Logistic Regression

Classification for 10 Years risk of Coronary heart disease

Data Preparation

In [235…]
```
import numpy as np
import pandas as pd

df = pd.read_csv("framingham_heart_disease.csv")
df.head()
```

Out[235…]

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose	TenYearCHD
0	1	39	4.0	0	0.0	0.0	0	0	0	195.0	106.0	70.0	26.97	80.0	77.0	0
1	0	46	2.0	0	0.0	0.0	0	0	0	250.0	121.0	81.0	28.73	95.0	76.0	0
2	1	48	1.0	1	20.0	0.0	0	0	0	245.0	127.5	80.0	25.34	75.0	70.0	0
3	0	61	3.0	1	30.0	0.0	0	1	0	225.0	150.0	95.0	28.58	65.0	103.0	1
4	0	46	3.0	1	23.0	0.0	0	0	0	285.0	130.0	84.0	23.10	85.0	85.0	0

In [236…]
```
df.isnull().sum()
```

Out[236…]
```
male                0
age                 0
education         105
currentSmoker       0
cigsPerDay         29
BPMeds             53
prevalentStroke     0
prevalentHyp        0
diabetes            0
totChol            50
sysBP               0
diaBP               0
BMI                19
heartRate           1
glucose           388
TenYearCHD          0
dtype: int64
```

In [237…]
```
series = pd.isnull(df['cigsPerDay'])
df[series]
```

Out[237…]

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose	TenYearCHD
131	1	43	4.0	1	NaN	0.0	0	0	0	222.0	109.5	69.0	25.50	75.0	NaN	0
139	1	49	4.0	1	NaN	0.0	0	0	0	256.0	127.5	81.5	28.21	93.0	85.0	1
1046	0	49	1.0	1	NaN	0.0	0	0	0	280.0	120.0	80.0	22.33	90.0	75.0	0
1292	1	42	3.0	1	NaN	0.0	0	0	0	225.0	122.5	80.0	25.54	90.0	90.0	0
1347	0	58	4.0	1	NaN	0.0	0	1	0	270.0	195.0	117.5	23.35	75.0	NaN	0
1451	1	54	1.0	1	NaN	0.0	0	0	0	219.0	110.0	72.0	26.05	95.0	86.0	0
1497	1	55	1.0	1	NaN	0.0	0	0	0	214.0	132.5	85.5	29.25	70.0	103.0	0
1610	0	61	1.0	1	NaN	0.0	0	1	0	356.0	168.0	98.0	27.30	103.0	106.0	0
1625	0	49	2.0	1	NaN	0.0	0	1	0	233.0	158.0	102.0	25.31	90.0	72.0	0
1870	0	47	2.0	1	NaN	0.0	0	0	0	365.0	127.0	76.0	24.44	72.0	80.0	0
1963	1	45	3.0	1	NaN	0.0	0	1	0	170.0	145.5	99.0	26.74	83.0	85.0	0
1980	0	60	2.0	1	NaN	0.0	0	0	0	228.0	112.0	74.0	24.51	63.0	NaN	0
2405	0	49	1.0	1	NaN	0.0	0	0	0	252.0	123.0	69.0	21.45	72.0	89.0	0
2513	1	42	1.0	1	NaN	0.0	0	0	0	226.0	119.0	80.0	25.29	62.0	98.0	0
2542	1	57	NaN	1	NaN	0.0	0	0	0	223.0	107.5	72.5	24.74	62.0	103.0	0
3021	1	53	2.0	1	NaN	0.0	0	0	0	276.0	130.0	86.0	24.21	58.0	82.0	0
3034	0	57	NaN	1	NaN	0.0	0	0	0	229.0	115.0	69.0	24.43	80.0	93.0	0
3094	0	49	1.0	1	NaN	0.0	0	1	0	214.0	172.0	111.0	40.51	80.0	70.0	1
3106	0	47	3.0	1	NaN	0.0	0	0	0	321.0	132.0	88.0	28.14	90.0	74.0	0
3108	1	39	2.0	1	NaN	0.0	0	0	0	285.0	121.0	82.0	27.62	85.0	65.0	0
3156	1	37	3.0	1	NaN	0.0	0	0	0	188.0	123.5	77.0	26.62	65.0	80.0	0
3177	1	45	2.0	1	NaN	0.0	0	0	0	248.0	121.0	72.0	27.88	64.0	88.0	0
3309	1	58	2.0	1	NaN	0.0	0	0	0	235.0	127.5	76.0	21.02	81.0	135.0	0
3432	0	55	1.0	1	NaN	0.0	0	1	0	213.0	163.0	91.0	28.66	69.0	66.0	0
3579	1	42	1.0	1	NaN	0.0	0	0	0	196.0	123.0	73.0	22.06	66.0	NaN	0
3715	0	57	2.0	1	NaN	0.0	0	0	0	270.0	120.0	79.0	24.83	95.0	81.0	0
3847	0	41	1.0	1	NaN	0.0	0	0	0	171.0	135.0	82.5	24.35	79.0	82.0	0
3924	0	52	2.0	1	NaN	0.0	0	1	0	292.0	157.0	112.0	29.56	95.0	84.0	0
3942	0	39	3.0	1	NaN	0.0	0	0	0	197.0	126.5	76.5	19.71	55.0	63.0	0

In [238…]
```
data = df.drop(['currentSmoker','education'], axis = 'columns')
data.head()
```

Out[238…]

	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose	TenYearCHD
0	1	39	0.0	0.0	0	0	0	195.0	106.0	70.0	26.97	80.0	77.0	0
1	0	46	0.0	0.0	0	0	0	250.0	121.0	81.0	28.73	95.0	76.0	0
2	1	48	20.0	0.0	0	0	0	245.0	127.5	80.0	25.34	75.0	70.0	0
3	0	61	30.0	0.0	0	1	0	225.0	150.0	95.0	28.58	65.0	103.0	1
4	0	46	23.0	0.0	0	0	0	285.0	130.0	84.0	23.10	85.0	85.0	0

Fill NaN for cigarettes per Day

In [239…]
```
cigs = data['cigsPerDay']
cigs.head()
```

Out[239…]
```
0     0.0
1     0.0
2    20.0
3    30.0
4    23.0
Name: cigsPerDay, dtype: float64
```

In [240…]
```
cig = cigs.mean()
```

In [241…]
```
import math
integer_value = math.floor(cig)
```

```
cigs.fillna(integer_value, inplace = True)
data.isnull().sum()
```

Out[241…
```
male               0
age                0
cigsPerDay         0
BPMeds            53
prevalentStroke    0
prevalentHyp       0
diabetes           0
totChol           50
sysBP              0
diaBP              0
BMI               19
heartRate          1
glucose          388
TenYearCHD         0
dtype: int64
```

In [242…
```
data.dropna( axis = 0, inplace = True)
data.isnull().sum()
```

Out[242…
```
male             0
age              0
cigsPerDay       0
BPMeds           0
prevalentStroke  0
prevalentHyp     0
diabetes         0
totChol          0
sysBP            0
diaBP            0
BMI              0
heartRate        0
glucose          0
TenYearCHD       0
dtype: int64
```

