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# Data Science Mini Project
# Advanced Explorations Using Python
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# 1. Import Libraries & Set Options
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

# Set pandas display options
pd.set_option("display.max_columns", None)
pd.set_option("display.max_rows", 50)
sns.set_style("whitegrid")

# 2. Read Data
df = pd.read_csv("insurance.csv")
print("Dataset Loaded Successfully!")
print("Shape of data:", df.shape)
```

Dataset Loaded Successfully!
Shape of data: (1338, 7)

```
# 3. Understand & Prepare Data
print("\n--- First 5 Rows ---")
print(df.head())
print("\n--- Data Info ---")
print(df.info())
print("\n--- Summary Statistics ---")
print(df.describe(include="all"))
```

--- First 5 Rows ---

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.92400
1	18	male	33.770	1	no	southeast	1725.55230
2	28	male	33.000	3	no	southeast	4449.46200
3	33	male	22.705	0	no	northwest	21984.47061
4	32	male	28.880	0	no	northwest	3866.85520

--- Data Info ---

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1338 entries, 0 to 1337
Data columns (total 7 columns):
#   Column      Non-Null Count  Dtype
---  -
0   age         1338 non-null   int64
1   sex         1338 non-null   object
2   bmi         1338 non-null   float64
3   children    1338 non-null   int64
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4   smoker    1338 non-null   object
5   region    1338 non-null   object
6   charges   1338 non-null   float64
dtypes: float64(2), int64(2), object(3)
memory usage: 73.3+ KB
None

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--- Summary Statistics ---

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	age	sex	bmi	children	smoker	region
\count	1338.000000	1338	1338.000000	1338.000000	1338	1338
unique	NaN	2	NaN	NaN	2	4
top	NaN	male	NaN	NaN	no	southeast
freq	NaN	676	NaN	NaN	1064	364
mean	39.207025	NaN	30.663397	1.094918	NaN	NaN
std	14.049960	NaN	6.098187	1.205493	NaN	NaN
min	18.000000	NaN	15.960000	0.000000	NaN	NaN
25%	27.000000	NaN	26.296250	0.000000	NaN	NaN
50%	39.000000	NaN	30.400000	1.000000	NaN	NaN
75%	51.000000	NaN	34.693750	2.000000	NaN	NaN
max	64.000000	NaN	53.130000	5.000000	NaN	NaN

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count    charges
unique    NaN
top       NaN
freq      NaN
mean    13270.422265
std     12110.011237
min      1121.873900
25%      4740.287150
50%      9382.033000
75%     16639.912515
max     63770.428010

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#4.()) Understand the Variables (EDA Basics)
print("\n--- Unique values per column ---")
for col in df.columns:
    print(col, ":", df[col].nunique())

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--- Unique values per column ---
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```
age : 47  
sex : 2  
bmi : 548  
children : 6  
smoker : 2  
region : 4  
charges : 1337
```

