

```
# Import necessary libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.datasets import load_breast_cancer
from sklearn.preprocessing import StandardScaler
from sklearn.feature_selection import SelectKBest, f_classif
```

```
# Load the breast cancer dataset
```

```
data = load_breast_cancer()
df = pd.DataFrame(data.data, columns=data.feature_names)
df['target'] = data.target
```

```
df
```

	mean radius	mean texture	mean perimeter	mean area	mean
smoothness \					
0	17.99	10.38	122.80	1001.0	
0.11840					
1	20.57	17.77	132.90	1326.0	
0.08474					
2	19.69	21.25	130.00	1203.0	
0.10960					
3	11.42	20.38	77.58	386.1	
0.14250					
4	20.29	14.34	135.10	1297.0	
0.10030					
..	
...					
564	21.56	22.39	142.00	1479.0	
0.11100					
565	20.13	28.25	131.20	1261.0	
0.09780					
566	16.60	28.08	108.30	858.1	
0.08455					
567	20.60	29.33	140.10	1265.0	
0.11780					
568	7.76	24.54	47.92	181.0	
0.05263					

	mean compactness	mean concavity	mean concave points	mean
symmetry \				
0	0.27760	0.30010	0.14710	
0.2419				
1	0.07864	0.08690	0.07017	
0.1812				
2	0.15990	0.19740	0.12790	
0.2069				
3	0.28390	0.24140	0.10520	

0.2597			
4	0.13280	0.19800	0.10430
0.1809			
..
...			
564	0.11590	0.24390	0.13890
0.1726			
565	0.10340	0.14400	0.09791
0.1752			
566	0.10230	0.09251	0.05302
0.1590			
567	0.27700	0.35140	0.15200
0.2397			
568	0.04362	0.00000	0.00000
0.1587			

	mean fractal dimension	...	worst radius	worst texture \
0	0.07871	...	25.380	17.33
1	0.05667	...	24.990	23.41
2	0.05999	...	23.570	25.53
3	0.09744	...	14.910	26.50
4	0.05883	...	22.540	16.67
..
564	0.05623	...	25.450	26.40
565	0.05533	...	23.690	38.25
566	0.05648	...	18.980	34.12
567	0.07016	...	25.740	39.42
568	0.05884	...	9.456	30.37

	worst perimeter	worst area	worst smoothness	worst compactness
\				
0	184.60	2019.0	0.16220	0.66560
1	158.80	1956.0	0.12380	0.18660
2	152.50	1709.0	0.14440	0.42450
3	98.87	567.7	0.20980	0.86630
4	152.20	1575.0	0.13740	0.20500
..
564	166.10	2027.0	0.14100	0.21130
565	155.00	1731.0	0.11660	0.19220
566	126.70	1124.0	0.11390	0.30940
567	184.60	1821.0	0.16500	0.86810

568	59.16	268.6	0.08996	0.06444
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	worst concavity	worst concave points	worst symmetry \
0	0.7119	0.2654	0.4601
1	0.2416	0.1860	0.2750
2	0.4504	0.2430	0.3613
3	0.6869	0.2575	0.6638
4	0.4000	0.1625	0.2364
..
564	0.4107	0.2216	0.2060
565	0.3215	0.1628	0.2572
566	0.3403	0.1418	0.2218
567	0.9387	0.2650	0.4087
568	0.0000	0.0000	0.2871

	worst fractal dimension
0	0.11890
1	0.08902
2	0.08758
3	0.17300
4	0.07678
..	...
564	0.07115
565	0.06637
566	0.07820
567	0.12400
568	0.07039

[569 rows x 30 columns]

Display the first few rows of the dataset

df.head()

	mean radius	mean texture	mean perimeter	mean area	mean smoothness \
0	17.99	10.38	122.80	1001.0	0.11840
1	20.57	17.77	132.90	1326.0	0.08474
2	19.69	21.25	130.00	1203.0	0.10960
3	11.42	20.38	77.58	386.1	0.14250
4	20.29	14.34	135.10	1297.0	0.10030
..
564	17.99	10.38	122.80	1001.0	0.11840
565	20.57	17.77	132.90	1326.0	0.08474
566	19.69	21.25	130.00	1203.0	0.10960
567	11.42	20.38	77.58	386.1	0.14250
568	20.29	14.34	135.10	1297.0	0.10030
..
569

	mean compactness	mean concavity	mean concave points	mean
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```

symmetry \
0          0.27760          0.3001          0.14710
0.2419
1          0.07864          0.0869          0.07017
0.1812
2          0.15990          0.1974          0.12790
0.2069
3          0.28390          0.2414          0.10520
0.2597
4          0.13280          0.1980          0.10430
0.1809

    mean fractal dimension ... worst texture worst perimeter worst
area \
0          0.07871 ...          17.33          184.60
2019.0
1          0.05667 ...          23.41          158.80
1956.0
2          0.05999 ...          25.53          152.50
1709.0
3          0.09744 ...          26.50          98.87
567.7
4          0.05883 ...          16.67          152.20
1575.0

    worst smoothness worst compactness worst concavity worst concave
points \
0          0.1622          0.6656          0.7119
0.2654
1          0.1238          0.1866          0.2416
0.1860
2          0.1444          0.4245          0.4504
0.2430
3          0.2098          0.8663          0.6869
0.2575
4          0.1374          0.2050          0.4000
0.1625

    worst symmetry worst fractal dimension target
0          0.4601          0.11890          0
1          0.2750          0.08902          0
2          0.3613          0.08758          0
3          0.6638          0.17300          0
4          0.2364          0.07678          0