Data Exploration

In [243…
```
df.mean()
```

Out[243…
```
male               0.429212
age               49.584946
education          1.978950
currentSmoker      0.494101
cigsPerDay         9.003089
BPMeds             0.029630
prevalentStroke    0.005899
prevalentHyp       0.310524
diabetes           0.025720
totChol          236.721585
sysBP            132.352407
diaBP             82.893464
BMI               25.802008
heartRate         75.878924
glucose           81.966753
TenYearCHD         0.151958
dtype: float64
```

In [244…
```
df.var()
```

Out[244…
```
male               0.245047
age               73.481926
education          1.039973
currentSmoker      0.250024
cigsPerDay       142.088631
BPMeds             0.028759
prevalentStroke    0.005866
prevalentHyp       0.214149
diabetes           0.025064
totChol         1988.297915
sysBP            485.677704
diaBP            141.868338
BMI               16.647306
heartRate        144.639020
glucose          574.081513
TenYearCHD         0.128898
dtype: float64
```

In [245…
```
df.skew()
```

Out[245…
```
male               0.286135
age                0.228146
education          0.690033
currentSmoker      0.023606
cigsPerDay         1.247910
BPMeds             5.550010
prevalentStroke   12.909062
prevalentHyp       0.819278
diabetes           5.994378
totChol            0.871422
sysBP              1.145362
diaBP              0.714102
BMI                0.981974
heartRate          0.644482
glucose            6.213402
TenYearCHD         1.939741
dtype: float64
```

In [246…
```
df.kurtosis()
```

Out[246…
```
male               -1.919033
age                -0.989636
education          -0.712394
currentSmoker      -2.000387
cigsPerDay          1.023356
BPMeds             28.816384
prevalentStroke   164.721624
prevalentHyp       -1.329411
diabetes           33.948587
totChol             4.131582
sysBP               2.155019
diaBP               1.277100
BMI                 2.656839
heartRate           0.907483
glucose            58.674278
TenYearCHD          1.763428
dtype: float64
```

In [247…
```
df.min()
```

Out[247…
```
male               0.00
age               32.00
education          1.00
currentSmoker      0.00
cigsPerDay         0.00
BPMeds             0.00
prevalentStroke    0.00
prevalentHyp       0.00
diabetes           0.00
totChol          107.00
sysBP             83.50
diaBP             48.00
BMI               15.54
heartRate         44.00
glucose           40.00
TenYearCHD         0.00
dtype: float64
```

In [248…
```
df.max()
```

Out[248…
```
male                1.0
age                70.0
education           4.0
currentSmoker       1.0
cigsPerDay         70.0
BPMeds              1.0
prevalentStroke     1.0
prevalentHyp        1.0
diabetes            1.0
totChol           696.0
sysBP             295.0
diaBP             142.5
BMI                56.8
heartRate         143.0
glucose           394.0
TenYearCHD          1.0
dtype: float64
```

In [249…
```
df.median()
```

Out[249…
```
male                0.0
age                49.0
education           2.0
currentSmoker       0.0
cigsPerDay          0.0
BPMeds              0.0
prevalentStroke     0.0
prevalentHyp        0.0
diabetes            0.0
totChol           234.0
sysBP             128.0
diaBP              82.0
BMI                25.4
heartRate          75.0
glucose            78.0
TenYearCHD          0.0
dtype: float64
```