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# Distribution plots for numerical features
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num_cols = df.select_dtypes(include=np.number).columns
```

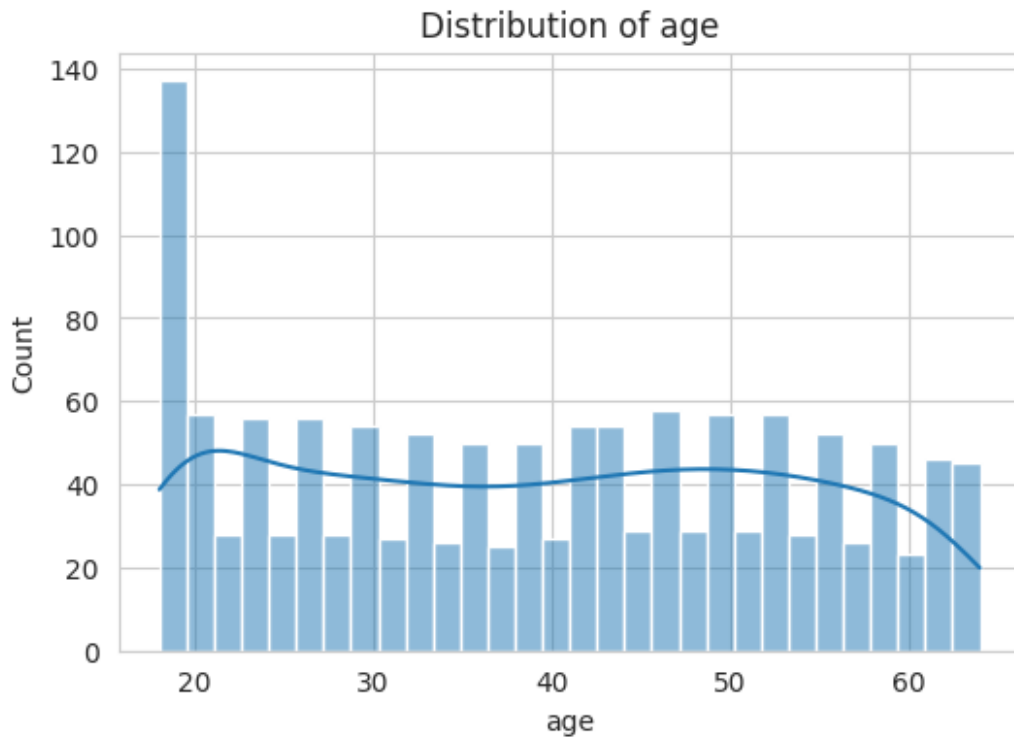
```
for col in num_cols:
```

