



```
In [2]: import pandas as pd
import numpy as np
from sklearn.preprocessing import MinMaxScaler, StandardScaler
import seaborn as sns
import matplotlib.pyplot as plt

df = pd.read_csv('diabetes.csv')
df.head()
```

```
Out[2]:
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesF
0	6	148	72	35	0	33.6	
1	1	85	66	29	0	26.6	
2	8	183	64	0	0	23.3	
3	1	89	66	23	94	28.1	
4	0	137	40	35	168	43.1	

```
In [3]: df.info()
print(df.isnull().sum())

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
 --- 
 0   Pregnancies      768 non-null    int64  
 1   Glucose          768 non-null    int64  
 2   BloodPressure    768 non-null    int64  
 3   SkinThickness    768 non-null    int64  
 4   Insulin          768 non-null    int64  
 5   BMI              768 non-null    float64 
 6   DiabetesPedigreeFunction 768 non-null    float64 
 7   Age              768 non-null    int64  
 8   Outcome          768 non-null    int64  
dtypes: float64(2), int64(7)
memory usage: 54.1 KB
Pregnancies          0
Glucose              0
BloodPressure        0
SkinThickness        0
Insulin              0
BMI                  0
DiabetesPedigreeFunction 0
Age                  0
Outcome              0
dtype: int64
```