In [250…
```
df.corr()
```

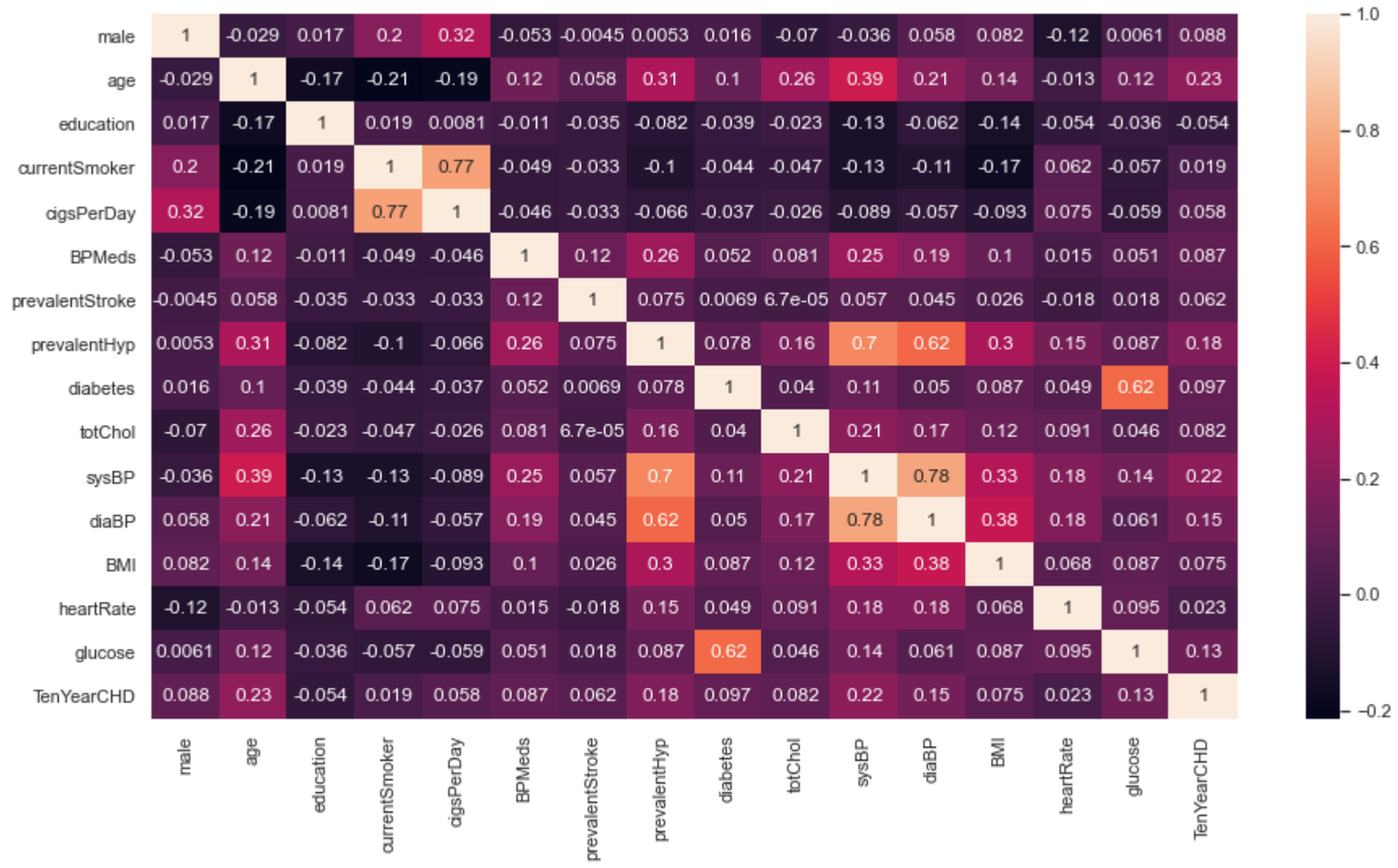
Out[250…

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose	TenYearCHD
male	1.000000	-0.028979	0.017352	0.197596	0.317930	-0.052506	-0.004546	0.005313	0.015708	-0.070322	-0.035989	0.057933	0.081672	-0.116620	0.006083	0.088428
age	-0.028979	1.000000	-0.165883	-0.213748	-0.192791	0.122995	0.057655	0.307194	0.101258	0.262131	0.394302	0.206104	0.135800	-0.012823	0.122256	0.225256
education	0.017352	-0.165883	1.000000	0.018532	0.008085	-0.010815	-0.035112	-0.081970	-0.038680	-0.023115	-0.129631	-0.062316	-0.137504	-0.054206	-0.035721	-0.054059
currentSmoker	0.197596	-0.213748	0.018532	1.000000	0.769690	-0.048938	-0.032988	-0.103260	-0.044295	-0.046562	-0.130230	-0.107746	-0.167650	0.062356	-0.056826	0.019456
cigsPerDay	0.317930	-0.192791	0.008085	0.769690	1.000000	-0.046134	-0.032707	-0.066146	-0.037067	-0.026320	-0.088780	-0.056632	-0.092856	0.075157	-0.058960	0.057884
BPMeds	-0.052506	0.122995	-0.010815	-0.048938	-0.046134	1.000000	0.117365	0.261187	0.052047	0.080558	0.254219	0.194227	0.100668	0.015233	0.051176	0.087489
prevalentStroke	-0.004546	0.057655	-0.035112	-0.032988	-0.032707	0.117365	1.000000	0.074830	0.006949	0.000067	0.057009	0.045190	0.025891	-0.017676	0.018431	0.061810
prevalentHyp	0.005313	0.307194	-0.081970	-0.103260	-0.066146	0.261187	0.074830	1.000000	0.077808	0.163993	0.696755	0.615751	0.301318	0.147261	0.086834	0.177603
diabetes	0.015708	0.101258	-0.038680	-0.044295	-0.037067	0.052047	0.006949	0.077808	1.000000	0.040278	0.111283	0.050329	0.087036	0.048994	0.617627	0.097317
totChol	-0.070322	0.262131	-0.023115	-0.046562	-0.026320	0.080558	0.000067	0.163993	0.040278	1.000000	0.208908	0.165182	0.115767	0.091125	0.046408	0.082184
sysBP	-0.035989	0.394302	-0.129631	-0.130230	-0.088780	0.254219	0.057009	0.696755	0.111283	0.208908	1.000000	0.784002	0.326981	0.182246	0.140621	0.216429
diaBP	0.057933	0.206104	-0.062316	-0.107746	-0.056632	0.194227	0.045190	0.615751	0.050329	0.165182	0.784002	1.000000	0.377588	0.181255	0.061231	0.145299
BMI	0.081672	0.135800	-0.137504	-0.167650	-0.092856	0.100668	0.025891	0.301318	0.087036	0.115767	0.326981	0.377588	1.000000	0.067678	0.087377	0.075192
heartRate	-0.116620	-0.012823	-0.054206	0.062356	0.075157	0.015233	-0.017676	0.147261	0.048994	0.091125	0.182246	0.181255	0.067678	1.000000	0.094500	0.022913
glucose	0.006083	0.122256	-0.035721	-0.056826	-0.058960	0.051176	0.018431	0.086834	0.617627	0.046408	0.140621	0.061231	0.087377	0.094500	1.000000	0.125544
TenYearCHD	0.088428	0.225256	-0.054059	0.019456	0.057884	0.087489	0.061810	0.177603	0.097317	0.082184	0.216429	0.145299	0.075192	0.022913	0.125544	1.000000

In [251…
```
import seaborn as sns
```

In [252…
```
sns.set(rc = {'figure.figsize':(15,8)})

sns.heatmap(df.corr(), annot=True,  annot_kws={'size':12 })
```

Out[252… <AxesSubplot:>

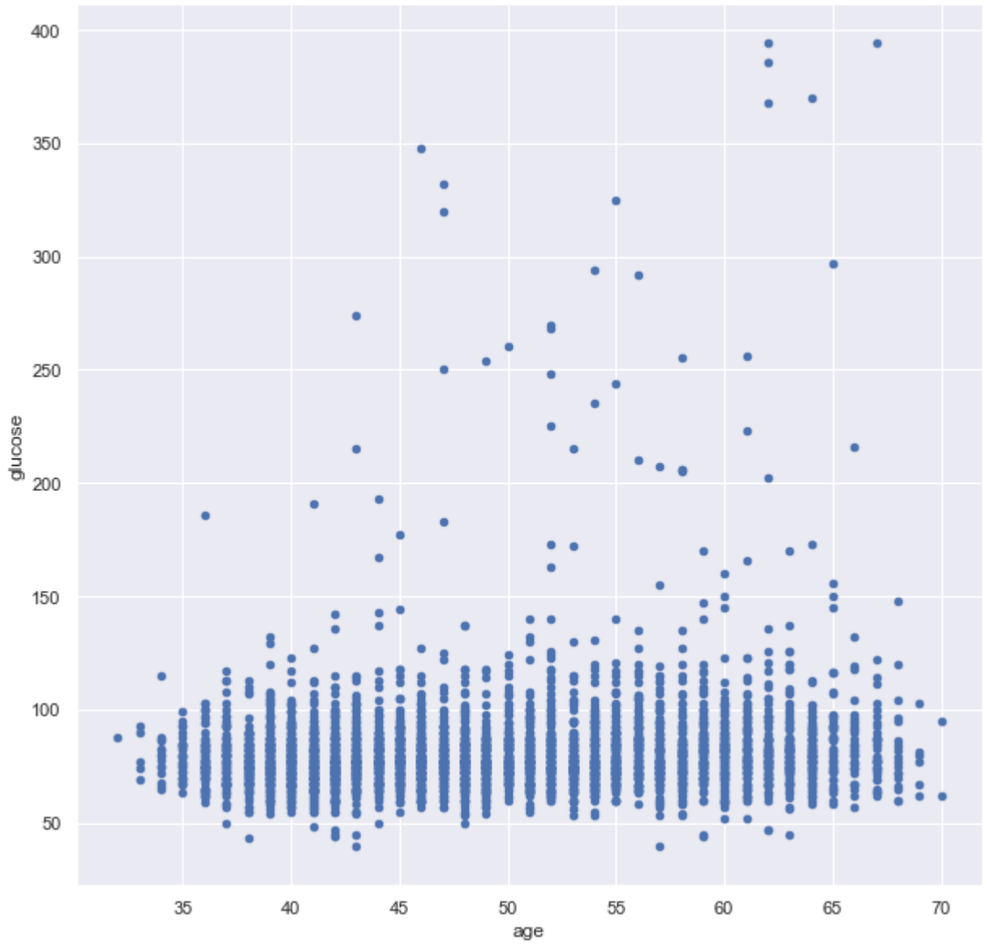


Plots

In [253…
```
df.plot.scatter('age', 'glucose',figsize=(10, 10))
```

```
*c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*.  Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.
```