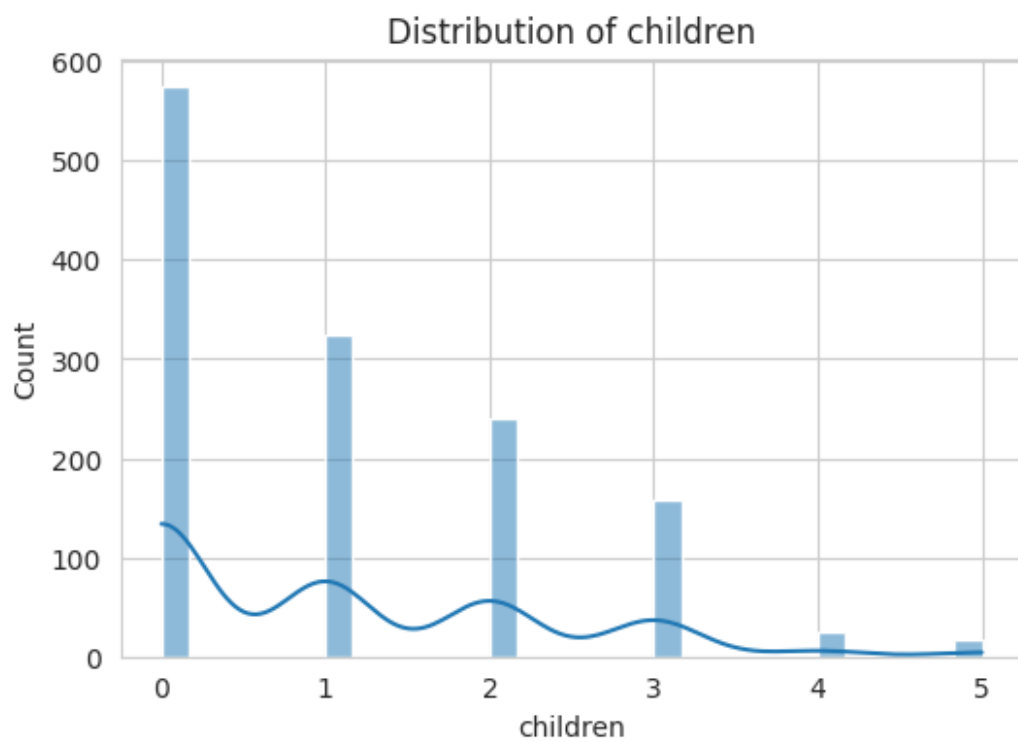
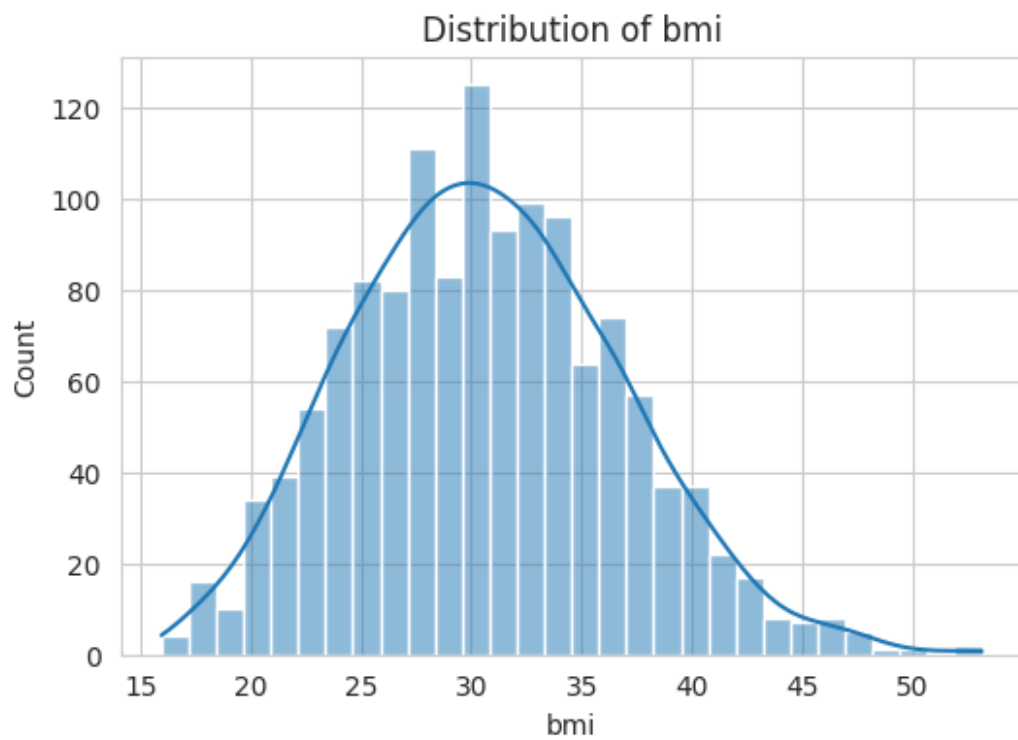
```
    plt.figure(figsize=(6,4))
```

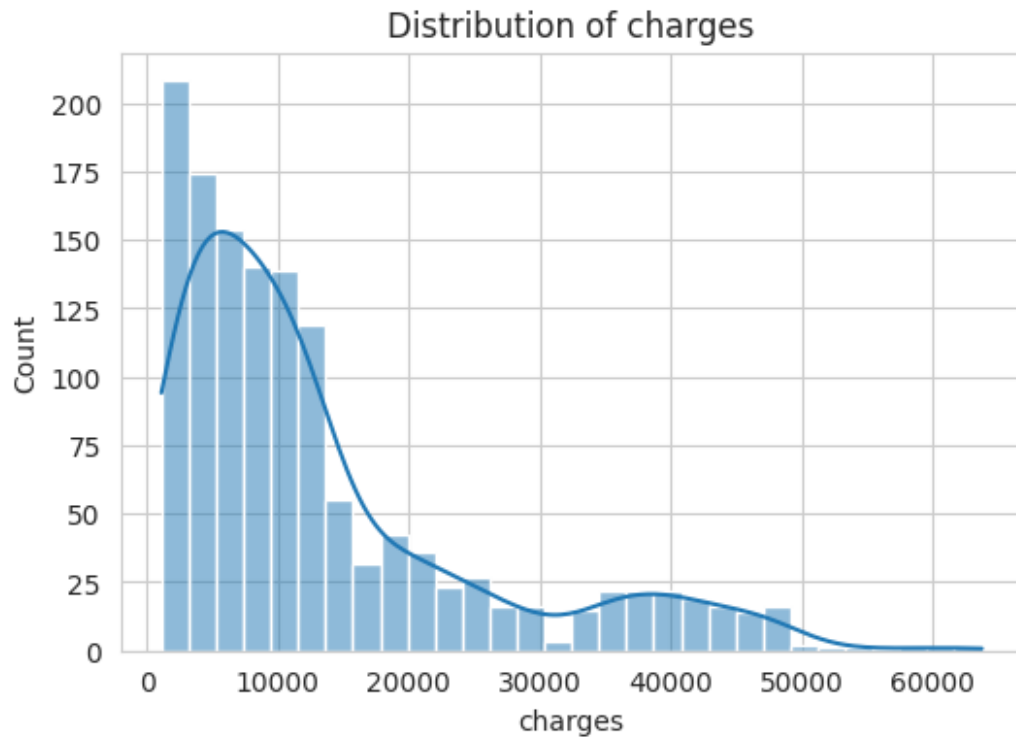
```
    sns.histplot(df[col], kde=True, bins=30)
```

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    plt.title(f"Distribution of {col}")
```

