

Name: difTime, dtype: int64

```
# # Calculate Velocity
def velocity( df, devices):
   df['velocity'] = 0
   for d in devices:
       idd = _df[_df['device_id'] == d].index
       print type(idd)
       iVe = idd[1:]
       _df['velocity'][iVe] = _df["distance"][iVe].astype("float")/\
                               df['difTime'][iVe].astype("float")
   return df
                                                   df = velocity(df, devices)
                                                   df
```

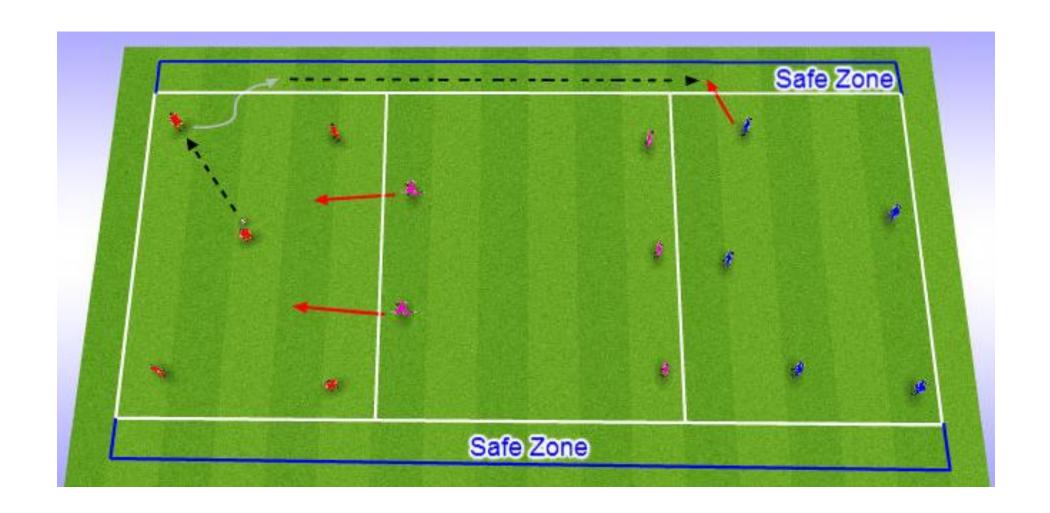
	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	distance	totalDistance	difTime	velocity
0	1.237747e+12	9	50.875526	25.792036	1	1	0	0.0000	0.00	0.000
1	1.237747e+12	8	28.525500	41.909308	1	1	0	0.0000	0.00	0.000
2	1.237747e+12	2	21.108100	12.481253	1	1	0	0.0000	0.00	0.000
3	1.237747e+12	6	45.247400	26.818432	1	1	0	0.0000	0.00	0.000
4	1.237747e+12	7	32.231844	12.636959	1	1	0	0.0000	0.00	0.000
5	1.237747e+12	5	28.107198	40.277207	1	1	0	0.0000	0.00	0.000
6	1.237747e+12	11	41.571300	61.727806	1	1	0	0.0000	0.00	0.000
7	1.237747e+12	1	4.295718	45.317342	1	1	0	0.0000	0.00	0.000
8	1.237747e+12	4	35.724664	29.717300	1	1	0	0.0000	0.00	0.000
9	1.237747e+12	3	26.620200	56.002000	1	1	0	0.0000	0.00	0.000
10	1.237747e+12	10	51.444286	49.769524	1	1	0	0.0000	0.00	0.000
11	1.237747e+12	9	50.907026	25.734036	1	1	0.0660	0.0660	0.05	1.320
12	1.237747e+12	8	28.521800	41.885908	1	1	0.0237	0.0237	0.05	0.474
13	1.237747e+12	2	21.154600	12.468753	1	1	0.0482	0.0482	0.05	0.964

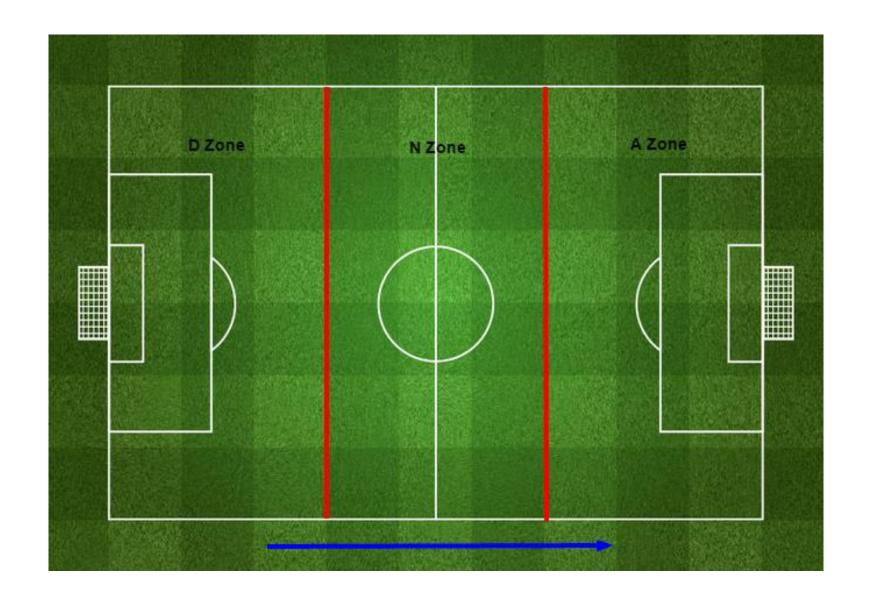
In this case the difference time are always 50 ms

	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	distance	totalDistance	difTime	velocity	speed
0	1.237747e+12	9	50.875526	25.792036	1	1	0	0.0000	0.00	0.000	0.000
1	1.237747e+12	8	28.525500	41.909308	1	1	0	0.0000	0.00	0.000	0.000
2	1.237747e+12	2	21.108100	12.481253	1	1	0	0.0000	0.00	0.000	0.000
3	1.237747e+12	6	45.247400	26.818432	1	1	0	0.0000	0.00	0.000	0.000
4	1.237747e+12	7	32.231844	12.636959	1	1	0	0.0000	0.00	0.000	0.000
5	1.237747e+12	5	28.107198	40.277207	1	1	0	0.0000	0.00	0.000	0.000
6	1.237747e+12	11	41.571300	61.727806	1	1	0	0.0000	0.00	0.000	0.000
7	1.237747e+12	1	4.295718	45.317342	1	1	0	0.0000	0.00	0.000	0.000
8	1.237747e+12	4	35.724664	29.717300	1	1	0	0.0000	0.00	0.000	0.000
9	1.237747e+12	3	26.620200	56.002000	1	1	0	0.0000	0.00	0.000	0.000
10	1.237747e+12	10	51.444286	49.769524	1	1	0	0.0000	0.00	0.000	0.000
11	1.237747e+12	9	50.907026	25.734036	1	1	0.0660	0.0660	0.05	1.320	1.320
12	1.237747e+12	8	28.521800	41.885908	1	1	0.0237	0.0237	0.05	0.474	0.474
13	1.237747e+12	2	21.154600	12.468753	1	1	0.0482	0.0482	0.05	0.964	0.964
14	1.237747e+12	6	45.247400	26.818432	1	1	0.0000	0.0000	0.05	0.000	0.000
15	1.237747e+12	7	32.231844	12.636959	1	1	0.0000	0.0000	0.05	0.000	0.000

#======= # # Describe Data #======= df.describe()

