import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear\_model import LinearRegressio
from sklearn.model\_selection import train\_test\_s
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import mean\_squared\_error,

from sklearn.linear\_model import LogisticRegress
# Import the datasets module from sklearn
from sklearn import datasets
# Access load\_breast\_cancer directly from the im
data = datasets.load\_breast\_cancer()

data

Show hidden output

data.feature\_names

array(['mean radius', 'mean texture', 'mean perimeter', 'mean area', 'mean smoothness', 'mean symmetry', 'mean fractal dimension', 'radius error', 'texture error', 'perimeter error', 'area error', 'smoothness error', 'compactness error', 'concavity error', 'concave points error', 'symmetry error', 'fractal dimension error', 'worst radius', 'worst texture', 'worst perimeter', 'worst area', 'worst smoothness', 'worst compactness', 'worst concavity', 'worst concave points', 'worst symmetry', 'worst fractal dimension'], dtype='<U23')</pre>

data.target

ValueError

NameError •••



Please explain the error:

NameError: name 'sklearn



## Hypothesis

The traceback indicates a
NameError: name
'sklearn' is not defined.
This error occurs when you try to
use a variable or module name
that hasn't been properly imported
or defined in the current scope.

In the provided code, you are trying to access sklearn.datasets.load\_bre While you have imported datasets from sklearn using from sklearn import datasets, this import statement only makes the datasets module available directly. It does not make the top-level sklearn package itself available under the name sklearn. Therefore, when you try to use

sklearn.datasets, the name sklearn is not recognized.

# Suggested Changes

To fix this, you need to import the entire sklearn package directly or use the imported datasets module correctly. The most

array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1])

common way to use modules from sklearn is to import the specific module or function you need directly, as you have done with datasets.

You can either change the line

data=
sklearn.datasets.load\_breato use the imported datasets
directly, or import sklearn as a
whole. The former is generally
preferred as it keeps the
namespace cleaner.

# Import the datasets mode
from sklearn import datas
# Access load\_breast\_cand
data = datasets.load\_brea

Use code with caution

∄□□

df=pd.DataFrame(data.data,columns=data.feature\_n

#### df.head(20)



worst		worst	worst	wor compactne
17.33	184.60	2019.0	0.1622	0.66
23.41	158.80	1956.0	0.1238	0.18
25.53	152.50	1709.0	0.1444	0.42
26.50	98.87	567.7	0.2098	0.86
16.67	152.20	1575.0	0.1374	0.20
23.75	103.40	741.6	0.1791	0.52
27.66	153.20	1606.0	0.1442	0.25
28.14	110.60	897.0	0.1654	0.36
30.73	106.20	739.3	0.1703	0.54
40.68	97.65	711.4	0.1853	1.05
33.88	123.80	1150.0	0.1181	0.15
27.28	136.50	1299.0	0.1396	0.56
29.94	151.70	1332.0	0.1037	0.39
27.66	112.00	876.5	0.1131	0.19
32.01	108.80	697.7	0.1651	0.77
37.13	124.10	943.2	0.1678	0.65
30.88	123.40	1138.0	0.1464	0.18
31.48	136.80	1315.0	0.1789	0.42
30.88	186.80	2398.0	0.1512	0.31
19.26	99.70	711.2	0.1440	0.17

df['target']=data.target

df.isnull().sum()

 $\overline{\Sigma}$ 

Show hidden output

df['target'].value\_counts()

 $\overline{\Rightarrow}$ 

count

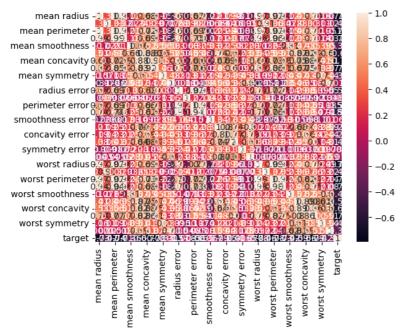
target	
1	357
0	212

dtype: int64

#### sns.heatmap(df.corr(),annot=True)



<Axes: >



df.shape

**→** (569, 31)

x=df.iloc[:,:-1]
y=df.iloc[:,-1]

Χ

Show hidden output

У



	target
0	0
1	0
2	0
3	0
4	0
564	0
565	0
566	0
567	0
568	1

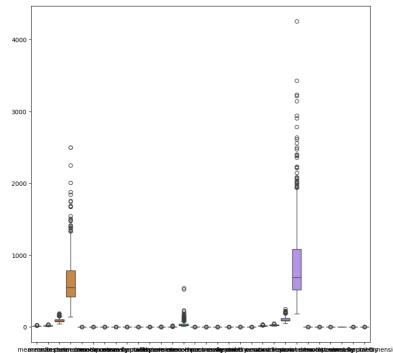
569 rows × 1 columns

dtype: int64

fig, ax = plt.subplots(figsize=(10, 10))
sns.boxplot(data=x, ax=ax)







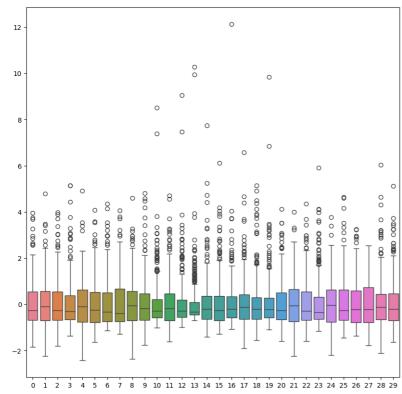
x\_train, x\_test, y\_train, y\_test = train\_test\_sp

scaler = StandardScaler()
x\_train\_norm = scaler.fit\_transform(x\_train)
x\_test\_norm = scaler.transform(x\_test)

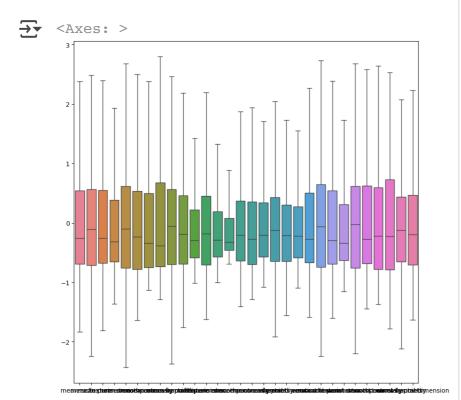
fig, ax = plt.subplots(figsize=(10, 10))
sns.boxplot(data=x\_train\_norm, ax=ax)



<Axes: >



fig, ax = plt.subplots(figsize=(10, 10))
sns.boxplot(data=x\_train\_df, ax=ax)



from sklearn.linear\_model import LogisticRegress
classification = LogisticRegression()
classification.fit(x\_train\_df, y\_train)



▼ LogisticRegression ① ?

LogisticRegression()

from imblearn.over\_sampling import SMOTE
smote = SMOTE(random\_state=42)
X\_train\_resampled, y\_train\_resampled = smote.fit\_
# Check resampled class distribution
print("\nResampled class distribution:")
print(pd.Series(y\_train\_resampled).value\_counts())
# Check resampled class distribution

**₹** 

Resampled class distribution: target
1 249

249
 249

Name: count, dtype: int64

y\_pred = classification.predict(x\_test\_norm)

/usr/local/lib/python3.11/dist-packages/skle
warnings.warn(

from sklearn.metrics import accuracy\_score
accuracy\_score (y\_test,y\_pred)\*100

**→** 98.24561403508771

### Enter a prompt here



0/2000

Gemini can make mistakes so double-check responses and use code with caution. <u>Learn more</u>