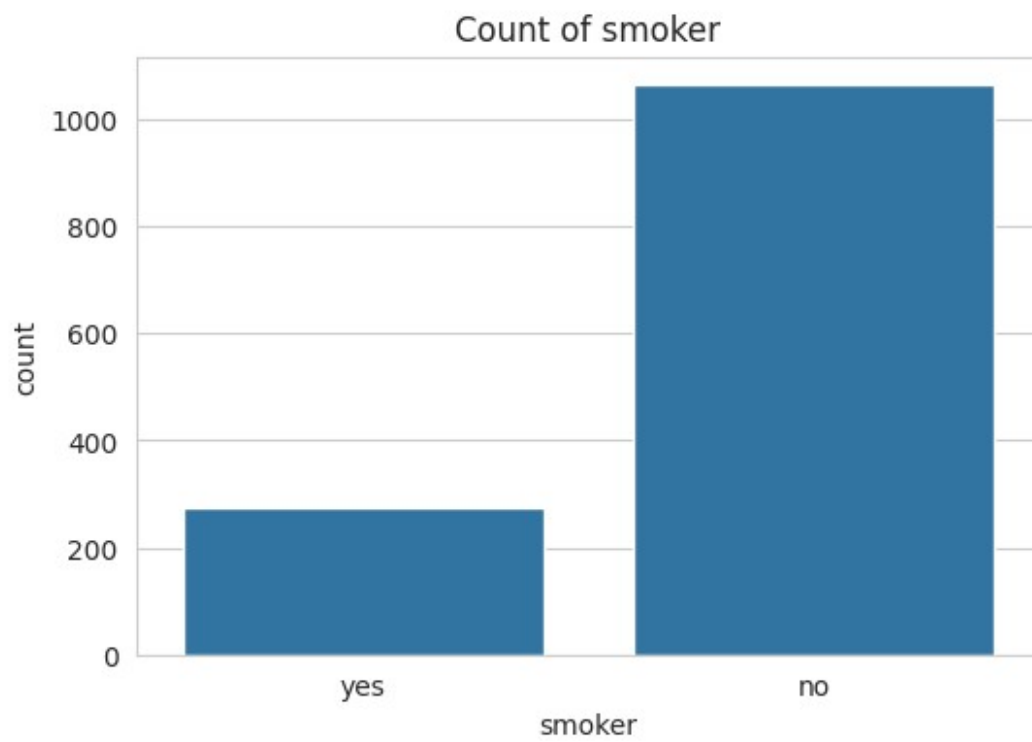
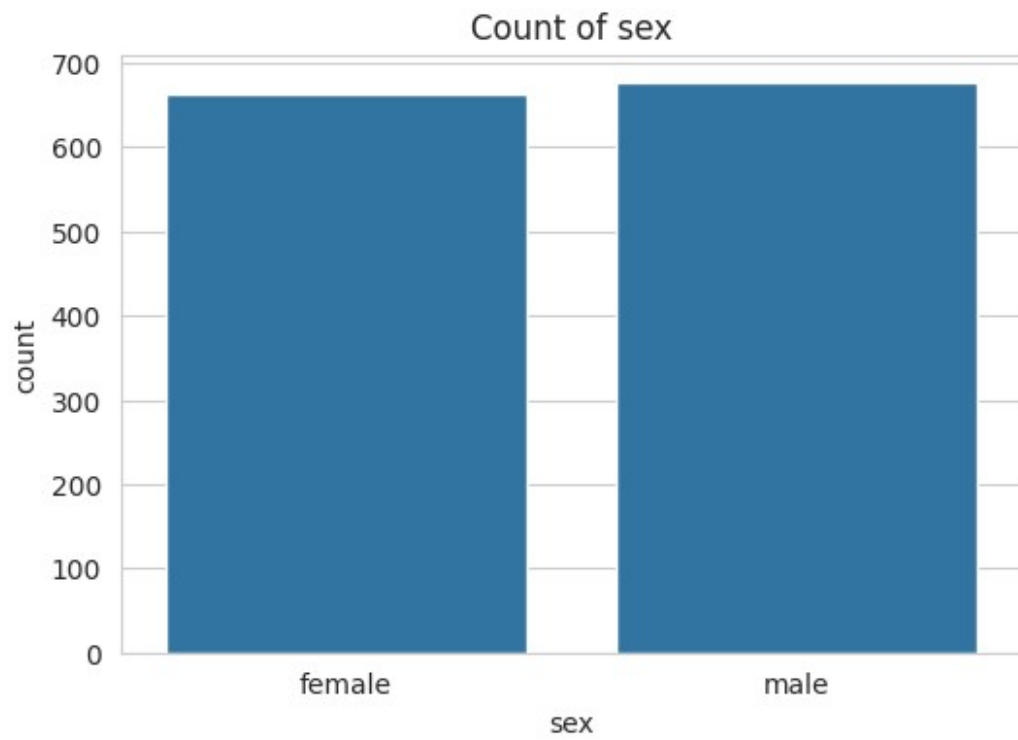
```
    plt.show()
```

