

Flipr Hackathon Solution

```
In [1]: #Importing required libraries
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [2]: %matplotlib inline
```

```
In [3]: from sklearn import metrics
```

```
In [4]: #Reading dataset using Pandas
train = pd.read_excel('Train_dataset.xlsx')
```

```
In [5]: train.head()
```

Out[5]:

	people_ID	Region	Gender	Designation	Name	Married	Children	Occupation	Mode_transport
0	1	Bhubaneswar	Female	Mrs	mansi	YES	1.0	Farmer	Put
1	2	Bhubaneswar	Female	Mrs	riya masi	YES	2.0	Farmer	Wi
2	3	Bhubaneswar	Female	Mrs	sunita	NO	1.0	Cleaner	Put
3	4	Bhubaneswar	Female	Mrs	anjali @ babli	YES	1.0	Driver	C
4	5	Bhubaneswar	Female	Mrs	champa karketta	NO	2.0	Manufacturing	C

5 rows × 28 columns

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

Out[6]: (10714, 28)

```
In [7]: train.columns
```

Out[7]: Index(['people_ID', 'Region', 'Gender', 'Designation', 'Name', 'Married', 'Children', 'Occupation', 'Mode_transport', 'cases/1M', 'Deaths/1M', 'comorbidity', 'Age', 'Coma score', 'Pulmonary score', 'cardiological pressure', 'Diuresis', 'Platelets', 'HBB', 'd-dimer', 'Heart rate', 'HDL cholesterol', 'Charlson Index', 'Blood Glucose', 'Insurance', 'salary', 'FT/month', 'Infect_Prob'], dtype='object')

```
In [8]: train[['cases/1M', 'Deaths/1M',
               'comorbidity', 'Age', 'Coma score', 'Pulmonary score',
               'cardiological pressure', 'Diuresis', 'Platelets']]
```

Out[8]:

	cases/1M	Deaths/1M	comorbidity	Age	Coma score	Pulmonary score	cardiological pressure	Diuresis	Platelets
0	2	0	Hypertension	68	8	<400	Normal	441.0	154.0
1	2	0	Diabetes	64	15	<100	Stage-02	NaN	121.0
2	2	0	None	19	13	<300	Elevated	416.0	124.0
3	2	0	Coronary Heart Disease	33	9	<200	Stage-01	410.0	98.0
4	2	0	Diabetes	23	7	<400	Normal	390.0	21.0
...
10709	8	2	Diabetes	20	14	<400	Normal	134.0	67.0
10710	8	2	None	42	4	<400	Normal	387.0	102.0
10711	8	2	Diabetes	59	3	<100	Stage-02	177.0	111.0
10712	8	2	Coronary Heart Disease	49	6	<300	Elevated	352.0	140.0
10713	8	2	Diabetes	17	7	<400	Normal	181.0	65.0

10714 rows × 9 columns

As parameters like people_ID, Name, Designation are not necessary parameters to calculate Infect_Probabilities, the following cell drops those columns from the dataset.

```
In [9]: train.drop(columns=['people_ID', 'Name', 'Designation'], inplace=True)
```

```
In [10]: #Replacing categorical Married column with numerical values
train['Married'].replace(to_replace=['YES', 'NO'], value=[1, 0], inplace=True)
```

```
In [11]: #Trying to find out which parameters influence Infect_Prob values using correlation function for numerical columns
train.corr()['Infect_Prob']
```

```
Out[11]: Married          -0.465114
Children          0.226795
cases/1M          0.172871
Deaths/1M         0.174994
Age              -0.331258
Coma score        0.038400
Diuresis          0.006887
Platelets         0.066727
HBB              0.019361
d-dimer           0.021304
Heart rate        -0.003647
HDL cholesterol   0.013288
Charlson Index    -0.011368
Blood Glucose     -0.009654
Insurance         0.009996
salary            -0.024621
FT/month          -0.001474
Infect_Prob       1.000000
Name: Infect_Prob, dtype: float64
```

From the above correlation values, the following parameters influence Infect_Prob:

- Married Status
- No of Children
- Age
- cases/1M
- deaths/1M

But for our model, we will remove the last two parameters and include 3 more parameters:

- Coma Score
- Salary
- Platelet Count

which somewhat influence the values of Infect_Prob.

```
In [12]: #Dropping records with null values in them.  
train.dropna(inplace=True)
```

```
In [13]: df = pd.DataFrame(train.corr()['Infect_Prob'])
```

```
In [14]: reqd_param = df[(abs(df['Infect_Prob'])>=0.03) & (df['Infect_Prob']!=1)].index
```

```
In [15]: reqd_param = list(reqd_param)
```

```
In [16]: reqd_param
```

```
Out[16]: ['Married',  
          'Children',  
          'cases/1M',  
          'Deaths/1M',  
          'Age',  
          'Coma score',  
          'Platelets',  
          'salary']
```

```
In [17]: reqd_param.remove('cases/1M')  
reqd_param.remove('Deaths/1M')
```

```
In [18]: #Trying to apply Multiple Linear Regression since Infect_Prob is a continuous value, hence Regression Algorithm  
from sklearn.linear_model import LinearRegression  
reg = LinearRegression()
```

```
In [19]: X = train[reqd_param]  
y = train['Infect_Prob']
```

In [20]: `X.head(10)`

Out[20]:

	Married	Children	Age	Coma score	Platelets	salary
0	1	1.0	68	8	154.0	1300000
2	0	1.0	19	13	124.0	900000
3	1	1.0	33	9	98.0	2300000
4	0	2.0	23	7	21.0	1100000
5	1	1.0	35	9	139.0	1900000
7	1	1.0	49	10	123.0	1200000
9	1	1.0	41	14	23.0	1400000
10	1	1.0	43	9	32.0	1100000
11	1	0.0	52	5	30.0	300000
13	1	2.0	52	14	17.0	2200000

In [21]: `y.head(10)`

Out[21]:

```
0    49.135010
2    73.224000
3    48.779225
4    87.868800
5    49.518345
7    49.121025
9    48.475097
10   46.970339
11   45.494822
13   48.948107
Name: Infect_Prob, dtype: float64
```

In [22]: `# Splitting the ML model into training and testing sets with 10% of the data assigned to testing dataset`
`from sklearn.model_selection import train_test_split`
`X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.1)`

In [23]: `print("X_train set dimension:{}".format(X_train.shape))`
`print("y_train set dimension:{}".format(y_train.shape))`
`print("X_test set dimension:{}".format(X_test.shape))`
`print("y_test set dimension:{}".format(y_test.shape))`

```
X_train set dimension:(6111, 6)
y_train set dimension:(6111,)
X_test set dimension:(680, 6)
y_test set dimension:(680,)
```

In [24]: `reg.fit(X_train,y_train)`

Out[24]: `LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False)`

In [25]: `print(reg.coef_)`

```
[-1.09632701e+01  1.92393370e+00 -2.47987043e-02  1.46632281e-01
 1.74491571e-02 -6.41527044e-07]
```

```
In [26]: #The relation between the parameters and the Infect_Prob value according to
         the model
         print("Infect_Prob = ({})*(Married or Not) + ({})*(No.of Children) + ({})*(Age) + ({})*(Coma Score) + ({})*(Platelet Count) + ({})*(Salary)".format(reg.coef_[0], reg.coef_[1], reg.coef_[2], reg.coef_[3], reg.coef_[4], reg.coef_[5]))
```

```
Infect_Prob = (-10.96327006719004)*(Married or Not) + (1.923933703873164)*(No.of Children) + (-0.024798704315948515)*(Age) + (0.14663228102830123)*(Coma Score) + (0.017449157072205786)*(Platelet Count) + (-6.415270444898437e-07)*(Salary)
```

```
In [27]: y_pred = reg.predict(X_test)
```

```
In [28]: y_pred.shape
```

```
Out[28]: (680,)
```

```
In [29]: #Calculating accuracy of the model using traditional statistics methods
         y_test = list(y_test)
         y_pred = list(y_pred)
         accu = []
         for i in range(len(y_test)):
             acc = 100-((abs(y_test[i]-y_pred[i])/y_test[i])*100)
             accu.append(acc)
```

```
In [30]: #Printing mean accuracy of the model.
         import statistics
         print("Accuracy of the Model:"+str(statistics.mean(accu)))
```

```
Accuracy of the Model:90.13875659535394
```

```
In [33]: y_test[:10]
```

```
Out[33]: [46.16209515,
          49.14719188,
          50.98113046,
          48.81404905,
          52.96879593,
          91.77408,
          48.96265471,
          50.58987314,
          46.31318291,
          48.71294823]
```

```
In [34]: y_pred[:10]
```

```
Out[34]: [47.177921218812784,
          62.099305858213064,
          53.70405894461851,
          49.11238759418449,
          52.4853319201207,
          63.312735991057124,
          49.76701405994029,
          51.032258404358686,
          47.36412448811653,
          60.80446383998665]
```

