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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

--- Summary Statistics ---
          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

          charges
\count  1338.000000
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

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

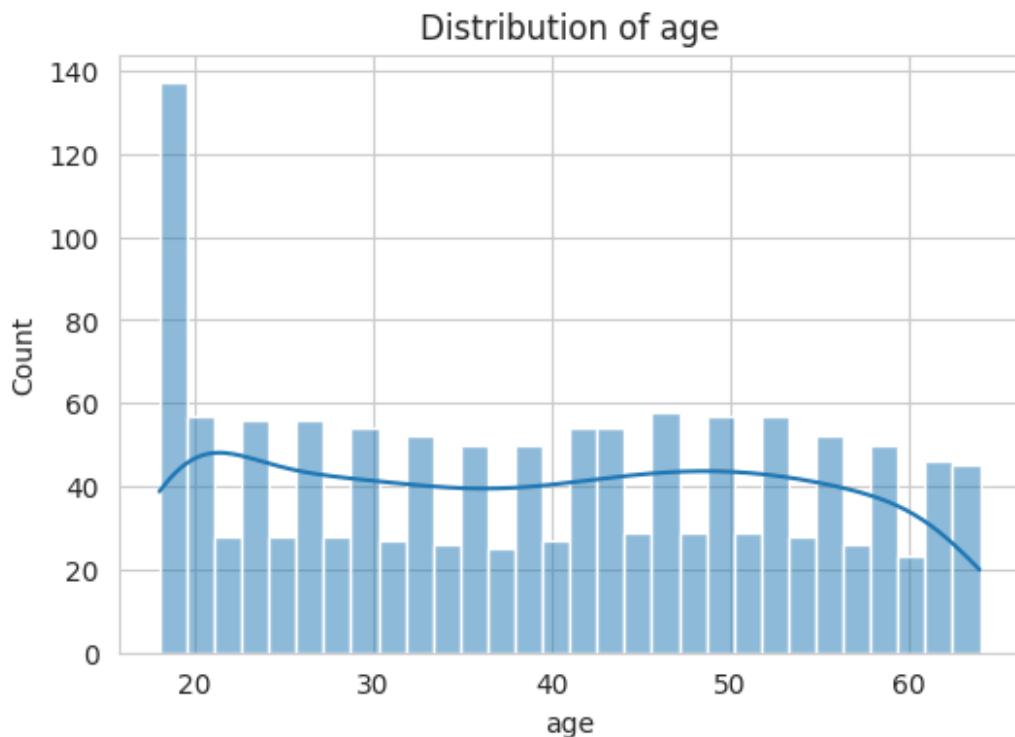
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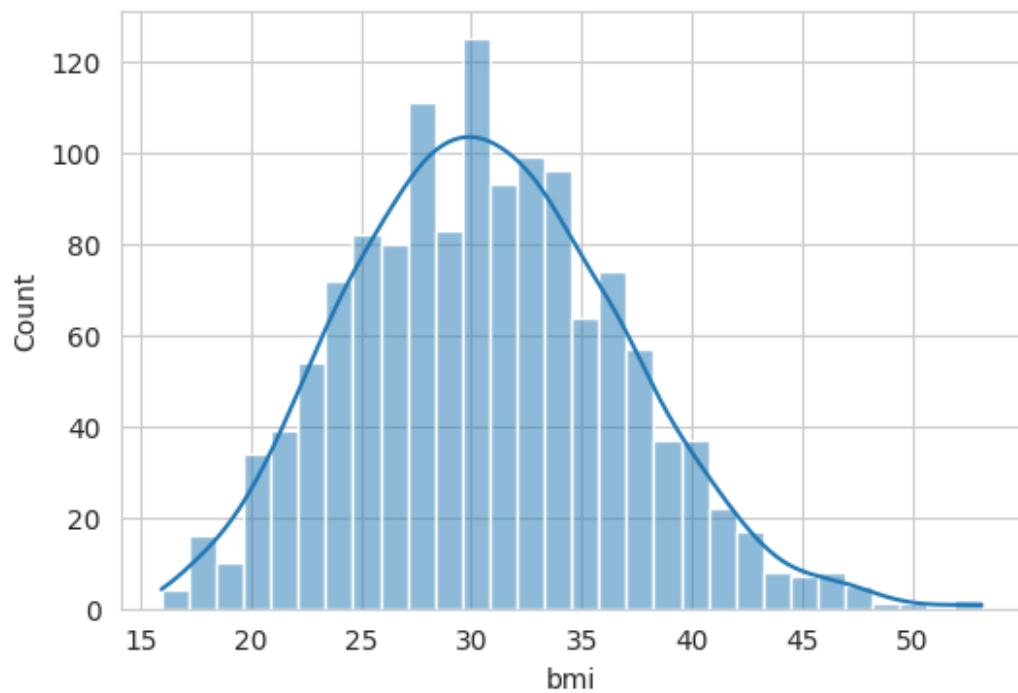
--- Unique values per column ---
age : 47
sex : 2
bmi : 548
children : 6
smoker : 2
region : 4
charges : 1337