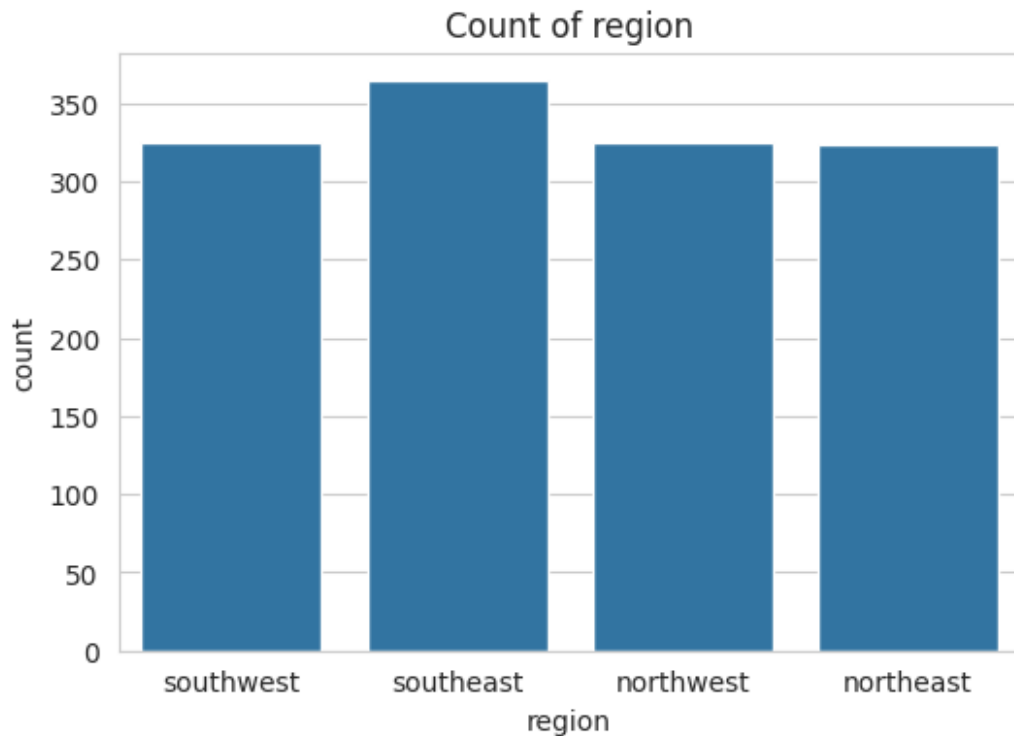






```
# Count plots for categorical features
cat_cols = df.select_dtypes(include="object").columns
for col in cat_cols:
    plt.figure(figsize=(6,4))
    sns.countplot(x=df[col])
    plt.title(f"Count of {col}")
    plt.show()
```