	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	totalDistance	difTime	velocity
count	5.940110e+05	594011.00000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000
mean	1.237748e+12	6.00000	40.995397	43.856025	23.000426	30.499454	3062.034685	0.049999	2.250519
std	7.794380e+05	3.16228	23.624175	19.094993	12.987441	17.318422	1757.739790	0.000215	1.419839
min	1.237747e+12	1.00000	-3.851690	-3.709834	1.000000	1.000000	0.000000	0.000000	0.000000
25%	1.237747e+12	3.00000	22.013650	29.903948	12.000000	15.000000	1598.513550	0.050000	1.170000
50%	1.237748e+12	6.00000	42.212144	43.642300	23.000000	30.000000	3025.199000	0.050000	1.998000
75%	1.237749e+12	9.00000	57.166875	57.709330	34.000000	45.000000	4474.828350	0.050000	3.112000
max	1.237749e+12	11.00000	104.214000	91.831700	46.000000	60.000000	7938.709100	0.050000	9.778000





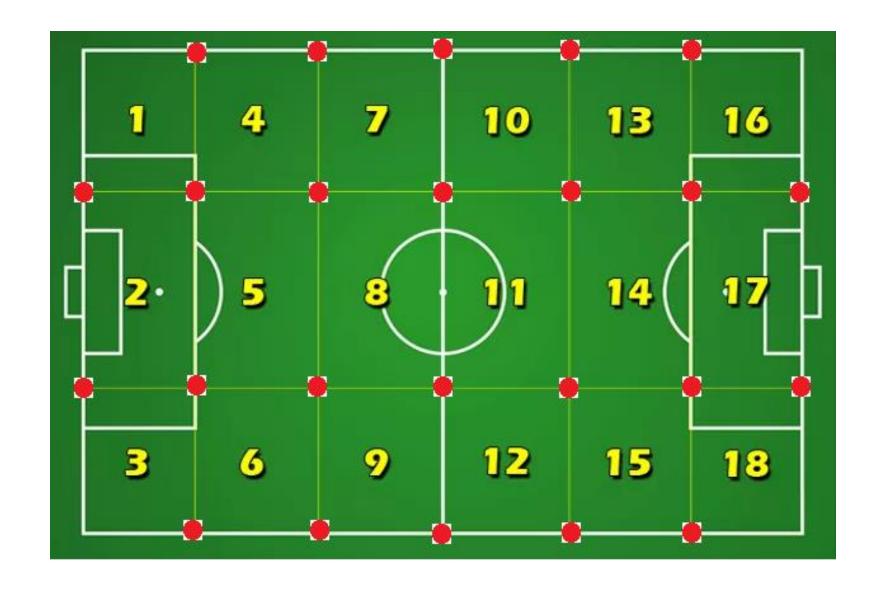
```
# # Calculate Zone (3 zones)
def calculate3Zones( df, x):
   df["3Zones"] = "'
   lendf = len(df)
   xZone1 = x/3
   x7one3 = 2*x7one1
   for i in range (0, lendf):
        if ( df.loc[i, "x pos"] < xZone1):</pre>
           df.loc[i, "3Zones"] = "Zone 1"
        else:
            if ( df.loc[i, "x pos"] > xZone3):
                df.loc[i, "3Zones"] = "Zone 3"
            else:
                df.loc[i, "3Zones"] = "Zone 2"
    return df
```

```
x = 100 # The long of the pitch is 100 m
y = 90 # The wide of the pitch is 90m
df = calculate3Zones(df, x)
df
```

	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	distance	totalDistance	difTime	velocity	3Zones
0	1.237747e+12	9	50.875526	25.792036	1	1	0.0000	0.0000	0.00	0.000	Zone_2
1	1.237747e+12	8	28.525500	41.909308	1	1	0.0000	0.0000	0.00	0.000	Zone_1
2	1.237747e+12	2	21.108100	12.481253	1	1	0.0000	0.0000	0.00	0.000	Zone_1
3	1.237747e+12	6	45.247400	26.818432	1	1	0.0000	0.0000	0.00	0.000	Zone_2
4	1.237747e+12	7	32.231844	12.636959	1	1	0.0000	0.0000	0.00	0.000	Zone_1
5	1.237747e+12	5	28.107198	40.277207	1	1	0.0000	0.0000	0.00	0.000	Zone_1
6	1.237747e+12	11	41.571300	61.727806	1	1	0.0000	0.0000	0.00	0.000	Zone_2
7	1.237747e+12	1	4.295718	45.317342	1	1	0.0000	0.0000	0.00	0.000	Zone_1
8	1.237747e+12	4	35.724664	29.717300	1	1	0.0000	0.0000	0.00	0.000	Zone_2
9	1.237747e+12	3	26.620200	56.002000	1	1	0.0000	0.0000	0.00	0.000	Zone_1
10	1.237747e+12	10	51.444286	49.769524	1	1	0.0000	0.0000	0.00	0.000	Zone_2
11	1.237747e+12	9	50.907026	25.734036	1	1	0.0660	0.0660	0.05	1.320	Zone_2
12	1.237747e+12	8	28.521800	41.885908	1	1	0.0237	0.0237	0.05	0.474	Zone_1







```
# # Get Xzone
def getXzone(_xp, _xUnit):
   if xp < xUnit:
       return "x1"
   else:
       if (_xp >= _xUnit) and (_xp <= 2*_xUnit):
           return "x2"
       else:
           if (xp \ge 2*_xUnit) and (xp < 3*_xUnit):
               return "x3"
           else:
               if (xp >= 3*xUnit) and (xp < 4*xUnit):
                   return "x4"
               else:
                   if (xp >= 4*_xUnit) and (xp < 5*_xUnit):
                       return "x5"
                   else:
                       return "x6"
```

```
# # Get Yzone
def getYzone(_yp, _yUnit):
    if _yp < _yUnit:</pre>
        return "y1"
    else:
        if (\_yp >= \_yUnit) and (\_yp <= 2*\_yUnit):
            return "y2"
        else:
             return "y3"
```

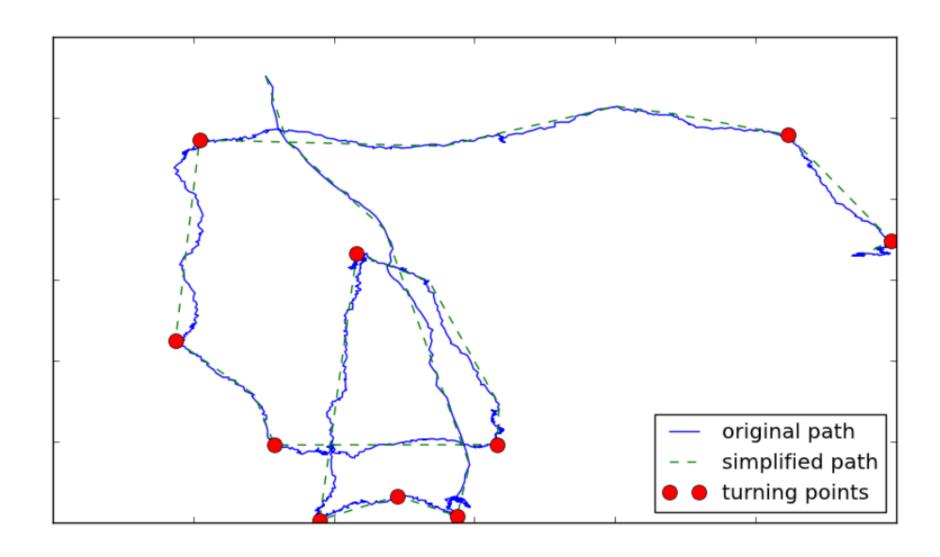
return df

```
# # Calculate Zone (18 zones)
def calculate18Zone( df, long, wide):
   df["18Zones"] =
   lendf = len(df)
   xUnit = long/6
   yUnit = wide/3
   dictZones = {"('x1', 'y3')" : "z1", "('x1', 'y2')" : "z2", "('x1', 'y1')" : "z3",\
                "('x2', 'y3')" : "z4", "('x2', 'y2')" : "z5", "('x2', 'y1')" : "z6",\
                "('x3', 'y3')" : "z7", "('x3', 'y2')" : "z8", "('x3', 'y1')" : "z9",\
                "('x4', 'y3')" : "z10", "('x4', 'y2')" : "z11", "('x4', 'y1')" : "z12",\
                "('x5', 'y3')" : "z13", "('x5', 'y2')" : "z14", "('x5', 'y1')" : "z15",\
                "('x6', 'y3')" : "z16", "('x6', 'y2')" : "z17", "('x6', 'y1')" : "z18"}
   for i in range (0, lendf):
       xZone = getXzone(_df.loc[i, "x pos"], xUnit)
       yZone = getYzone( df.loc[i, "y pos"], yUnit)
       df.loc[i, "18Zones"] = dictZones[str((xZone, yZone))]
```

```
df = calculate18Zones(df, x, y)
df
```

