Glucose 0.129459 1.000000 BloodPressure 0.141282 0.152590 SkinThickness -0.081672 0.057328 Insulin -0.073535 0.331357 BMI 0.017683 0.221071 DiabetesPedigreeFunction -0.033523 0.137337 Age 0.544341 0.263514	19.664000 68.792000 30.304200 22.164179 100.335821 35.142537 10.711111	BMI DiabetesPedigre 0.017683 0.221071 0.281805 0.392573 0.197859 1.000000 0.140647 0.036242 0.292695 Ze =20) C of squared corr	8Function Age Outcom -0.033523 0.544341 0.22189 0.137337 0.263514 0.46658 0.041265 0.239528 0.06506 0.183928 -0.113970 0.07475 0.185071 -0.042163 0.13054 0.140647 0.036242 0.29269 1.0003561 1.000000 0.23835 0.173844 0.238356 1.00000	98 31 58 52 48 95 44
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Pregnancies 1.000000 0.129459 Glucose 0.129459 1.000000 Bloodfressure 0.141292 0.152590 SkinThickness -0.081672 0.057378 Insulin -0.073535 0.331357 BMI 0.017683 0.221071 DiabetesPedigreeFunction -0.033553 0.337337 Age 0.544341 0.263514 Outcome 0.221898 0.466881 fig = plt.figure(figsize = (19,15)) plt.macstow(df.corr()**2, fignum = fig.number plt.stakes(range(df.shape[1]), df.columns, for plt.plt.colorbar().shape[1]), df.columns, for plt.plt.colorbar().shape[1]), df.columns, for plt.state(params(labets)ze = 14) cb.ax.tick.params(labets)ze = 14) print(df("Correlation matrix of squared correction of the plt.colorbar().shape[1]), df.columns, for plt.state(state) = (19,15) plt.state(stat	0.141282 -0.081672 -0.073535 0.152590	0.017683 0.221071 0.281805 0.392573 0.197859 1.000000 0.140647 0.036242 0.292695	-0.033523	98 81 88 82 95 96 90
plt.matshow(df.corr()**2,fignum = fig.number plt.xticks(range(df.shape[1]),df.columns,for ob = plt.colorbar() cb = plt.colorbar() plt.title("correlation matrix of squared cor Text(0.5, 1.0, 'correlation matrix of squared cor Text(0.5, 1.0, 'correlation matrix of squared cor Text(0.5, 1.0, 'correlation matrix of squared cor BloodPressure - BloodPressure - SkinThickness - SkinThickness - SkinThickness - Mage - Outcome - DiabetesPedigreeFunction - Mage - Outcome - DiabetesPedigreeFunction - BMI - DiabetesPedigreeFunction - Age - Outcome - C:Users\square (figsize = (10, 6)) sns.distplot(df["Glucose"][df["Outcome"] == plt.xticks([i for i in range(0, 201,15)], rote plt.xticks([i for i in range(0, 201,15)], rote plt.xticks([i folucose"] fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose", fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose", fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose", fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose", fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose", fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose", fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose", fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose", fontsize = 20) C:Users\square (figsize = (20, 6)) sns.distplot(df["Glucose")] column (figsize = (20, 6)) dia	ntsize = 14, rotation = 45) rrelations \n\n\n\n\n\n', fontsiz d correlations \n\n\n\n\n\n') correlation matrix	of squared corr		Outcome
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<pre>sns.distplot(df["Insulin"][df["Outcome"]==1] plt.xticks() plt.title("Insulin", fontsize = 20) C:\Users\somu reddy\anaconda3\lib\site-packaged in a future version. Please adapt your convel function for histograms). warnings.warn(msg, FutureWarning) Text(0.5, 1.0, 'Insulin')</pre>	ges\seaborn\distributions.py:2			