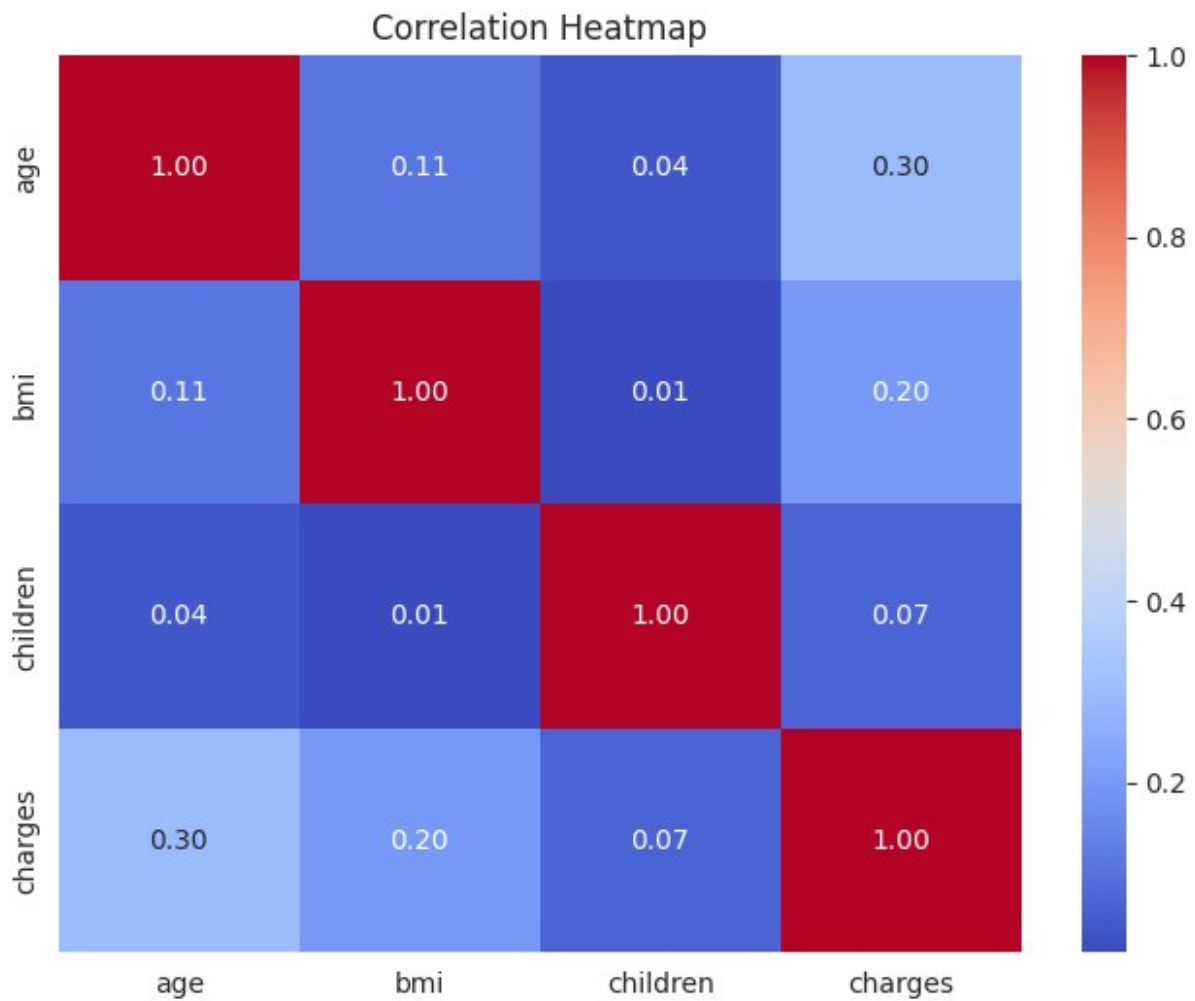




```
# 5. Check for Missing Values
print("\n--- Missing Values ---")
print(df.isnull().sum())

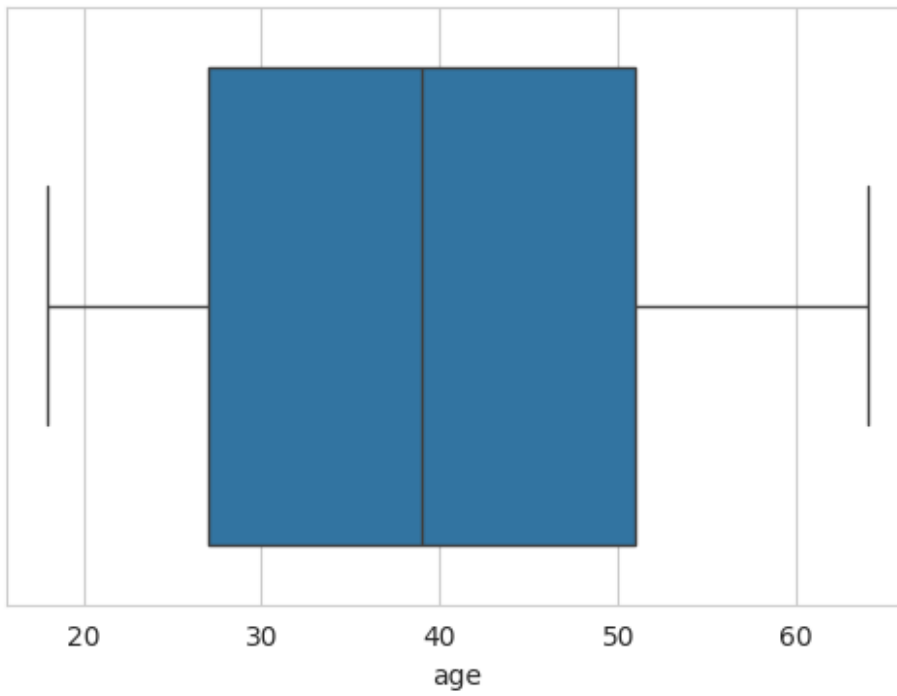
--- Missing Values ---
age      0
sex      0
bmi      0
children 0
smoker   0
region   0
charges  0
dtype: int64

# 6. Study Correlation (numeric only)
plt.figure(figsize=(8,6))
numeric_df = df.select_dtypes(include=[np.number]) # keep only
numeric columns
sns.heatmap(numeric_df.corr(), annot=True, cmap="coolwarm", fmt=".2f")
plt.title("Correlation Heatmap")
plt.show()
```