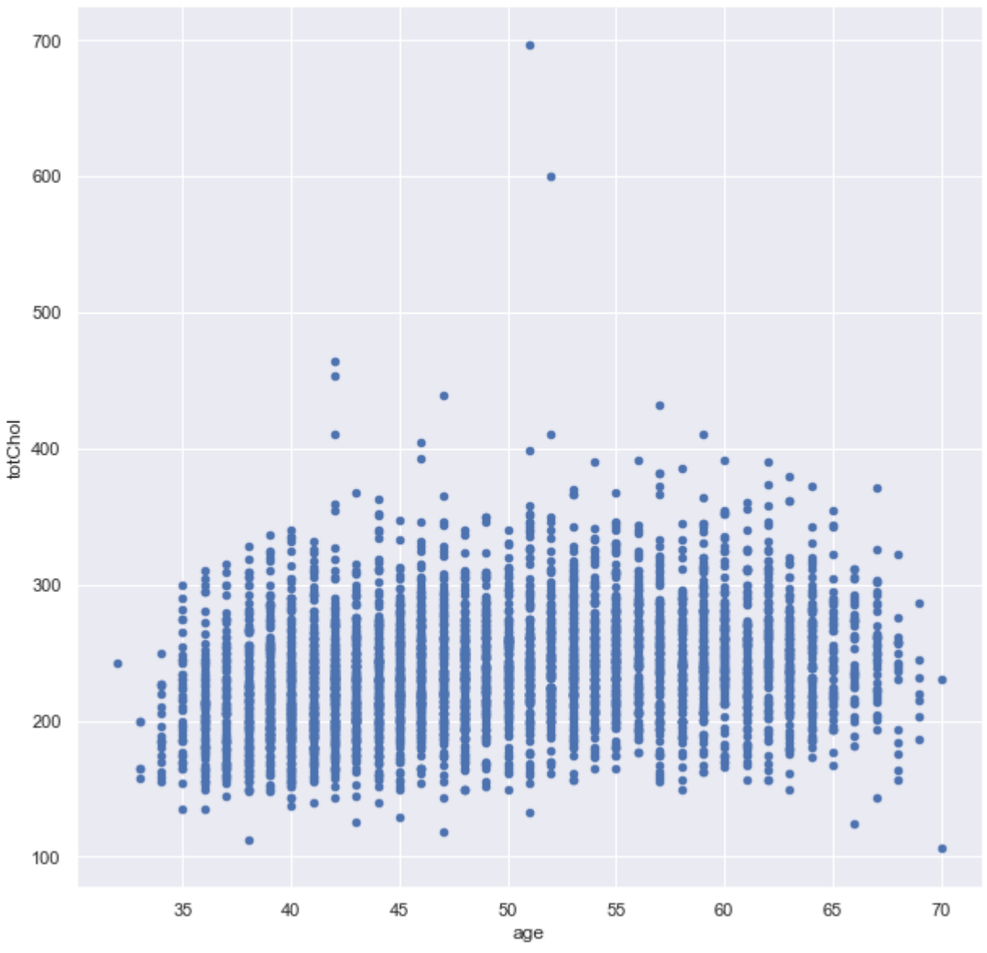
Out[253… <AxesSubplot:xlabel='age', ylabel='glucose'>



In [254…
```
df.plot.scatter('age', 'totChol',figsize=(10, 10))
```

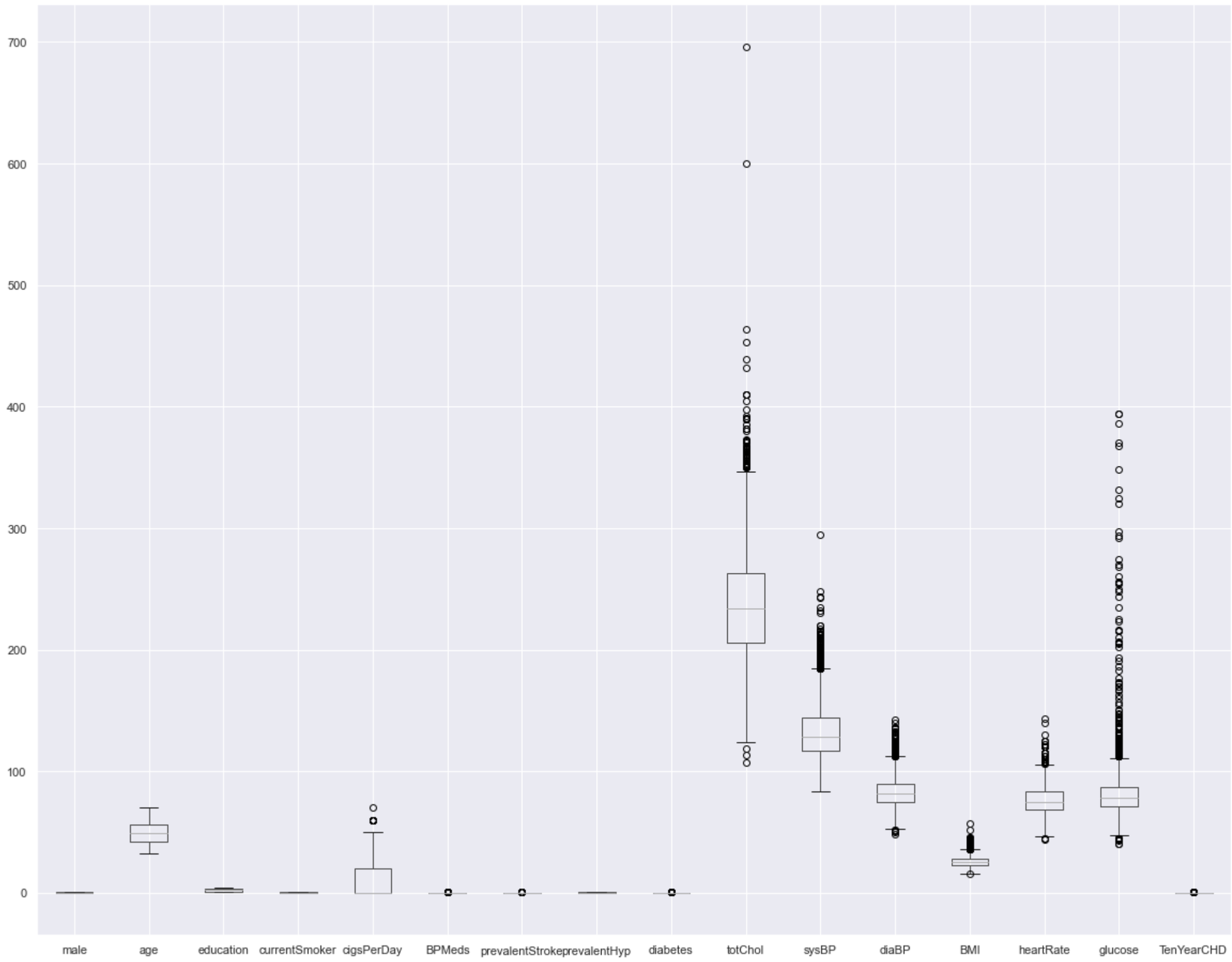
```
*c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with *x* & *y*.  Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all points.
```