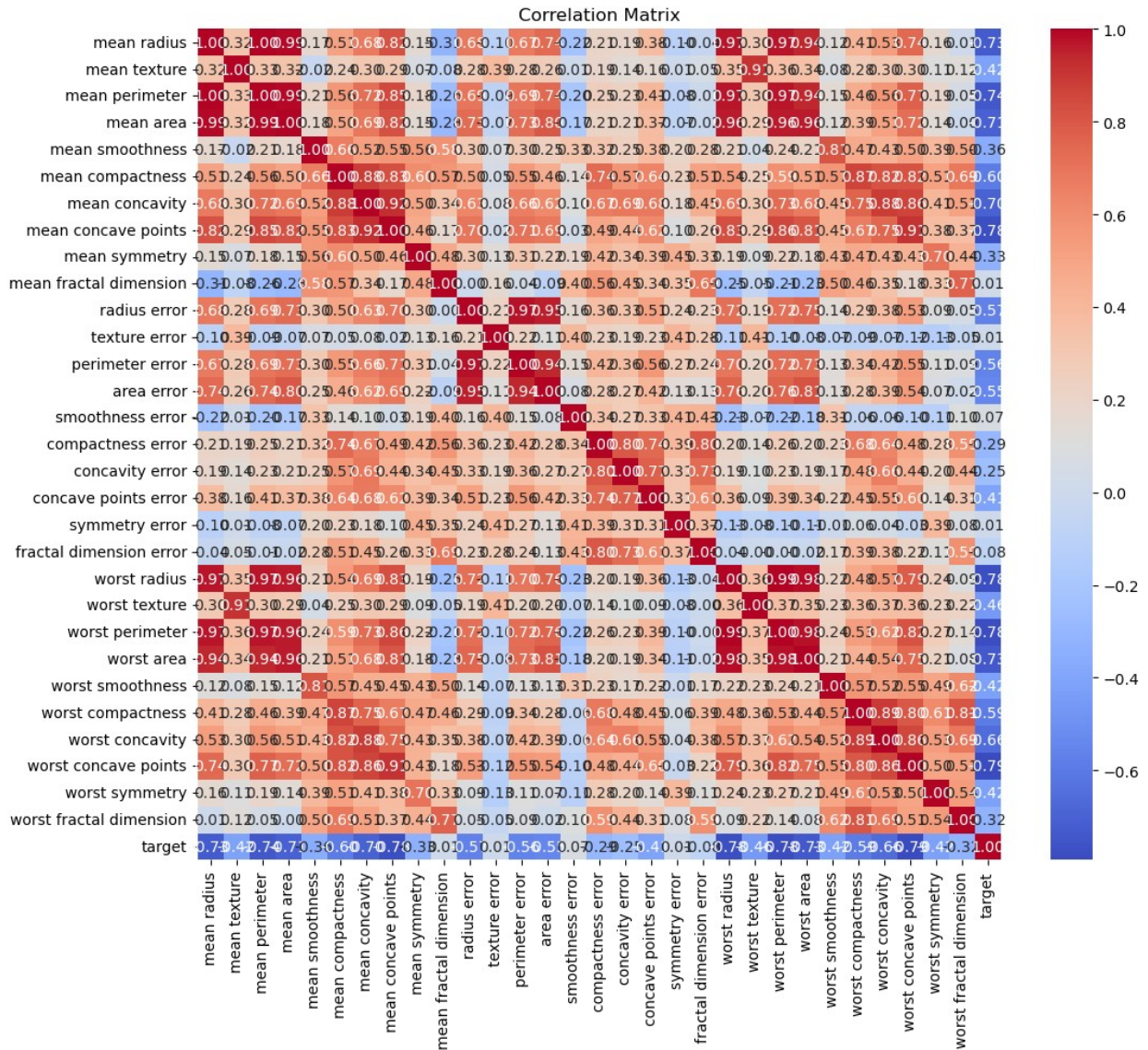
```

```
[5 rows x 31 columns]
```

```
# Task 3: Remove constant columns
```

```
df = df.loc[:, (df != df.iloc[0]).any()]
```

```
# Task 4: Encode categorical features (if any)  
# In this dataset, encoding is not necessary as all features are  
numerical  
  
# Task 5: Feature scaling  
numerical_features = df.drop('target', axis=1).columns  
scaler = StandardScaler()  
df[numerical_features] = scaler.fit_transform(df[numerical_features])  
  
# Task 6: Correlation analysis  
correlation_matrix = df.corr()  
plt.figure(figsize=(12, 10))  
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm',  
fmt=".2f")  
plt.title("Correlation Matrix")  
plt.show()
```



```
# Task 7: Feature selection
```

```
X = df.drop('target', axis=1)
```

```
y = df['target']
```

```
k_best_selector = SelectKBest(score_func=f_classif, k=5)
```

```
X_new = k_best_selector.fit_transform(X, y)
```

```
# Display the selected features
```

```
selected_features = X.columns[k_best_selector.get_support()]
```

```
print("\nSelected Features:")
```

```
print(selected_features)
```

Selected Features:

Index(['mean perimeter', 'mean concave points', 'worst radius',

```
'worst perimeter', 'worst concave points'],  
dtype='object')
```