	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	distance	totalDistance	difTime	velocity	3Zones	18Zones
0	1.237747e+12	9	50.875526	25.792036	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z12
1	1.237747e+12	8	28.525500	41.909308	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z 5
2	1.237747e+12	2	21.108100	12.481253	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z6
3	1.237747e+12	6	45.247400	26.818432	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z9
4	1.237747e+12	7	32.231844	12.636959	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z9
5	1.237747e+12	5	28.107198	40.277207	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z 5
6	1.237747e+12	11	41.571300	61.727806	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z7
7	1.237747e+12	1	4.295718	45.317342	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z2
8	1.237747e+12	4	35.724664	29.717300	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z9
9	1.237747e+12	3	26.620200	56.002000	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z 5
10	1.237747e+12	10	51.444286	49.769524	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z11
11	1.237747e+12	9	50.907026	25.734036	1	1	0.0660	0.0660	0.05	1.320	Zone_2	z12
12	1.237747e+12	8	28.521800	41.885908	1	1	0.0237	0.0237	0.05	0.474	Zone_1	z 5
13	1.237747e+12	2	21.154600	12.468753	1	1	0.0482	0.0482	0.05	0.964	Zone_1	z6

Turning points

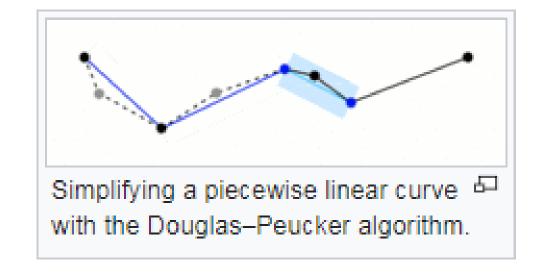


Ramer - Douglas - Peucker algorithm

https://en.wikipedia.org/wiki/Ramer%E2%80%93Douglas%E2%80%93Peucker algorithm

Given a <u>curve composed of</u>
<u>line segments</u> to find a similar curve with fewer points.

The algorithm defines 'dissimilar' based on the maximum distance between the original curve and the simplified curve.



The simplified curve consists of a subset of the points that defined the original curve.

The true turning point is always debatable and will depend on the angle that one specifies that has to lie between points.

- 1. Calculate distance between subsequent points
- 2. Calculate angle between subsequent points
- 3. Look how distance / angle changes between subsequent points

Use the RamerDouglas-Peucker
(RDP) algorithm to simplify the path

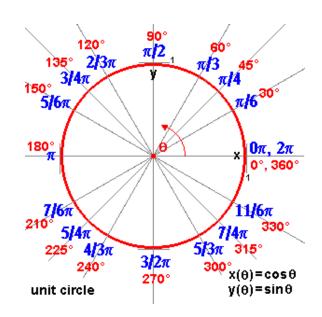
```
# # Reduces a series of points to a simplified version that loses detail,
# # but maintains the general shape of the series.
def rdp(_points, _epsilon):
    dmax = 0.0
    index = 0
    lenP = len(points) - 1
   # Find the point with the maximum distance
   for i in range(1, lenP):
        d = pointLineDistance( points[i], points[0], points[-1])
       if d > dmax:
            index = i
            dmax = d
    # If max distance is greater than epsilon, recursively simplify
    if (dmax >= epsilon):
       # Recursive call
       results = rdp( points[:index+1], epsilon)[:-1] +\
                    rdp( points[index:], epsilon)
    else:
        results = [_points[0], _points[-1]]
    return results
```

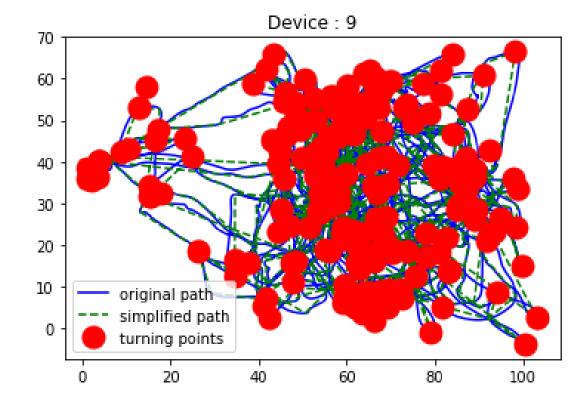
```
# Returns the angles between vectors.
     Parameters:
    dir is a 2D-array of shape (N,M) representing N vectors in M-dimensional space.
    The return value is a 1D-array of values of shape (N-1,), with each value
    between 0 and pi.
    0 implies the vectors point in the same direction
    pi/2 implies the vectors are orthogonal
     pi implies the vectors point in opposite directions
def angle( arr):
   dir2 = arr[1:]
   dir1 = arr[:-1]
    return np.arccos((dir1*dir2).sum(axis=1)/\
                     (np.sqrt((dir1**2).sum(axis=1)*(dir2**2).sum(axis=1))))
```

```
# # Use the Ramer-Douglas-Peucker algorithm to simplify the path
def turn points( df, xCol, yCol, tolerance, min angle, turnArr, d):
    num samples = len(df)
   if (num samples > 0):
       _min_angle = np.pi * _min_angle
       npcoord = _df[[_xCol, _yCol]].as_matrix()
       simplified = np.array(rdp(npcoord.tolist(), _tolerance))
       sx, sy = simplified.T
       directions = np.diff(simplified, axis=0)
       theta = angle(directions)
       idx = np.where(theta > min angle)[0] + 1
       lenIdx = len(idx)
       for i in range(0, lenIdx):
           exp = (_df[_xCol] == sx[idx[i]]) & (_df[_yCol] == sy[idx[i]])
           # add index into turnArr
           turnArr.append( df.index[exp])
        plotTurning( df[ xCol], df[ yCol], sx, sy, idx, d)
       return turnArr
   else:
        return turnArr
```