Out[254…
```
<AxesSubplot:xlabel='age', ylabel='totChol'>
```



In [255…
```
df.boxplot(figsize=(20, 16))
```

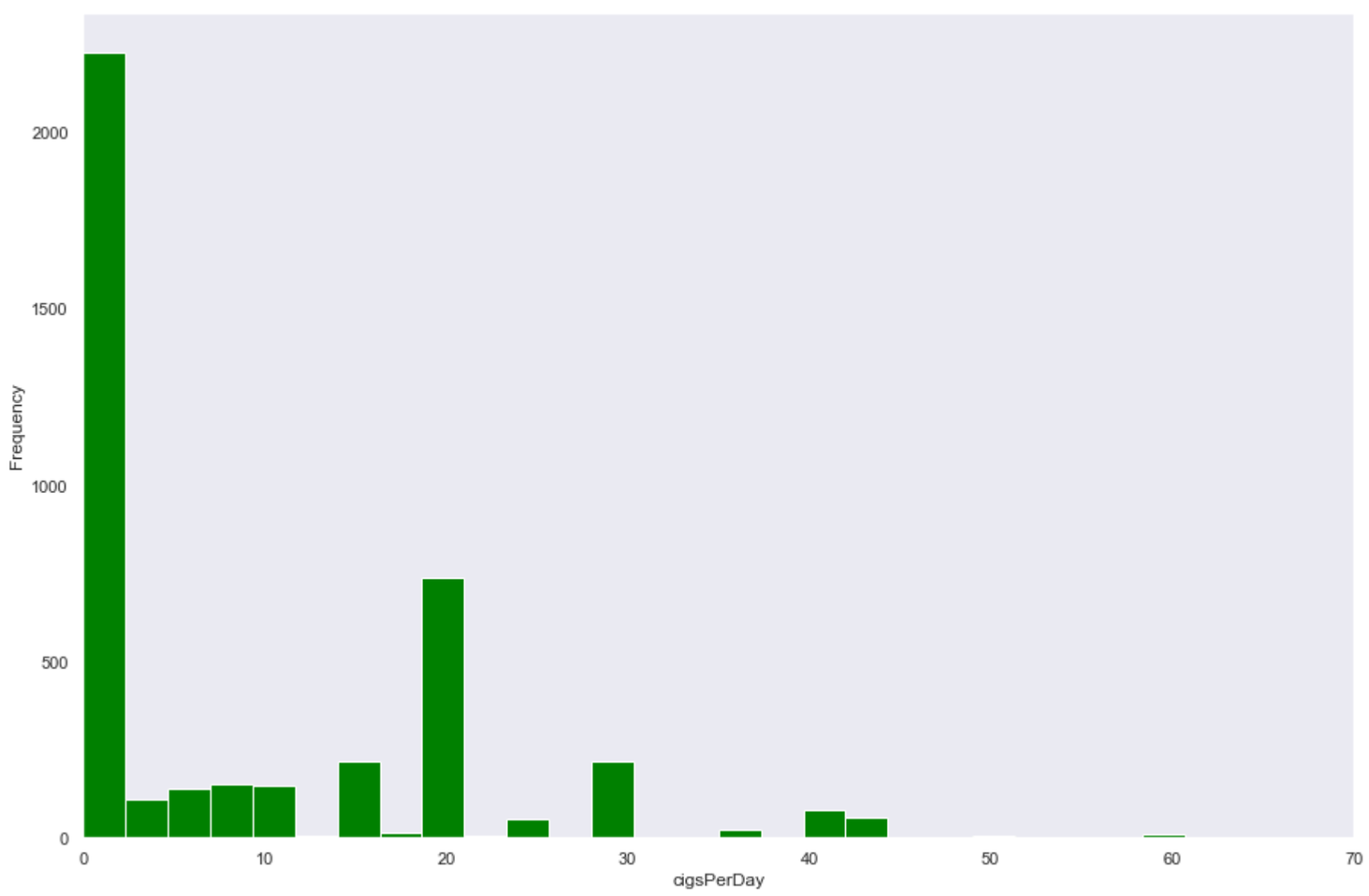
Out[255…
```
<AxesSubplot:>
```



In [256…

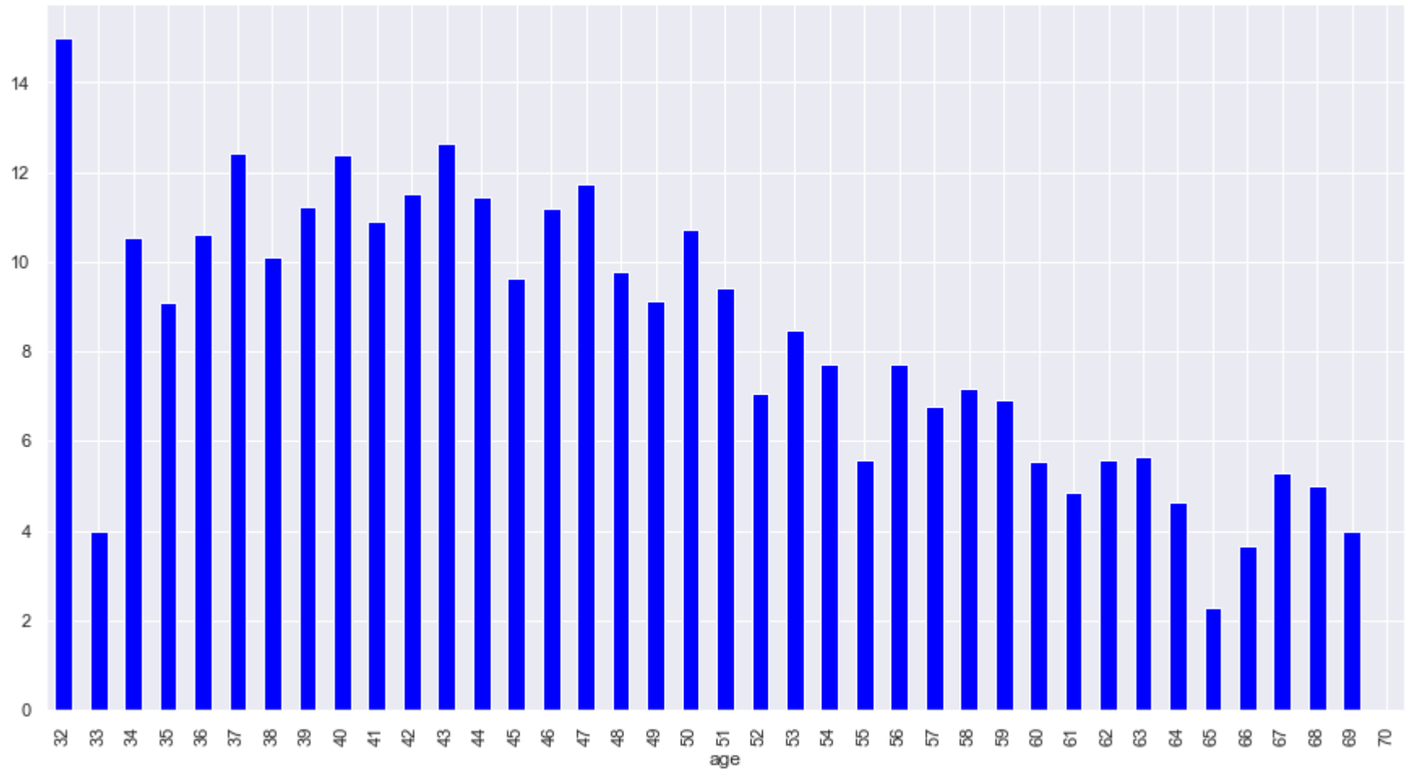
```
ax = df['cigsPerDay'].hist(bins=30, grid=False, color='green', figsize=(15, 10))
ax.set_xlabel('cigsPerDay')
ax.set_ylabel('Frequency')

ax.set_xlim(0, 70)
#ax.set_ylim(0, 120)
```