```
In [4]: df.describe()
```

Out[4]:

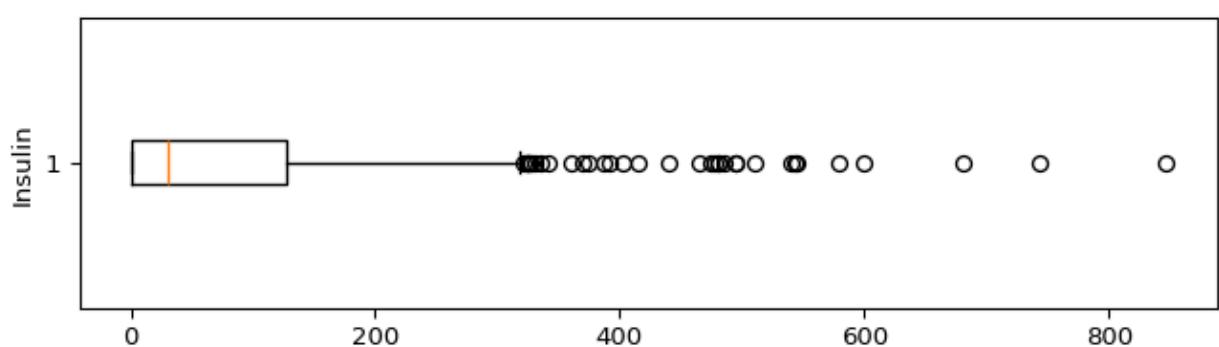
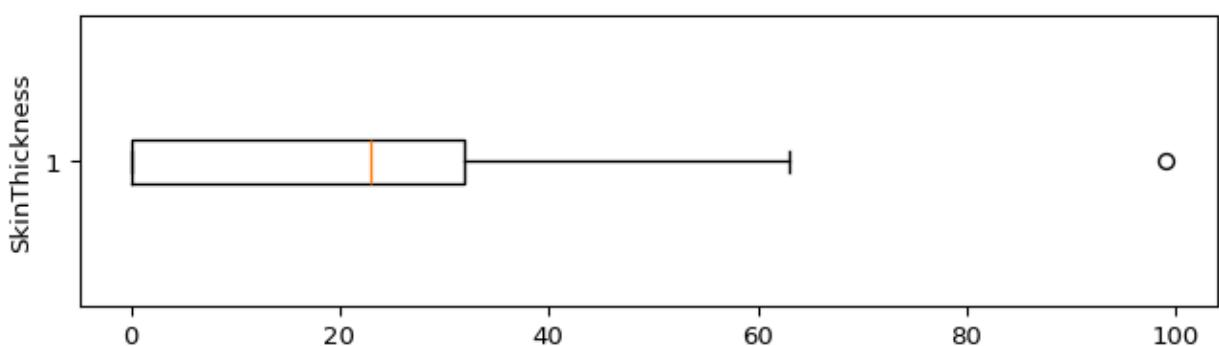
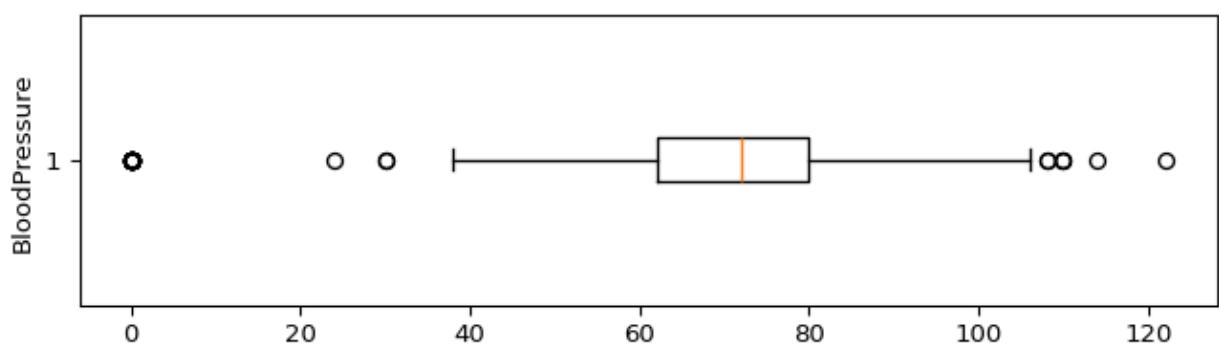
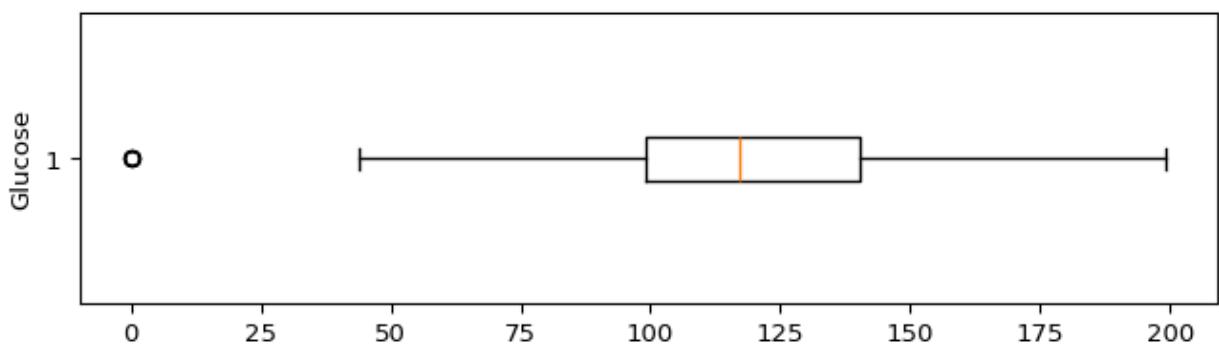
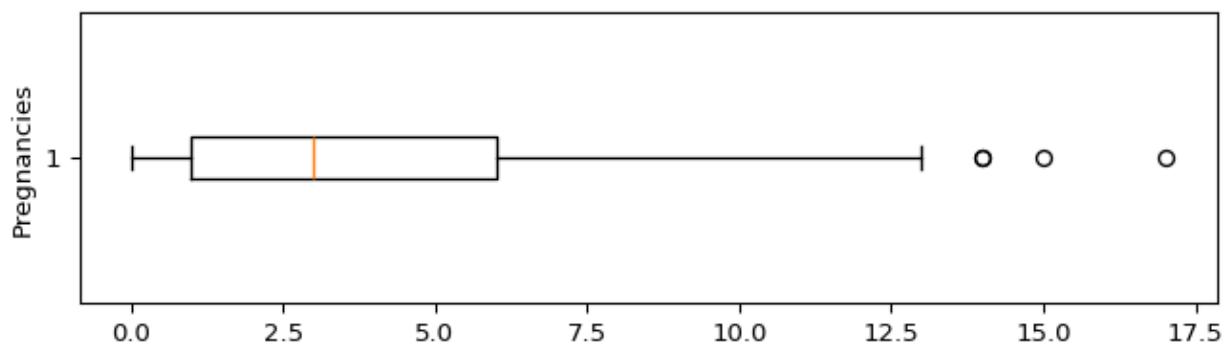
	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	
<b>count</b>	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000
<b>mean</b>	3.845052	120.894531	69.105469	20.536458	79.799479	31.0
<b>std</b>	3.369578	31.972618	19.355807	15.952218	115.244002	7.8
<b>min</b>	0.000000	0.000000	0.000000	0.000000	0.000000	0.0
<b>25%</b>	1.000000	99.000000	62.000000	0.000000	0.000000	27.0
<b>50%</b>	3.000000	117.000000	72.000000	23.000000	30.500000	32.0
<b>75%</b>	6.000000	140.250000	80.000000	32.000000	127.250000	36.0
<b>max</b>	17.000000	199.000000	122.000000	99.000000	846.000000	67.0

In [5]:

```
fig, axs = plt.subplots(len(df.columns), 1, figsize=(7, 18), dpi=95)

for i, col in enumerate(df.columns):
    axs[i].boxplot(df[col], vert=False)
    axs[i].set_ylabel(col)

plt.tight_layout()
plt.show()
```



```
In [6]: q1, q3 = np.percentile(df['Insulin'], [25, 75])
iqr = q3 - q1

lower = q1 - 1.5 * iqr
upper = q3 + 1.5 * iqr

clean_df = df[(df['Insulin'] >= lower) & (df['Insulin'] <= upper)]

print("Before:", df.shape)
print("After:", clean_df.shape)
```

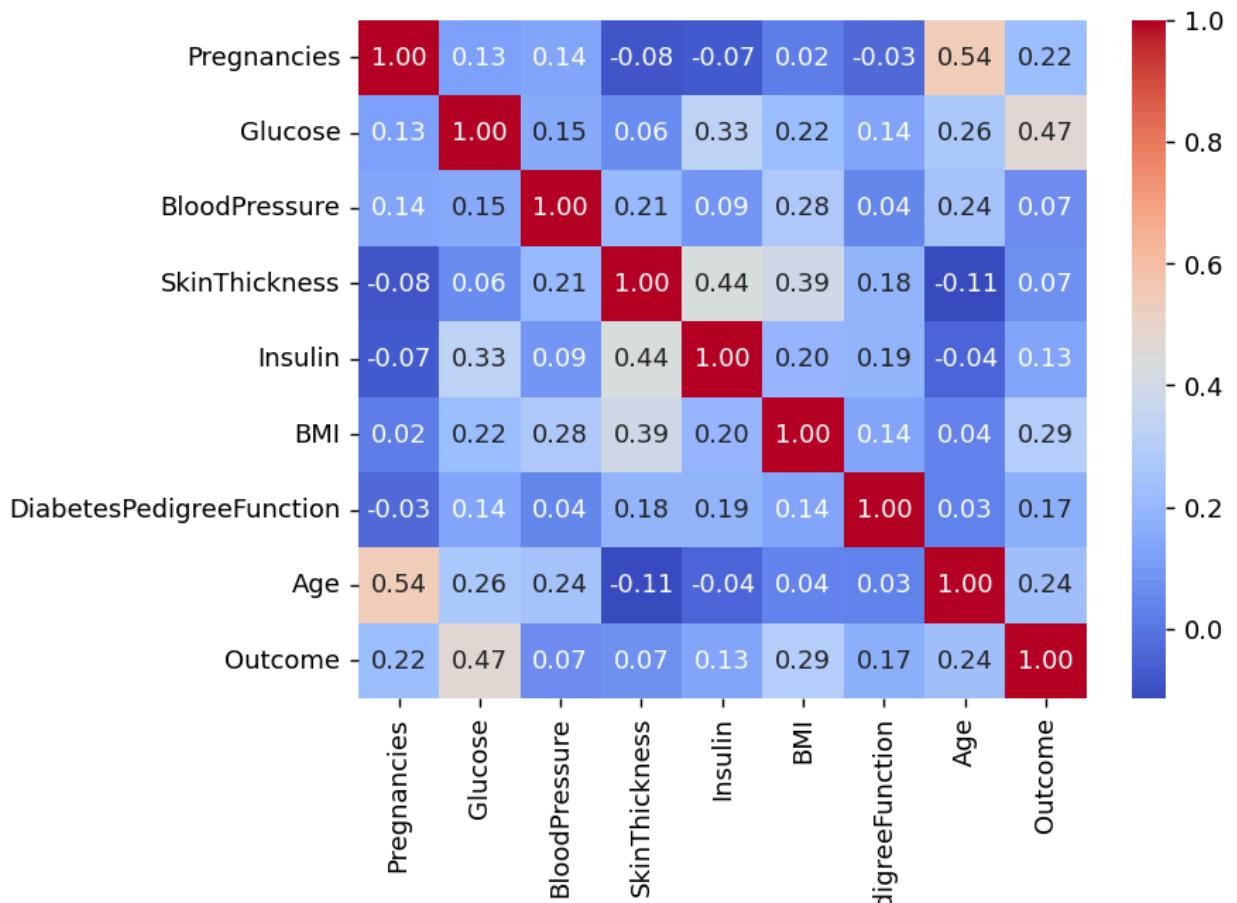
Before: (768, 9)

After: (734, 9)