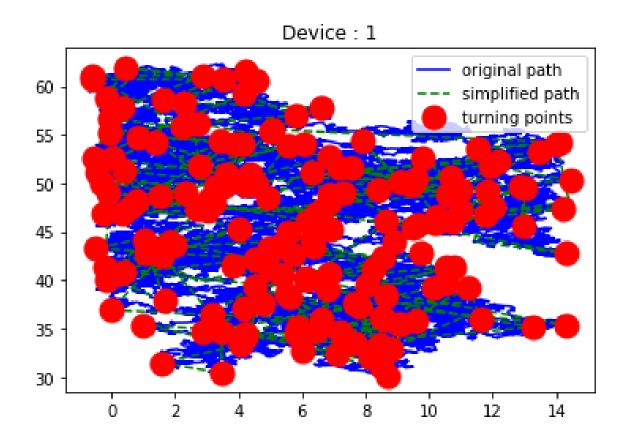
```
# # Plot turning point of a player
def plotTurning(coord_x, coord_y, sx, sy, idx, _d):
   fig = plt.figure()
   figsize=(1, 1)
    ax = fig.add_subplot(111)
    ax.plot(coord x, coord y, 'b-', label='original path')
    ax.plot(sx, sy, 'g--', label='simplified path')
    ax.plot(sx[idx], sy[idx], 'ro', markersize = 15, label='turning points')
    plt.legend(loc='best')
    plt.title('Device : '+ str( d))
    plt.show()
```

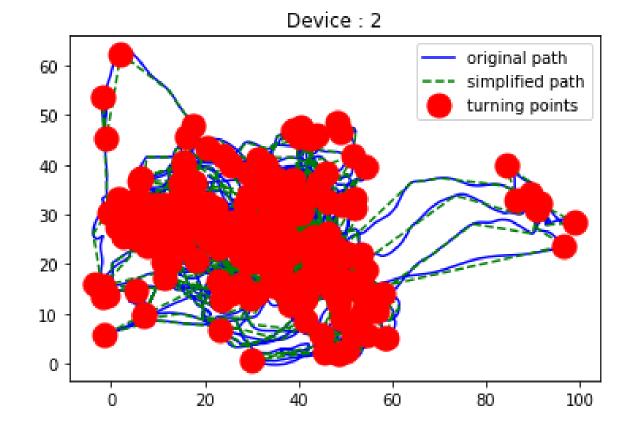
```
# Calculate Turning Points
def calculateTurningPoints(_df, _devices, _tolerance, _min_angle):
   turnArr = []
   for d in _devices:
        dfd = df[ df['device id'] == d]
        print "\n\nTurning points of device >>> ", d
       turnArr = turn_points(dfd, 'x_pos', 'y_pos', _tolerance, _min_angle, turnArr, d)
   # Add turnAng to DataFrame
    _df['turnAng'] = 0
   for i in turnArr:
       _df.loc[i, 'turnAng'] = 1
    return df
```