Out[256… (0.0, 70.0)



In [257…

```
df_avg_Cig = df.groupby('age')['cigsPerDay'].mean()
df_avg_Cig.plot.bar(color='blue');
```



CHD Stats

In [258…
```
Heart_Attack = data[data.TenYearCHD == 1]
Heart_Attack.head()
```

Out[258…

	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose	TenYearCHD
3	0	61	30.0	0.0	0	1	0	225.0	150.0	95.0	28.58	65.0	103.0	1
6	0	63	0.0	0.0	0	0	0	205.0	138.0	71.0	33.11	60.0	85.0	1
15	0	38	20.0	0.0	0	1	0	221.0	140.0	90.0	21.35	95.0	70.0	1
17	0	46	20.0	0.0	0	0	0	291.0	112.0	78.0	23.38	80.0	89.0	1
25	1	47	20.0	0.0	0	0	0	294.0	102.0	68.0	24.18	62.0	66.0	1

In [259…
```
Heart_Attack = data[data.TenYearCHD == 1]
Heart_Attack.head()
```

Out[259…

	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose	TenYearCHD
3	0	61	30.0	0.0	0	1	0	225.0	150.0	95.0	28.58	65.0	103.0	1
6	0	63	0.0	0.0	0	0	0	205.0	138.0	71.0	33.11	60.0	85.0	1
15	0	38	20.0	0.0	0	1	0	221.0	140.0	90.0	21.35	95.0	70.0	1
17	0	46	20.0	0.0	0	0	0	291.0	112.0	78.0	23.38	80.0	89.0	1
25	1	47	20.0	0.0	0	0	0	294.0	102.0	68.0	24.18	62.0	66.0	1

In [260…
```
No_Heart_Attack = data[data.TenYearCHD == 0]
No_Heart_Attack.head()
```

Out[260…

	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose	TenYearCHD
0	1	39	0.0	0.0	0	0	0	195.0	106.0	70.0	26.97	80.0	77.0	0
1	0	46	0.0	0.0	0	0	0	250.0	121.0	81.0	28.73	95.0	76.0	0
2	1	48	20.0	0.0	0	0	0	245.0	127.5	80.0	25.34	75.0	70.0	0
4	0	46	23.0	0.0	0	0	0	285.0	130.0	84.0	23.10	85.0	85.0	0
5	0	43	0.0	0.0	0	1	0	228.0	180.0	110.0	30.30	77.0	99.0	0

In [261…
```
data.groupby('TenYearCHD').mean()
```

Out[261…

TenYearCHD	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	diaBP	BMI	heartRate	glucose
0	0.425312	48.740313	8.723750	0.024063	0.004063	0.276250	0.020625	235.347187	130.308437	82.182969	25.662769	75.609688	80.665000
1	0.557491	54.233449	10.574913	0.064460	0.013937	0.506969	0.062718	246.311847	143.827526	87.161150	26.644094	76.378049	88.827526

In [262…
```
final = data.drop(['diaBP','BMI','heartRate'], axis = 'columns')
```

In [263…
```
No_Heart_Attack = final[final.TenYearCHD == 0]
No_Heart_Attack.head()
```

Out[263…

	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	glucose	TenYearCHD
0	1	39	0.0	0.0	0	0	0	195.0	106.0	77.0	0
1	0	46	0.0	0.0	0	0	0	250.0	121.0	76.0	0
2	1	48	20.0	0.0	0	0	0	245.0	127.5	70.0	0
4	0	46	23.0	0.0	0	0	0	285.0	130.0	85.0	0
5	0	43	0.0	0.0	0	1	0	228.0	180.0	99.0	0

In [264…
```
Heart_Attack = final[final.TenYearCHD == 1]
Heart_Attack.head()
```

Out[264…

	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	glucose	TenYearCHD
3	0	61	30.0	0.0	0	1	0	225.0	150.0	103.0	1
6	0	63	0.0	0.0	0	0	0	205.0	138.0	85.0	1
15	0	38	20.0	0.0	0	1	0	221.0	140.0	70.0	1
17	0	46	20.0	0.0	0	0	0	291.0	112.0	89.0	1
25	1	47	20.0	0.0	0	0	0	294.0	102.0	66.0	1

In [265…
```
final.groupby('TenYearCHD').mean()
```

Out[265…

TenYearCHD	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	glucose
0	0.425312	48.740313	8.723750	0.024063	0.004063	0.276250	0.020625	235.347187	130.308437	80.665000
1	0.557491	54.233449	10.574913	0.064460	0.013937	0.506969	0.062718	246.311847	143.827526	88.827526

In [266…
```
X = final[['male','age','cigsPerDay','BPMeds','prevalentStroke','prevalentHyp','diabetes','totChol','sysBP','glucose']]
y = final['TenYearCHD']
```

In [267…
```
X
```

Out[267…

	male	age	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	glucose
0	1	39	0.0	0.0	0	0	0	195.0	106.0	77.0
1	0	46	0.0	0.0	0	0	0	250.0	121.0	76.0
2	1	48	20.0	0.0	0	0	0	245.0	127.5	70.0
3	0	61	30.0	0.0	0	1	0	225.0	150.0	103.0
4	0	46	23.0	0.0	0	0	0	285.0	130.0	85.0
...	...	...	...	...	...	...	...	...	...	...
4231	1	58	0.0	0.0	0	1	0	187.0	141.0	81.0
4232	1	68	0.0	0.0	0	1	0	176.0	168.0	79.0
4233	1	50	1.0	0.0	0	1	0	313.0	179.0	86.0
4234	1	51	43.0	0.0	0	0	0	207.0	126.5	68.0
4237	0	52	0.0	0.0	0	0	0	269.0	133.5	107.0