```
In [7]: corr = df.corr()

plt.figure(dpi=130)
sns.heatmap(corr, annot=True, fmt='.2f', cmap='coolwarm')
plt.show()

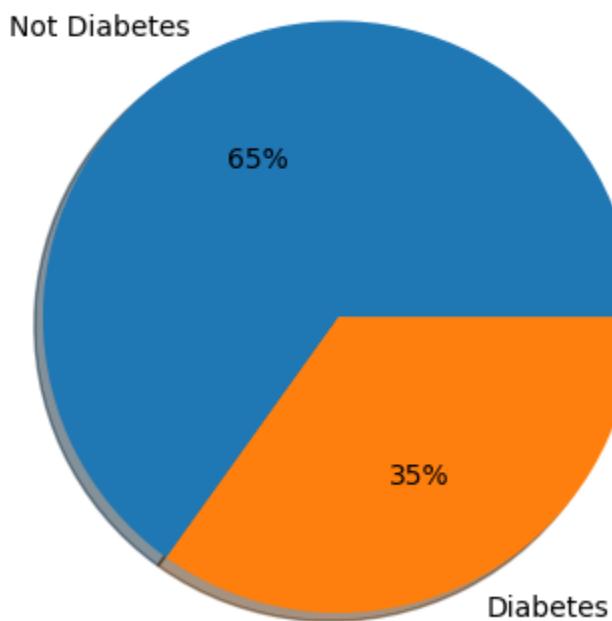
print(corr['Outcome'].sort_values(ascending=False))
```



```
Outcome           1.000000
Glucose          0.466581
BMI              0.292695
Age              0.238356
Pregnancies      0.221898
DiabetesPedigreeFunction 0.173844
Insulin          0.130548
SkinThickness    0.074752
BloodPressure    0.065068
Name: Outcome, dtype: float64
```

```
In [8]: plt.pie(
    df['Outcome'].value_counts(),
    labels=['Not Diabetes', 'Diabetes'],
    autopct='%.f%%',
    shadow=True
)
plt.title('Outcome Proportionality')
plt.show()
```

Outcome Proportionality



```
In [9]: X = df.drop(columns=['Outcome'])
y = df['Outcome']

print(X.shape, y.shape)
```

(768, 8) (768,)

```
In [10]: scaler = MinMaxScaler()
X_normalized = scaler.fit_transform(X)
```

```
print(X_normalized[:5])
```

[0.35294118 0.74371859 0.59016393 0.35353535 0.	0.50074516
0.23441503 0.48333333]	
[0.05882353 0.42713568 0.54098361 0.29292929 0.	0.39642325
0.11656704 0.16666667]	
[0.47058824 0.91959799 0.52459016 0. 0.	0.34724292
0.25362938 0.18333333]	
[0.05882353 0.44723618 0.54098361 0.23232323 0.11111111	0.41877794
0.03800171 0. ]	
[0. 0.68844221 0.32786885 0.35353535 0.19858156	0.64232489
0.94363792 0.2 ]]	

```
In [11]: scaler = StandardScaler()  
X_standardized = scaler.fit_transform(X)  
  
print(X_standardized[:5])
```

[[ 0.63994726 0.84832379 0.14964075 0.90726993 -0.69289057 0.20401277 0.46849198 1.4259954 ]]	
[-0.84488505 -1.12339636 -0.16054575 0.53090156 -0.69289057 -0.68442195 -0.36506078 -0.19067191]	
[ 1.23388019 1.94372388 -0.26394125 -1.28821221 -0.69289057 -1.10325546 0.60439732 -0.10558415]	
[-0.84488505 -0.99820778 -0.16054575 0.15453319 0.12330164 -0.49404308 -0.92076261 -1.04154944]	
[-1.14185152 0.5040552 -1.50468724 0.90726993 0.76583594 1.4097456 5.4849091 -0.0204964 ]]	