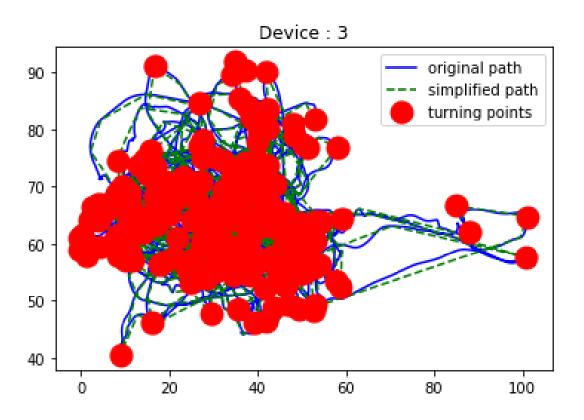


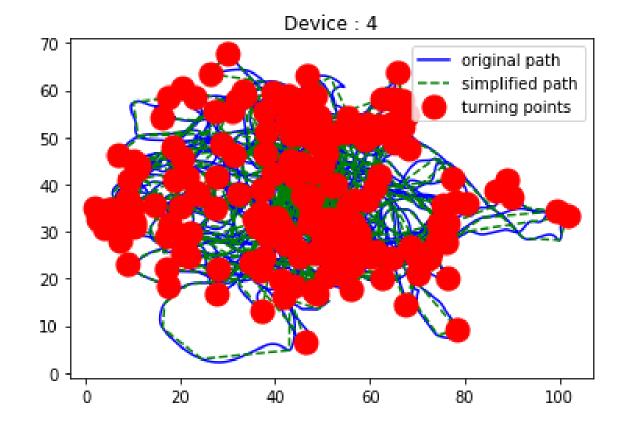
Turning points of device >>> 1



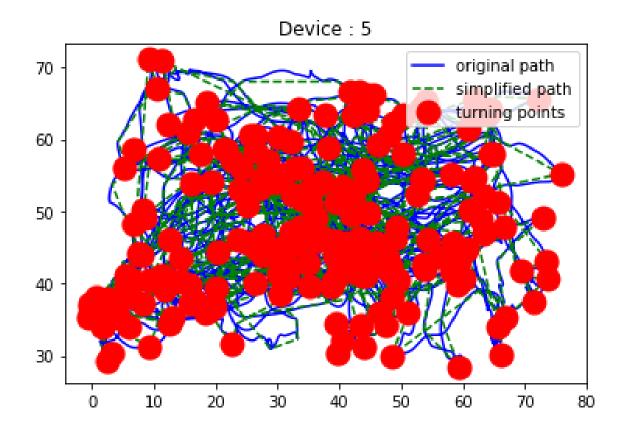


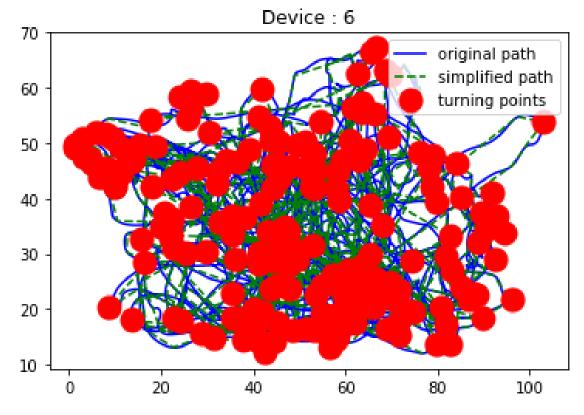
Turning points of device >>> 3



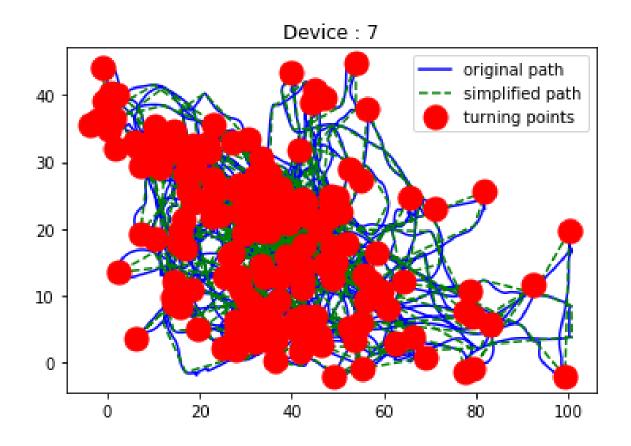


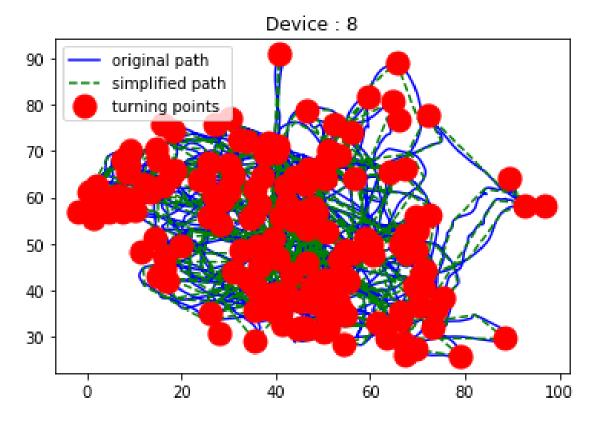
Turning points of device >>> 5



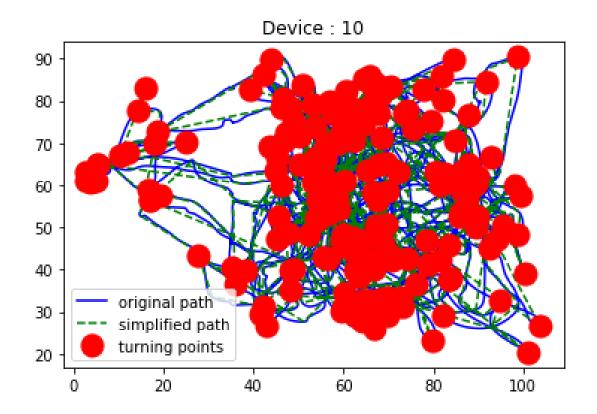


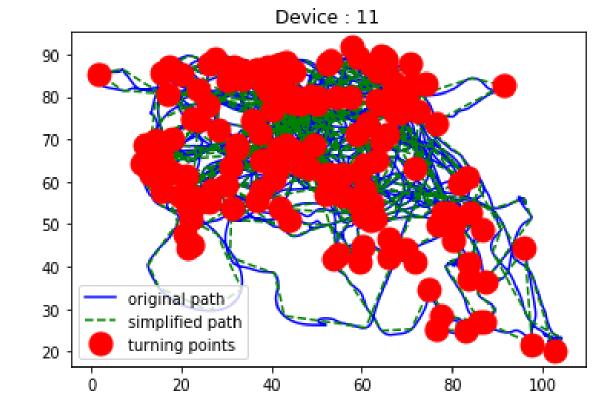
Turning points of device >>> 7





Turning points of device >>> 10





	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	distance	totalDistance	difTime	velocity	3Zones	18Zones	turnAng
0	1.237747e+12	9	50.875526	25.792036	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z12	0
1	1.237747e+12	8	28.525500	41.909308	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z5	0
2	1.237747e+12	2	21.108100	12.481253	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z6	0
3	1.237747e+12	6	45.247400	26.818432	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z9	0
4	1.237747e+12	7	32.231844	12.636959	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z9	0
5	1.237747e+12	5	28.107198	40.277207	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z5	0
6	1.237747e+12	11	41.571300	61.727806	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z7	0
7	1.237747e+12	1	4.295718	45.317342	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z2	0
8	1.237747e+12	4	35.724664	29.717300	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z9	0
9	1.237747e+12	3	26.620200	56.002000	1	1	0.0000	0.0000	0.00	0.000	Zone_1	z5	0
10	1.237747e+12	10	51.444286	49.769524	1	1	0.0000	0.0000	0.00	0.000	Zone_2	z11	0
11	1.237747e+12	9	50.907026	25.734036	1	1	0.0660	0.0660	0.05	1.320	Zone_2	z12	0
12	1.237747e+12	8	28.521800	41.885908	1	1	0.0237	0.0237	0.05	0.474	Zone_1	z 5	0

Describe Data

#======= # # Describe Data #======= df.describe()

	timestamp	device_id	x_pos	y_pos	matchTime_minute	matchTime_second	distance	totalDistance	difTime	velocity	turnAng
count	5.940110e+05	594011.00000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000	594011.000000
mean	1.237748e+12	6.00000	40.995397	43.856025	23.000426	30.499454	0.112526	3062.034685	0.049999	2.250519	0.004414
std	7.794380e+05	3.16228	23.624175	19.094993	12.987441	17.318422	0.070992	1757.739790	0.000215	1.419839	0.066292
min	1.237747e+12	1.00000	-3.851690	-3.709834	1.000000	1.000000	0.000000	0.000000	0.000000	0.000000	0.000000