3774 rows × 10 columns

In [268…
```
y
```

Out[268…
```
0       0
1       0
2       0
3       1
4       0
       ..
4231    0
4232    1
4233    1
4234    0
4237    0
Name: TenYearCHD, Length: 3774, dtype: int64
```

Model Training

In [269…
```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size = 0.20, random_state = 99)
```

In [270…
```
from sklearn.linear_model import LogisticRegression
model = LogisticRegression(max_iter = 1000, C = 1)
```

In [271…
```
model.fit(X_train, y_train)
print("Accuracy:", model.score(X_test,y_test))
```

```
Accuracy: 0.8649006622516556
```

Confusion Matrix

In [272…
```
y_pred = model.predict(X_test)

from sklearn.metrics import confusion_matrix
matrix = confusion_matrix(y_test,y_pred)
matrix
```

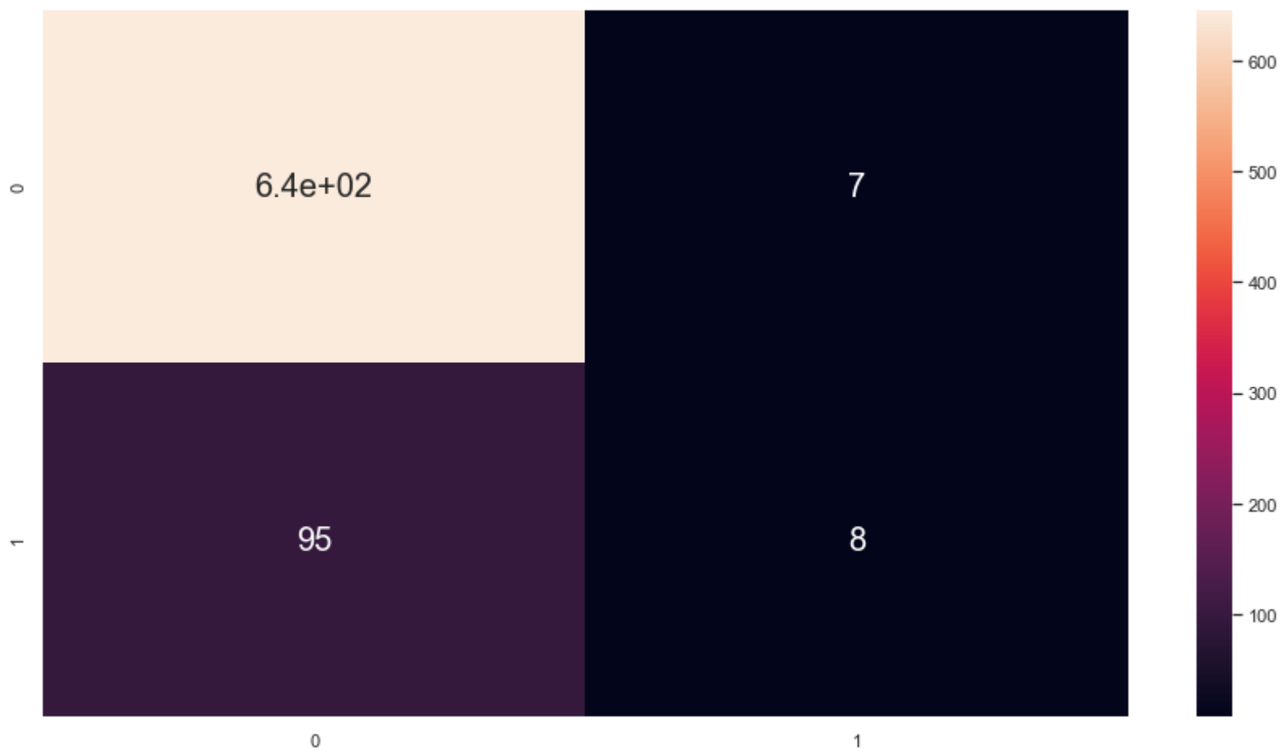
Out[272…
```
array([[645,   7],
       [ 95,   8]], dtype=int64)
```

In [273…
```
print(len(y_test),len(y_pred))
print(X_test.size)
```

```
755 755
7550
```