# Distribution plots for numerical features
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)
    plt.title(f"Distribution of {col}")
    plt.show()

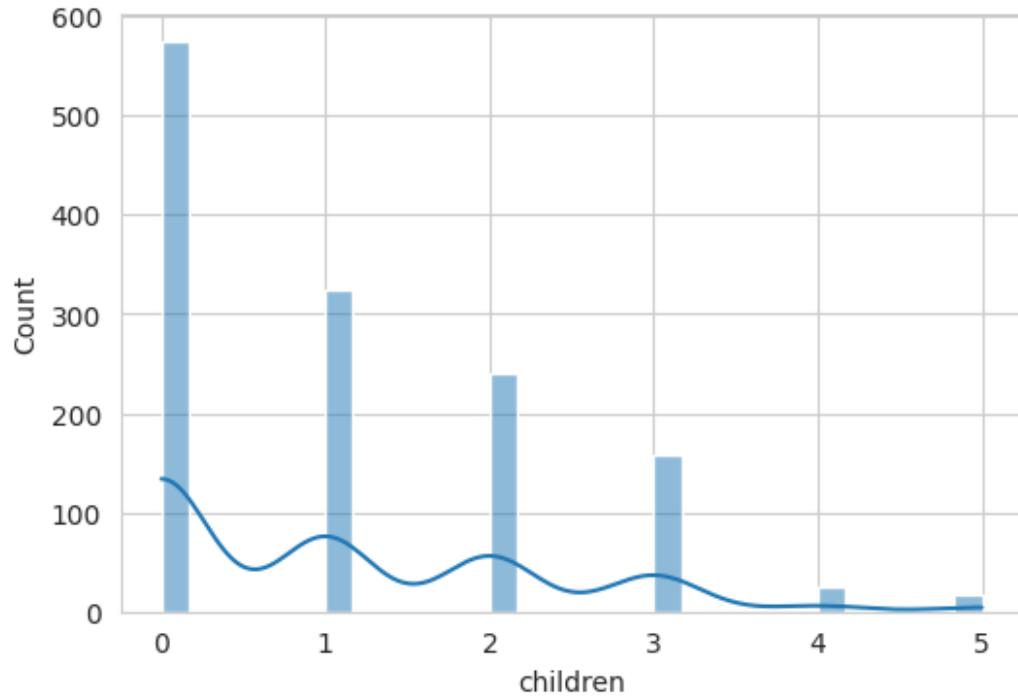
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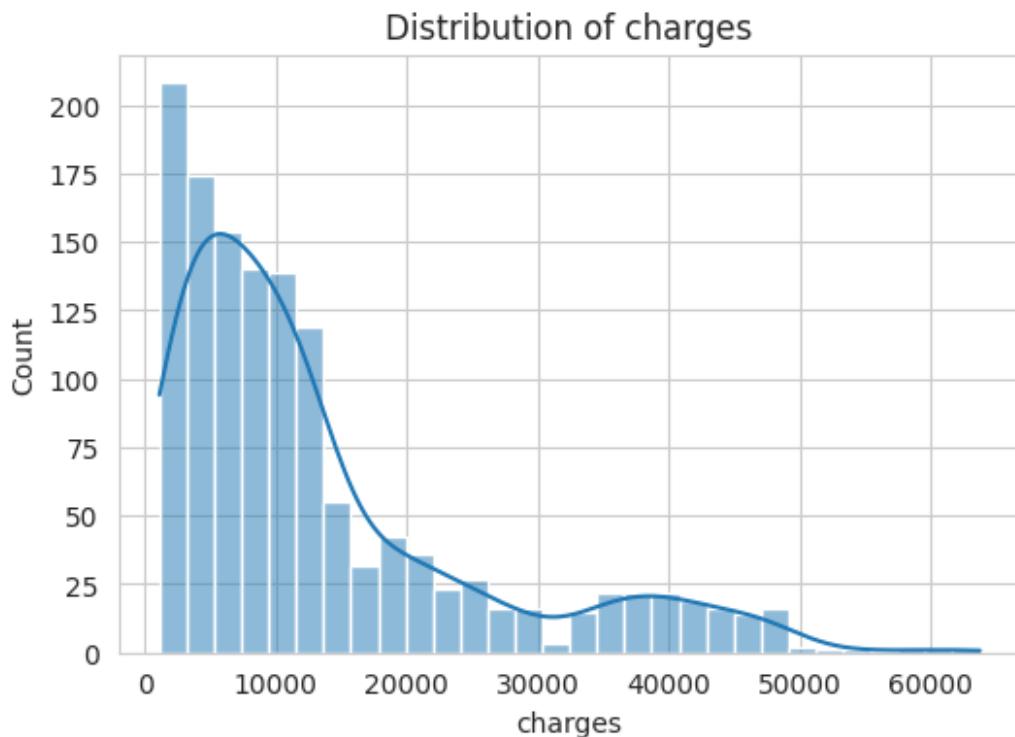