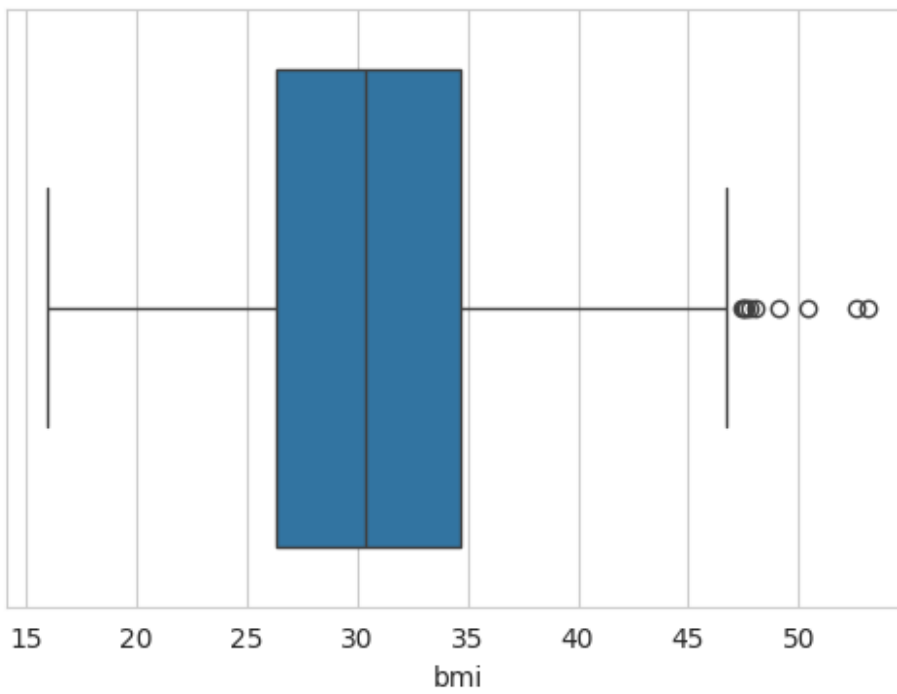


```
# 7. Detect Outliers
# Using Boxplots
for col in num_cols:
    plt.figure(figsize=(6,4))
    sns.boxplot(x=df[col])
    plt.title(f"Outlier Detection for {col}")
    plt.show()
```