25%	1.237747e+12	3.00000	22.013650	29.903948	12.000000	15.000000	0.058500	1598.513550	0.050000	1.170000	0.000000
50%	1.237748e+12	6.00000	42.212144	43.642300	23.000000	30.000000	0.099900	3025.199000	0.050000	1.998000	0.000000
75%	1.237749e+12	9.00000	57.166875	57.709330	34.000000	45.000000	0.155600	4474.828350	0.050000	3.112000	0.000000
max	1.237749e+12	11.00000	104.214000	91.831700	46.000000	60.000000	0.488900	7938.709100	0.050000	9.778000	1.000000

Describe Data

Describe Data

Device	>>> 1				Device	>>> 10								
count	timestamp 5.400100e+04	Device		المراجعة المراجعة	count	timestamp 5.400100e+04	device_id 54001.0	F4004	x_pos .000000	54001.	y_pos		me_minute	
mean	1.237748e+12	count	timestamp 5.400100e+04	devio	count	1.237748e+12	10.0		. 239701		744580	546	23.000426	
std	7.794445e+05	mean	1.237748e+12	54	mean std	7.794445e+05	0.0		614398		013511		12.987550	
min	1.237747e+12	std	7.794445e+05		min	1.237747e+12								
25% 50%	1.237747e+12 1.237748e+12	min	1.237747e+12		25%		10.0		723298		267654		1.000000	
75%	1.237748e+12 1.237749e+12	25%	1.237747e+12			1.237747e+12	10.0		836486		256924		12.000000	
max	1.237749e+12	50%	1.237748e+12		50%	1.237748e+12	10.0		444511		040736		23.000000	
		75%	1.237749e+12		75%	1.237749e+12	10.0		984686		982824		34.000000	
	matchTime_se	max	1.237749e+12		max	1.237749e+12	10.0	103	882186	90.	441724		46.000000	9
count	54001.00		matchTime coc	and										
mean std	30.49 17.31	count	matchTime_sec 54001.000			matchTime_sec		istance		istance		difTime	\	
min	1.00		30.499		count	54001.000		.000000		.000000		.000000		
25%	15.00		17.318568		mean	30.499		.109908		.837325		.049999		
50%	30.00	min	1.000		std	17.318		.073984		.685696		.000215		
75%	45.00	25%	15.000	000	min	1.000		.000000		.000000		.000000		
max	60.00	50%	30.000000		25%	15.000		.055200		.902600		.050000		
	velocity 75% 45.00000		999	50%	30.000	000 0	.095600	2987	.869800	0	.050000			
count	54001.000000	max	60.000	1000	75%	45.000	000 0	.150000	4404	.093500	0	.050000		
mean	2.940208					60.000	000 0	.483500	5935	.146500	0	.050000		
std	1.242827		velocity	- 4004										
min	0.000000			54001 0		velocity	turn	Ang						
25% 50%	2.008000 2.882000		1.97829 1.27514	9	count	54001.000000	54001.000	999						
75%	3.812000		0.00000	9	mean	2.198162	0.006	611						
max	6.644000	25%	1.03200	9	std	1.479688	0.081	949						
		50%	1.72600	0	min	0.000000	0.000	999						
		75%	2.74600	0	25%	1.104000	0.000	999						
		max	9.76800	1	50%	1.912000	0.000	999						
					75%	3.000000	0.000	999						
					max	9.670000	1.000	999						

