## **Problem Statement**

Building digital maps is challenging, and maintaining it up to date in an ever-changing world is even more challenging. Various machine learning techniques helps us to detect road signs and changes in real world, and process it to update maps.

The problem presented here is related to a step after detecting a sign on a road. This step has to now identify each road geometry on which this sign is applicable. While sounds like a simple problem, signs in junctions makes this more challenging.

For example, given a sign detected on a road from a 4-camera setting on vehicle, the closest sighting of the sign may be in the right facing camera, with a sharp sign angle with respect to the direction of the car on which cameras set is mounted. Next step for updating map using this sign is to identify the exact road on which this sign is to be placed or applied.

On a + junction, when a sign is detected on the right camera, its hard now to tell if this sign is for the straight road, or for the right-side road, unless you consider parameters like sign bounding box aspect ratio.

For example, a sign detected from Front camera will have a natural aspect ratio of the sign when it is actually facing front of the car, however when same sign is detected on a right-side camera with a sharp angle from front, sign bounding box gets skewed, giving a hint that although its detected in right, it's still facing the front of the car.

Dataset provided here has details on camera sign was detected, Angle of sign with respect to front in degrees, Sign's reported bounding box aspect ratio (width/height), Sign Width and Height, and the target feature Sign Facing, which is where the sign is actually facing.

Goal here is to predict where the sign is actually facing with respect to the vehicle, given above set of inputs

```
In [1]:
        import pandas as pd
        from sklearn.ensemble import RandomForestClassifier
        train = pd.read csv("train.csv")
        test = pd.read csv("test.csv")
```

In [2]: train.head()

Out[2]:

	ld	DetectedCamera	AngleOfSign	SignAspectRatic
0	2c9180975a056a64015a1e0a52e57021	Rear	195	1.02
1	2c9180975a056a64015a1e17b32171e4	Rear	203	1.09
2	2c9180975a056a64015a1de4deb16bd5	Front	26	0.96
3	2c9180975a056a64015a1de4deb16bdd	Rear	199	0.81
4	2c9180975a056a64015a1de4deb16bd6	Rear	208	0.93

```
In [3]: test.head()
```

Out[3]:

```
ld
                                      DetectedCamera | AngleOfSign | SignAspectRatio
0 2c9180975a056a64015a1e10d3f270fe
                                      Right
                                                       67
                                                                    0.63
  2c9180975a056a64015a1de4deb16bdc Front
                                                       16
                                                                    88.0
  2c9180975a056a64015a1e0e70ea70ce
                                      Right
                                                       44
                                                                    1.15
  2c9180975a056a64015a1dfed0c46ec6
                                                       50
                                      Right
                                                                    1.10
  2c9180975a056a64015a1dfed0c46ec7
                                                       30
                                                                    0.95
                                      Front
```

```
In [4]: train['DetectedCamera'].value_counts()
```

```
Out[4]: Front
                  10910
         Right
                  10516
                   9298
         Left
                   7761
         Rear
```

Name: DetectedCamera, dtype: int64

```
In [5]: #encode as integer
        mapping = {'Front':0, 'Right':1, 'Left':2, 'Rear':3}
        train = train.replace({'DetectedCamera':mapping})
        test = test.replace({'DetectedCamera':mapping})
```

```
In [6]: #renaming column
        train.rename(columns = {'SignFacing (Target)': 'Target'}, inplace=True)
```

```
In [7]: #encode Target Variable based on sample submission file
        mapping = {'Front':0, 'Left':1, 'Rear':2, 'Right':3}
        train = train.replace({'Target':mapping})
```

```
In [8]:
        #target variable
        y_train = train['Target']
        test id = test['Id']
```

```
In [9]:
        #drop columns
        train.drop(['Target','Id'], inplace=True, axis=1)
        test.drop('Id',inplace=True,axis=1)
```

```
In [28]: #train model
         #clf = RandomForestClassifier(n estimators=500,max features=3,min samples spli
         t=5,oob_score=True)
         clf=RandomForestClassifier(n estimators = 1200, oob score=True, n jobs =-1,ran
         dom_state =42,
                                     max_features ="auto", min_samples_leaf=1, min_sampl
         es split=50)
         clf.fit(train, y_train)
Out[28]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                     max_depth=None, max_features='auto', max_leaf_nodes=None,
                     min_samples_leaf=1, min_samples_split=50,
                     min weight fraction leaf=0.0, n estimators=1200, n jobs=-1,
                     oob_score=True, random_state=42, verbose=0, warm_start=False)
In [ ]: #predict on test data
         from sklearn.svm import SVC
         model =SVC()
         model.fit(train,y train)
         pred = model.predict_proba(test)
In [37]: #write submission file and submit
         columns = ['Front','Left','Rear','Right']
         sub = pd.DataFrame(data=pred, columns=columns)
         sub['Id'] = test_id
         sub = sub[['Id','Front','Left','Rear','Right']]
         sub.to_csv("sub2.csv",index=False, float_format='%0.6f')
In [10]: | from sklearn.svm import SVC
 In [ ]: # Support Vector Machines
         svc = SVC(probability=True)
         svc.fit(train, y train)
         Y pred = svc.predict(test)
         #acc_svc = round(svc.score(train, y_train) * 100, 2)
         #acc svc
In [ ]:
In [ ]:
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In [ ]:
```