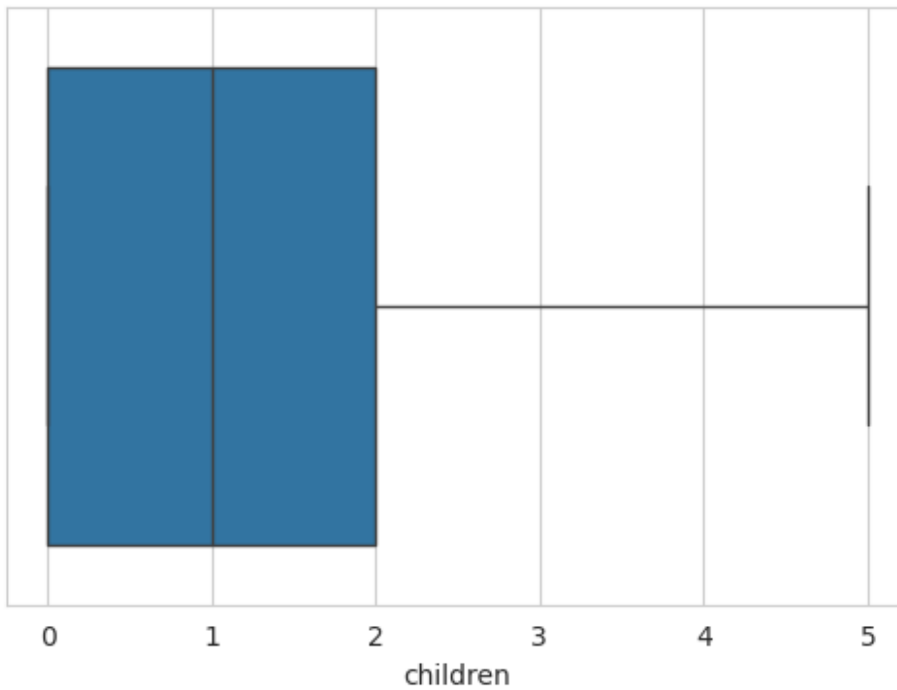

Outlier Detection for age



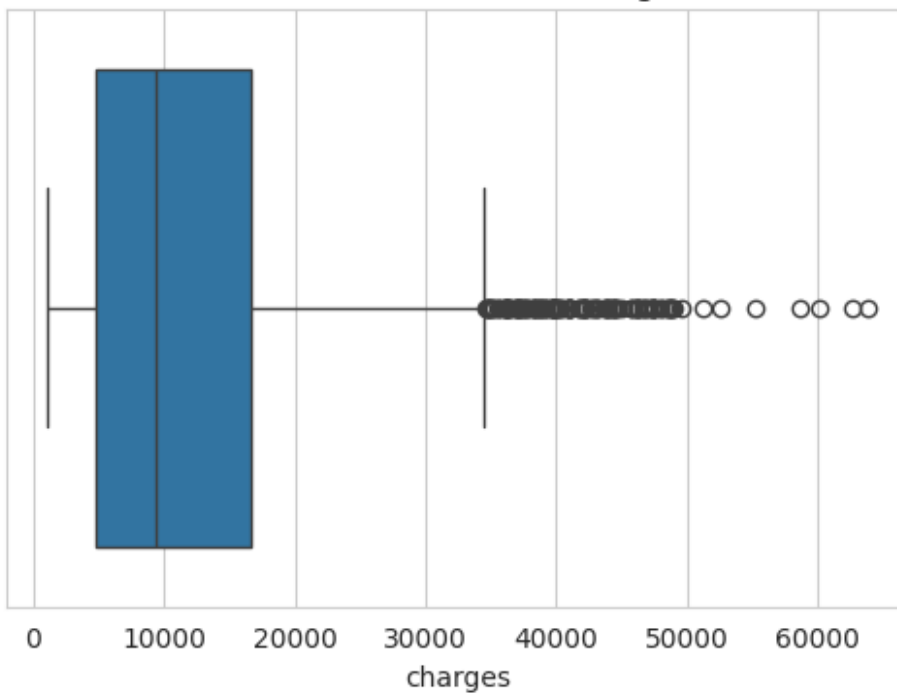
Outlier Detection for bmi



Outlier Detection for children



Outlier Detection for charges



```
# Outlier detection using IQR
def detect_outliers_iqr(data, col):
    Q1 = data[col].quantile(0.25)
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Q3 = data[col].quantile(0.75)
IQR = Q3 - Q1
lower = Q1 - 1.5 * IQR
upper = Q3 + 1.5 * IQR
outliers = data[(data[col] < lower) | (data[col] > upper)]
return outliers

for col in num_cols:
    outliers = detect_outliers_iqr(df, col)
    print(f"{col}: {len(outliers)} outliers")

age: 0 outliers
bmi: 9 outliers
children: 0 outliers
charges: 139 outliers

# 8. Feature Engineering – 'region' Variable
print("\n--- Feature Engineering on 'region' ---")
print(df['region'].value_counts())

--- Feature Engineering on 'region' ---
region
southeast    364
southwest    325
northwest    325
northeast    324
Name: count, dtype: int64

# Convert region into dummy variables
df_encoded = pd.get_dummies(df, columns=['region'], drop_first=True)
print("After Encoding, new columns:", df_encoded.columns)

After Encoding, new columns: Index(['age', 'sex', 'bmi', 'children',
    'smoker', 'charges',
    'region_northwest', 'region_southeast', 'region_southwest'],
    dtype='object')

# Example: Create a new feature "is_southeast"
df["is_southeast"] = np.where(df["region"] == "southeast", 1, 0)
print(df[["region", "is_southeast"]].head())

print("\nFinal Dataset Shape:", df.shape)

   region  is_southeast
0  southwest           0
1  southeast           1
2  southeast           1
3  northwest           0
4  northwest           0

```

Final Dataset Shape: (1338, 8)