DATA



- 1. Timestamp
- 2. Devices_id
- 3. Positions (x, y)
- 4. matchTime_minute
- 5. matchTime_second
- 6. Distance
- 7. totalDistance
- 8. Velocity
- 9. 3Zones
- 10. 18Zones
- 11. Turning Points

Write to CSV

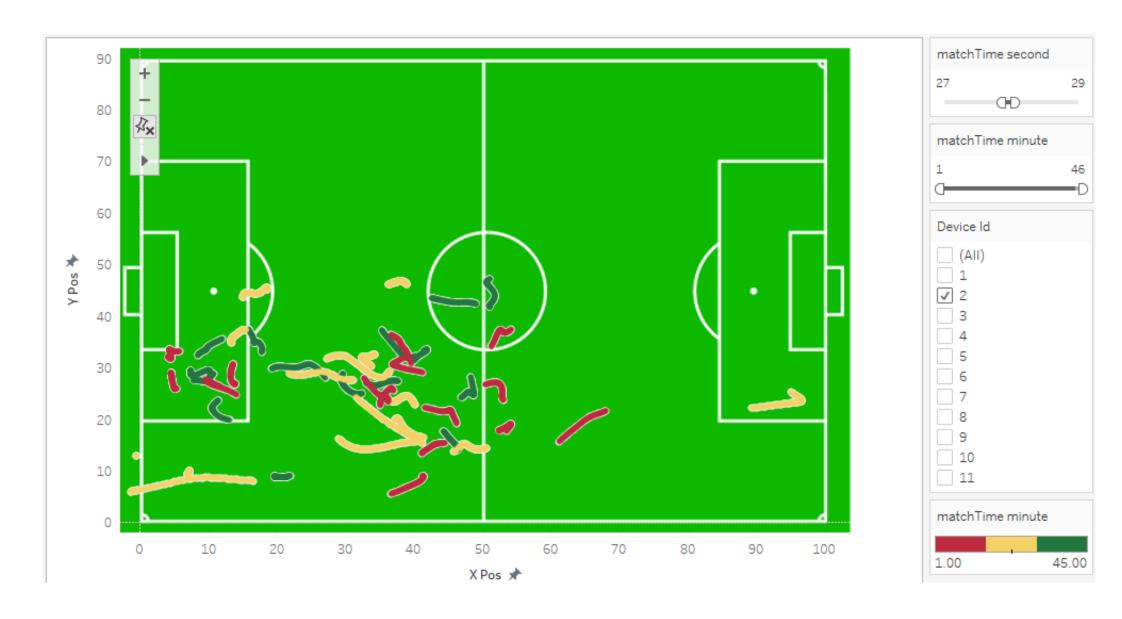
OUTLINE

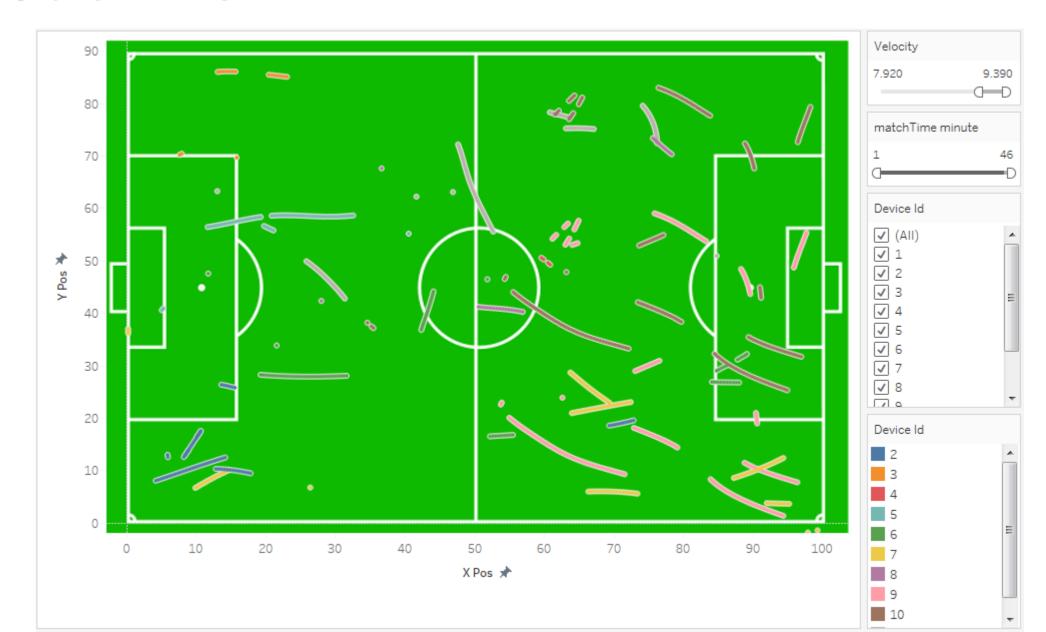
Data analysis with Python

Data visualization with Tableau

The 4-4-2 Formation



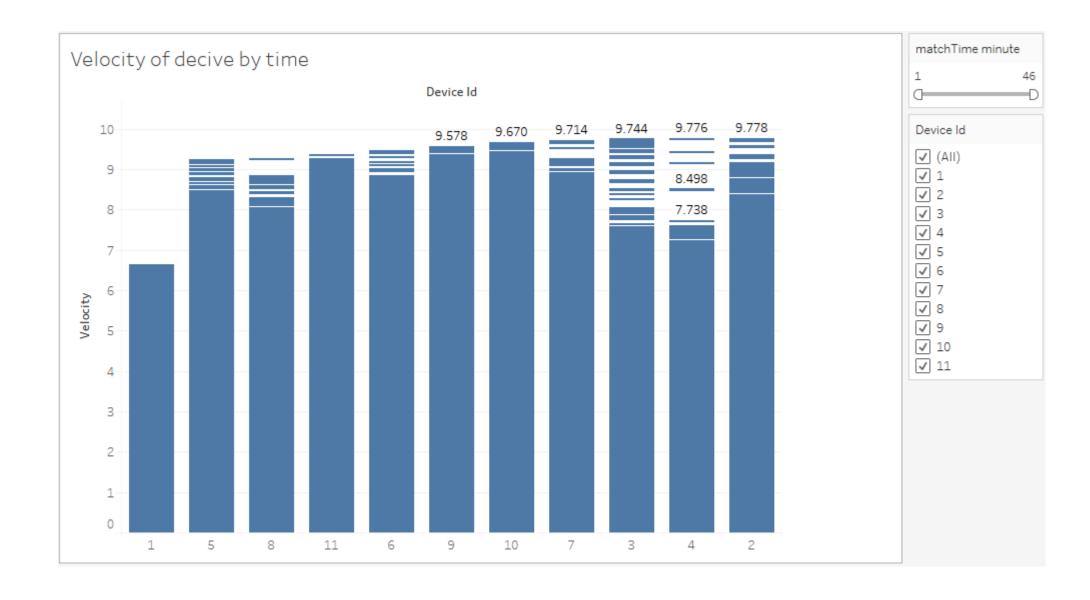


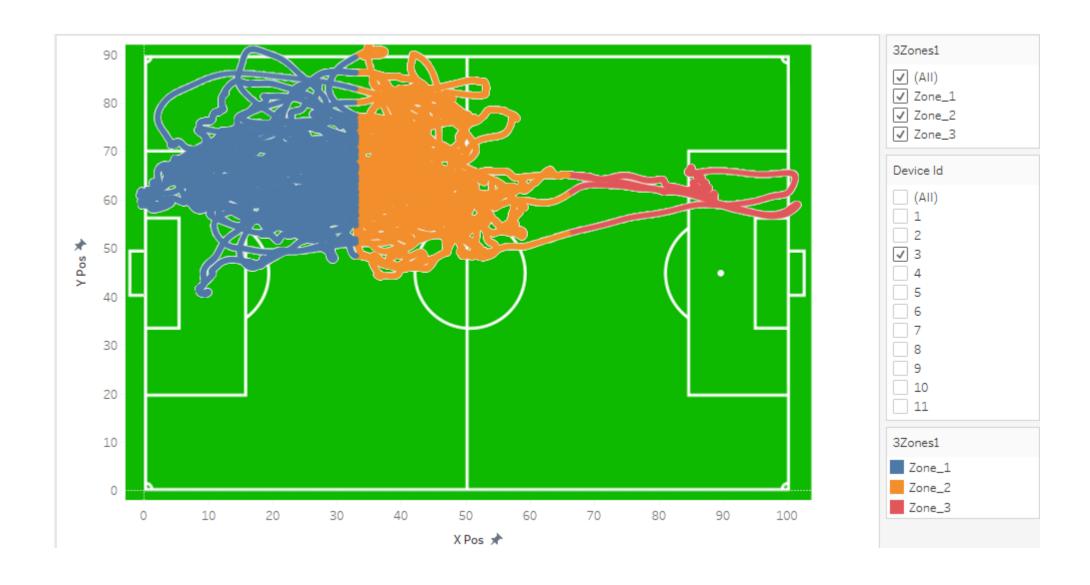


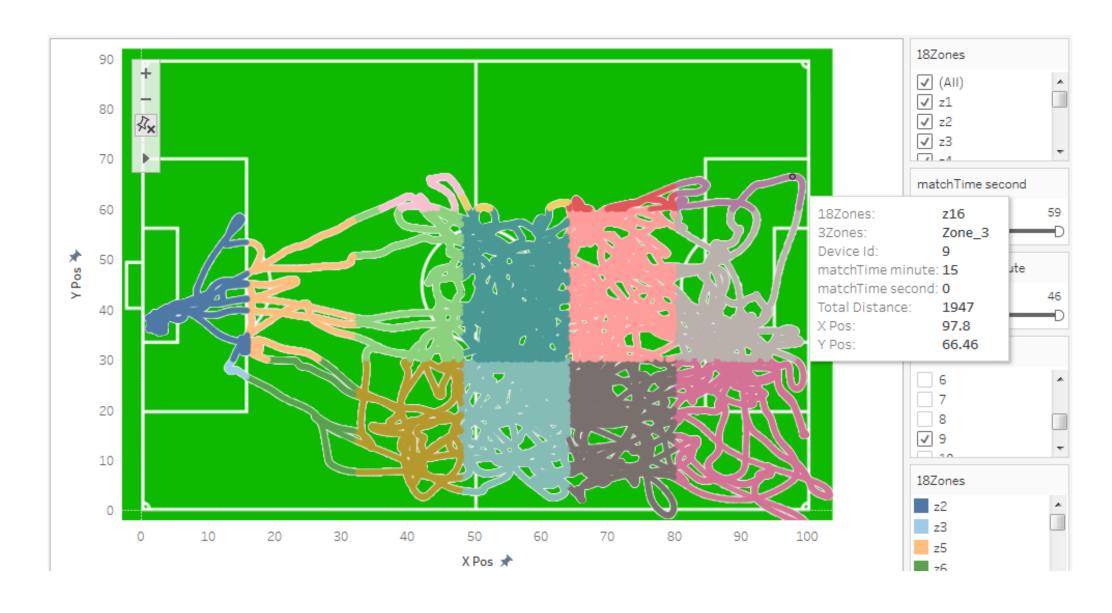
Velocity (>6.05 m/s²)