Using the dataset given and training the model with 6 parameters mentioned above, we can find that our model predicts the values of Infect_Prob at an accuracy of about **90%**, which is considered to be a good model to predict the Infect_Prob values

Now let us predict the values of Test_dataset using the above model

```
In [35]: test = pd.read_excel('Test_dataset.xlsx')
```

```
In [36]: test.head(10)
```

Out[36]:

	people_ID	Region	Gender	Designation	Name	Married	Children	Occupation	Mode_transport	ca
0	5942	Delhi	Female	Mrs	smt rekha prajapat	YES	2	Driver	Public	
1	18664	Delhi	Male	Mr	nirmal	YES	2	Legal	Walk	
2	5603	Delhi	Female	Mrs	pinky	YES	2	Sales	Car	
3	5649	Delhi	Female	Mrs	pooja @aafri	YES	2	Sales	Car	
4	5099	Delhi	Female	Mrs	anjali	YES	2	Business	Car	
5	18749	Delhi	Male	Mr	diwan chand	YES	2	Sales	Walk	
6	5228	Delhi	Female	Mrs	sunita	YES	2	Driver	Car	
7	5559	Delhi	Female	Mrs	gaytri	YES	2	Manufacturing	Walk	
8	5220	Delhi	Female	Mrs	ritu	YES	2	Researcher	Public	
9	5476	Delhi	Female	Mrs	poonam	YES	2	Researcher	Public	

10 rows × 27 columns

```
In [37]: test.drop(columns=['people_ID', 'Region', 'Designation', 'Name'], inplace=True)
```

```
In [38]: test['Married'].replace(to_replace=['YES', 'NO'], value=[1,0], inplace=True)
```

```
In [39]: test.head(10)
```

Out[39]:

	Gender	Married	Children	Occupation	Mode_transport	cases/1M	Deaths/1M	comorbidity	Age	Co sci
0	Female	1	2	Driver	Public	4	1	Diabetes	52	
1	Male	1	2	Legal	Walk	4	1	Diabetes	53	
2	Female	1	2	Sales	Car	4	1	Diabetes	35	
3	Female	1	2	Sales	Car	4	1	None	31	
4	Female	1	2	Business	Car	4	1	Diabetes	51	
5	Male	1	2	Sales	Walk	4	1	Diabetes	34	
6	Female	1	2	Driver	Car	4	1	None	61	
7	Female	1	2	Manufacturing	Walk	4	1	None	55	
8	Female	1	2	Researcher	Public	4	1	Hypertension	28	
9	Female	1	2	Researcher	Public	4	1	Hypertension	55	

10 rows × 23 columns

```
In [40]: test.dropna(inplace=True)
```

```
In [41]: test.shape
```

```
Out[41]: (14498, 23)
```

```
In [42]: reqd_param
```

```
Out[42]: ['Married', 'Children', 'Age', 'Coma score', 'Platelets', 'salary']
```

```
In [43]: X1_test = test[reqd_param]  
X1_test.shape
```

```
Out[43]: (14498, 6)
```

```
In [44]: y1_pred = reg.predict(X1_test)
```

```
In [45]: y1_pred
```

```
Out[45]: array([51.63894912, 51.77515755, 50.5194971 , ..., 46.29255063,  
                46.08116227, 46.3596052 ])
```

```
In [46]: y1_pred = list(y1_pred)
```

```
In [47]: test['Infect_Prob'] = y1_pred
```

```
In [48]: #Writing the final solution in an Excel Format  
test.to_excel('Solution_Sheet.xlsx', sheet_name='Solution')
```

Thank You!!!