0.008 - 0.006 - 0.004 - 0.002 - 0.000	400	600	800 10	000
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<pre>fig = plt.figure(figsize = (16,5)) sns.distplot(df["DiabetesPedigreeFunction"][plt.xticks([i*0.15 for i in range(1,12)]) plt.title("diabetespedigreefunction") C:\Users\somu reddy\anaconda3\lib\site-packaged in a future version. Please adapt your code</pre>	ges\seaborn\distributions.py:2	2557: FutureWarning: figure-level functio	`distplot` is a depreca n with similar flexibil	ted function and will ity) or `histplot` (a
vel function for histograms). warnings.warn(msg, FutureWarning) Text(0.5, 1.0, 'diabetespedigreefunction') 2.00 1.75 1.50 1.25 1.00 0.75	diabetespedigreefunction			
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C:\Users\somu reddy\anaconda3\lib\site-packaged in a future version. Please adapt your convel function for histograms). warnings.warn(msg, FutureWarning) Text(0.5, 1.0, 'Age') 0.035- 0.030- 0.025-				
x = df.drop(["Pregnancies", "BloodPressure", 'y = df.iloc[:,-1]	Age "SkinThickness"],axis = 1)			
<pre>X_train, X_test, y_train, y_test = train_test_s print("X_train size:", X_train.shape) print("y_train size: ", y_train.shape, "\n") print("X_test size:", X_test.shape) print("y_test size:", y_test.shape) X_train size: (614, 8) y_train size: (614,) X_test size: (154, 8) y_test size: (154,) sc= StandardScaler()</pre>	split(X,y,test_size = 0.2,rand	dom_state=0)		
<pre>X_train = sc.fit_transform(X_train) X_test = sc.fit_transform(X_test) print(sc) StandardScaler() print(X_train) [[0.90832902 0.91569367 0.44912368 0</pre>	.50667229 -0.07049698			
[0.03644676 -0.84620959 -0.216349720 -1.04898095] [2.0708387 -1.12937261 0.244362640 0.11706589] [0.32707418 0.47521786 0.653884734 2.94889395]] print(X_test) [[-0.89295432 2.39507259 0.39763774 1 -0.93064283] [-0.56553774 -0.42589245 0.2898275 0 -0.83598035] [0.08929543 -1.37643502 -0.35703388 0 -0.64665539]	.26640405 -0.50001442 .07275877 0.52121586 .52657475 2.78935129 .31944116 -0.27698825 .37136088 -0.31725331			
<pre>[0.08929543 0.64730077 0.93668889 1 -0.93064283] [-0.23812115 -0.14992848 0.28982750 -0.74131787] [-0.89295432 -0.42589245 0.182017270 -0.74131787]] kn_classifier = KNeighborsClassifier(n_neighted) kn_classifier.fit(X_train,y_train) KNeighborsClassifier()</pre> kn_y_pred = kn_classifier.predict(X_test)	.62809381 -1.19689011 .04399691 1.01459113	p = 2)		
<pre>cm_kn = confusion_matrix(y_test, kn_y_pred) print(cm_kn) [[96 11] [17 30]] print("Correct:", sum(kn_y_pred == y_test)) print("Incorrect : ", sum(kn_y_pred != y_test) print("Accuracy:", sum(kn_y_pred == y_test)/le Correct: 126 Incorrect : 28 Accuracy: 0.81818181818182</pre>	en(kn_y_pred))			
<pre>svc_classifier = SVC(kernel ="linear", random svc_classifier.fit(X_train, y_train) SVC(kernel='linear', random_state=0) svc_y_pred = svc_classifier.predict(X_test) svc_cm = confusion_matrix(y_test, svc_y_pred) print(svc_cm) [[96 11] [18 29]]</pre>				
<pre>print("Correct:", sum(svc_y_pred == y_test)) print("Incorrect : ", sum(svc_y_pred != y_test)) print("Accuracy:", sum(svc_y_pred ==y_test)/] Correct: 125 Incorrect : 29 Accuracy: 0.8116883116883117 nb_classifier = GaussianNB() nb_classifier.fit(X_train, y_train) GaussianNB()</pre>				
<pre>nb_y_pred =nb_classifier.predict(X_test) nb_cm = confusion_matrix(nb_y_pred,y_test) print(nb_cm) [[87 16] [20 31]]</pre>	t)) en(nb_y_pred))			