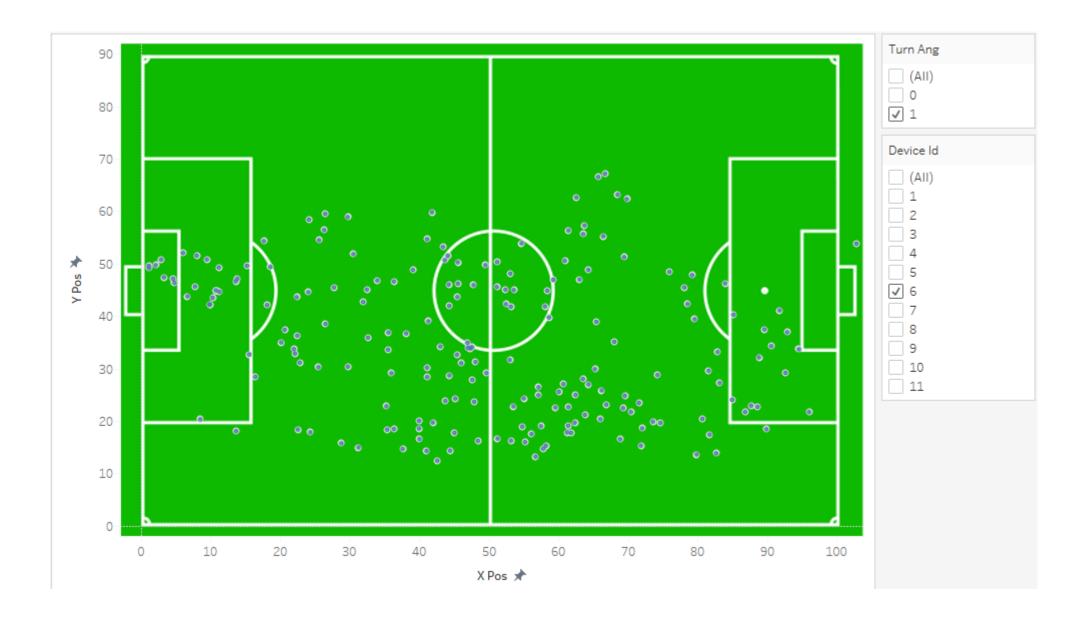
Device_id = 10

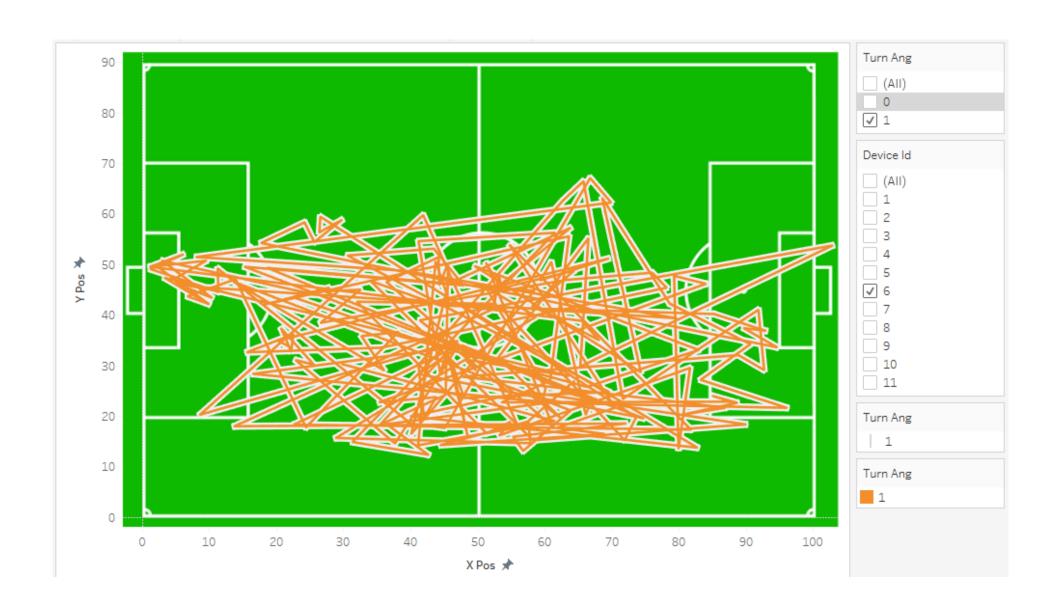


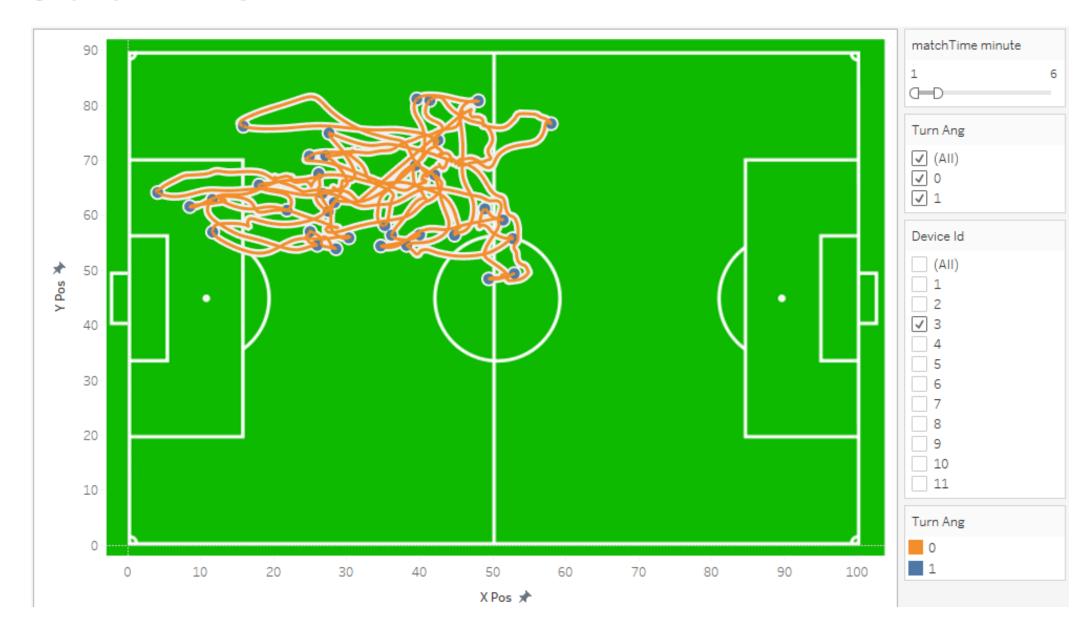


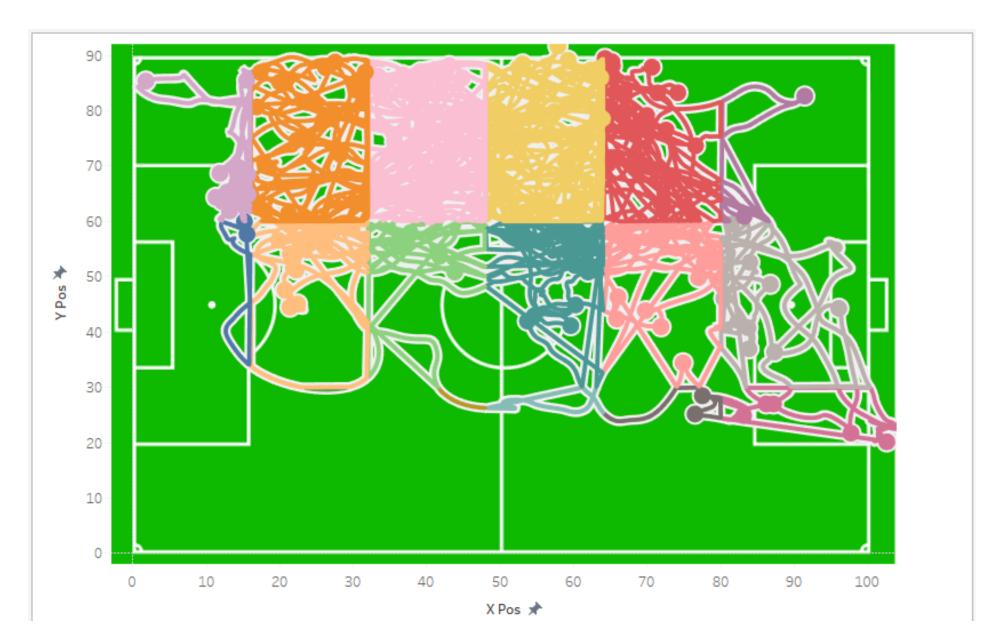


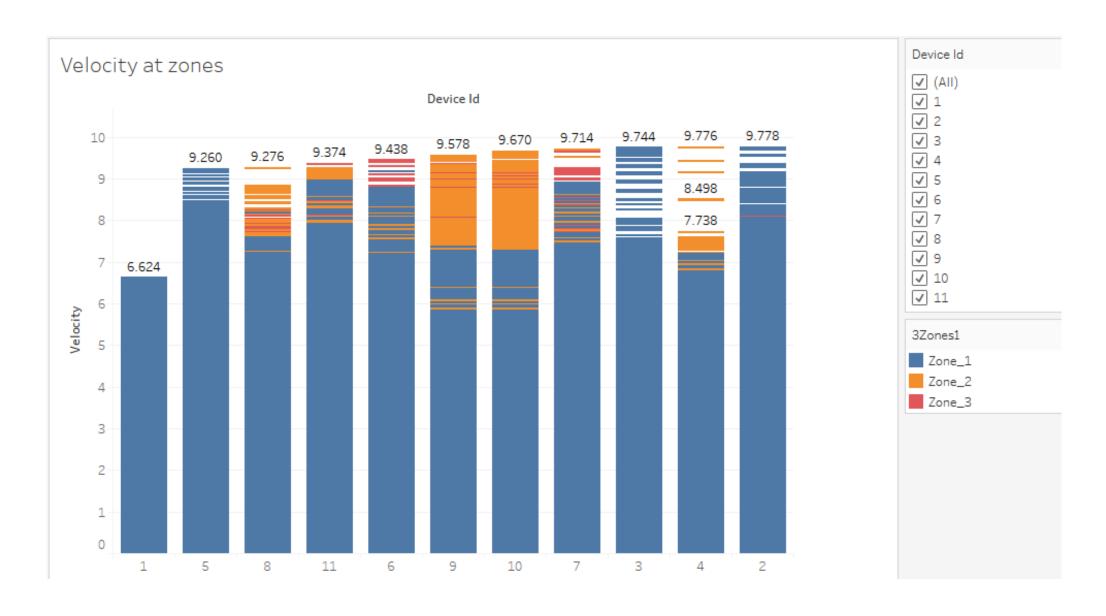












Calculate

- 1. Real Time
- 2. Distance
- 3. Total distance
- 4. Velocity
- 5. Zone
- 6. Turning points

See you in next workshops

- Energy
- > Pass
- Combination
- Create dashboards
- Performance with big Data

. . . .

THANK YOU FOR YOUR ATTENTION