In [274…
```
sns.heatmap(matrix, annot = True, annot_kws={'size':20 })
```

Out[274…
```
<AxesSubplot:>
```



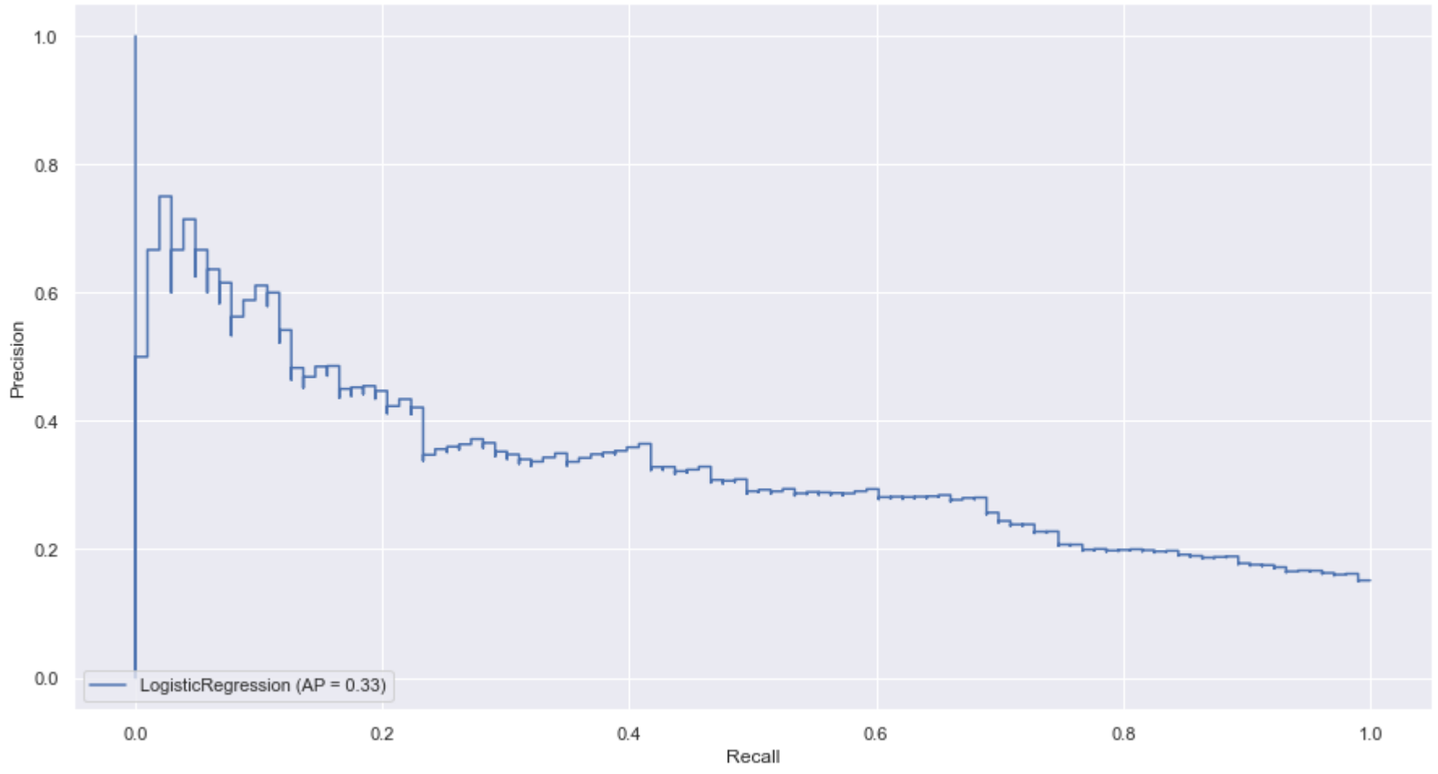
Precision Recall

In [275…
```
from sklearn.metrics import recall_score, precision_score

print("Recall: ", recall_score(y_test, y_pred))
print("Precision: ", precision_score(y_test, y_pred))
```

```
Recall:  0.07766990291262135
Precision:  0.5333333333333333
```

In [276…
```
from sklearn.metrics import plot_precision_recall_curve

plot_precision_recall_curve(model, X_test, y_test)
```

Out[276…
```
<sklearn.metrics._plot.precision_recall_curve.PrecisionRecallDisplay at 0x231f2ca1b80>
```



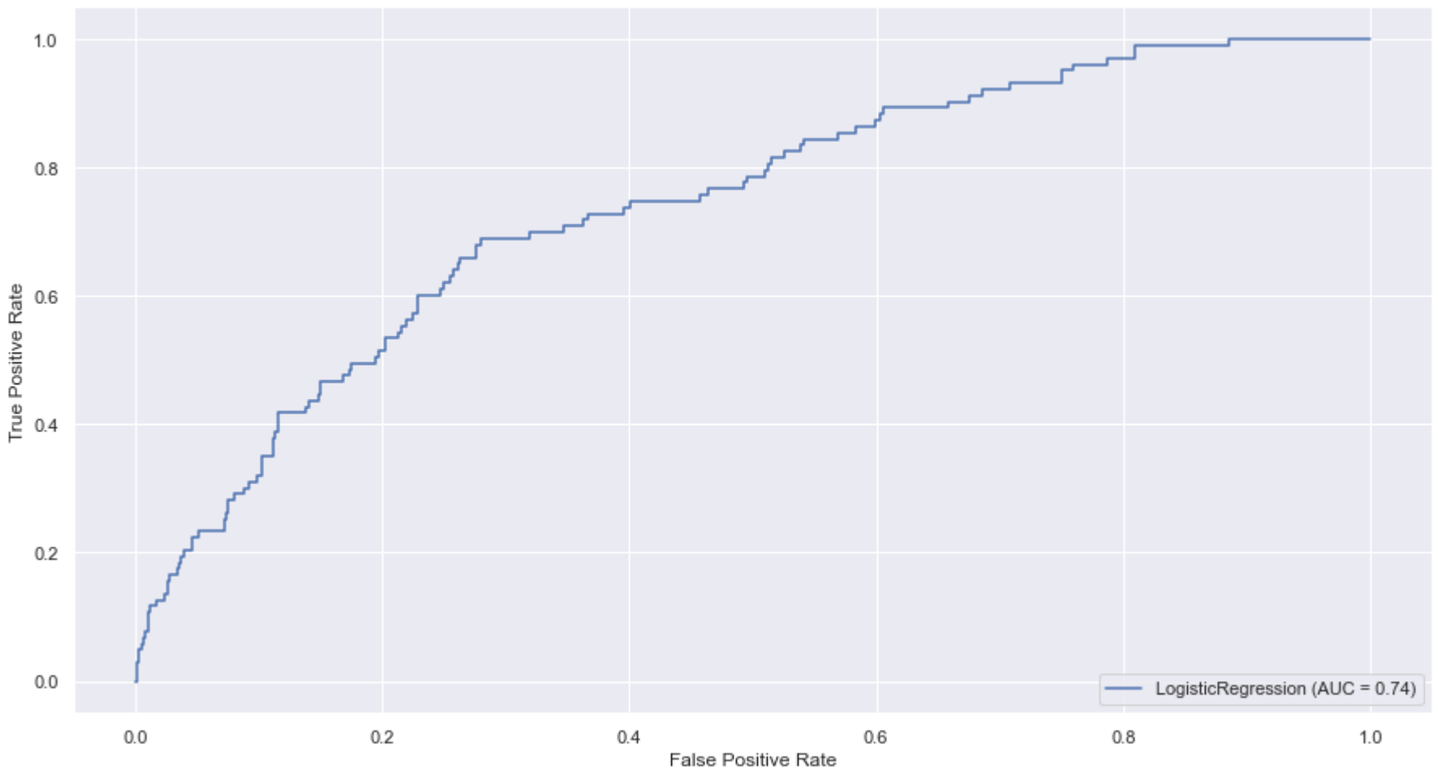
ROC Curve

In [277…
```
from sklearn.metrics import plot_roc_curve, roc_curve, auc

fpr_lr, tpr_lr, _ = roc_curve(y_test, y_pred)
roc_auc = auc(fpr_lr, tpr_lr)

roc_auc_lr
```

Out[277…
```
0.5334668532968014
```

In [278…
```
plot_roc_curve(model, X_test, y_test)
```

Out[278…
```
<sklearn.metrics._plot.roc_curve.RocCurveDisplay at 0x231f19687c0>
```



Using k-Fold Cross Validation

In [279…
```
from sklearn.model_selection import KFold
from sklearn.metrics import accuracy_score
```

In [280…
```
k = 10
kf = KFold(n_splits=k, random_state=None)
#model = LogisticRegression(solver= 'liblinear')

acc_score = []

for train_index , test_index in kf.split(X):

    X_train , X_test = X.iloc[train_index,:],X.iloc[test_index,:]
    y_train , y_test = y.iloc[train_index] , y.iloc[test_index]

    model.fit(X_train,y_train)
    pred_values = model.predict(X_test)

    acc = accuracy_score(pred_values , y_test)
    acc_score.append(acc)

avg_acc_score = sum(acc_score)/k

print('accuracy of each fold - {}'.format(acc_score))
print('Avg accuracy : {}'.format(avg_acc_score))
```

```
accuracy of each fold - [0.8121693121693122, 0.8571428571428571, 0.8624338624338624, 0.8835978835978836, 0.843501326259947, 0.8567639257294429, 0.8647214854111406, 0.8673740053050398, 0.8514588859
416445, 0.8355437665782494]
Avg accuracy : 0.853470731056938
```