Distribution of bmi



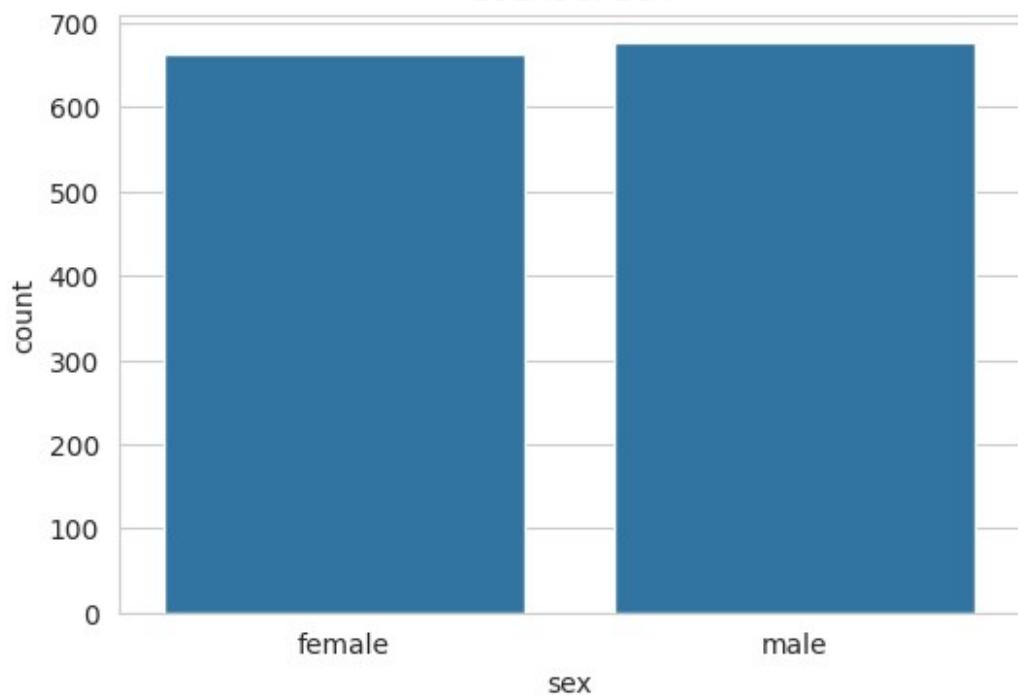
Distribution of children



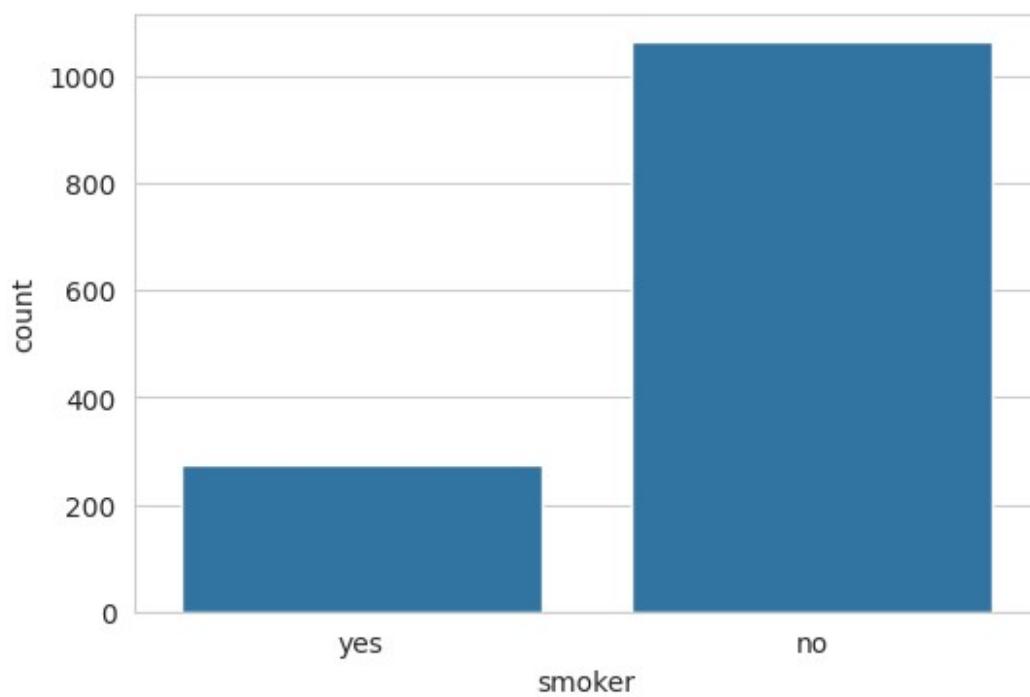


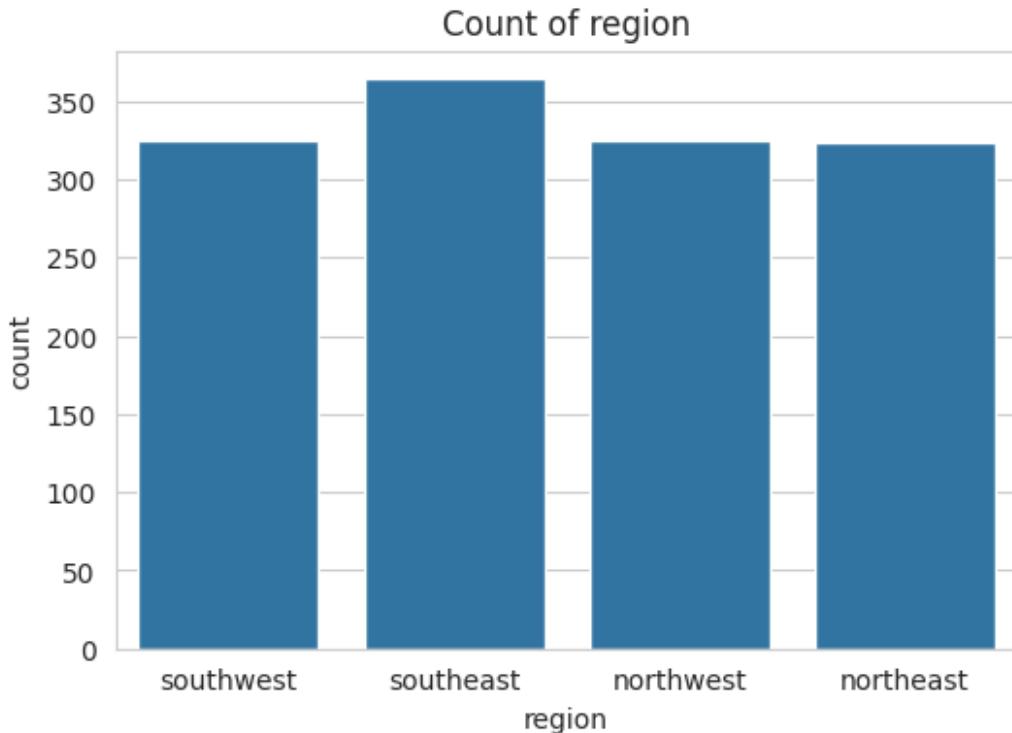
```
# Count plots for categorical features
cat_cols = df.select_dtypes(include="object").columns
for col in cat_cols:
    plt.figure(figsize=(6,4))
    sns.countplot(x=df[col])
    plt.title(f"Count of {col}")
    plt.show()
```

Count of sex



Count of smoker

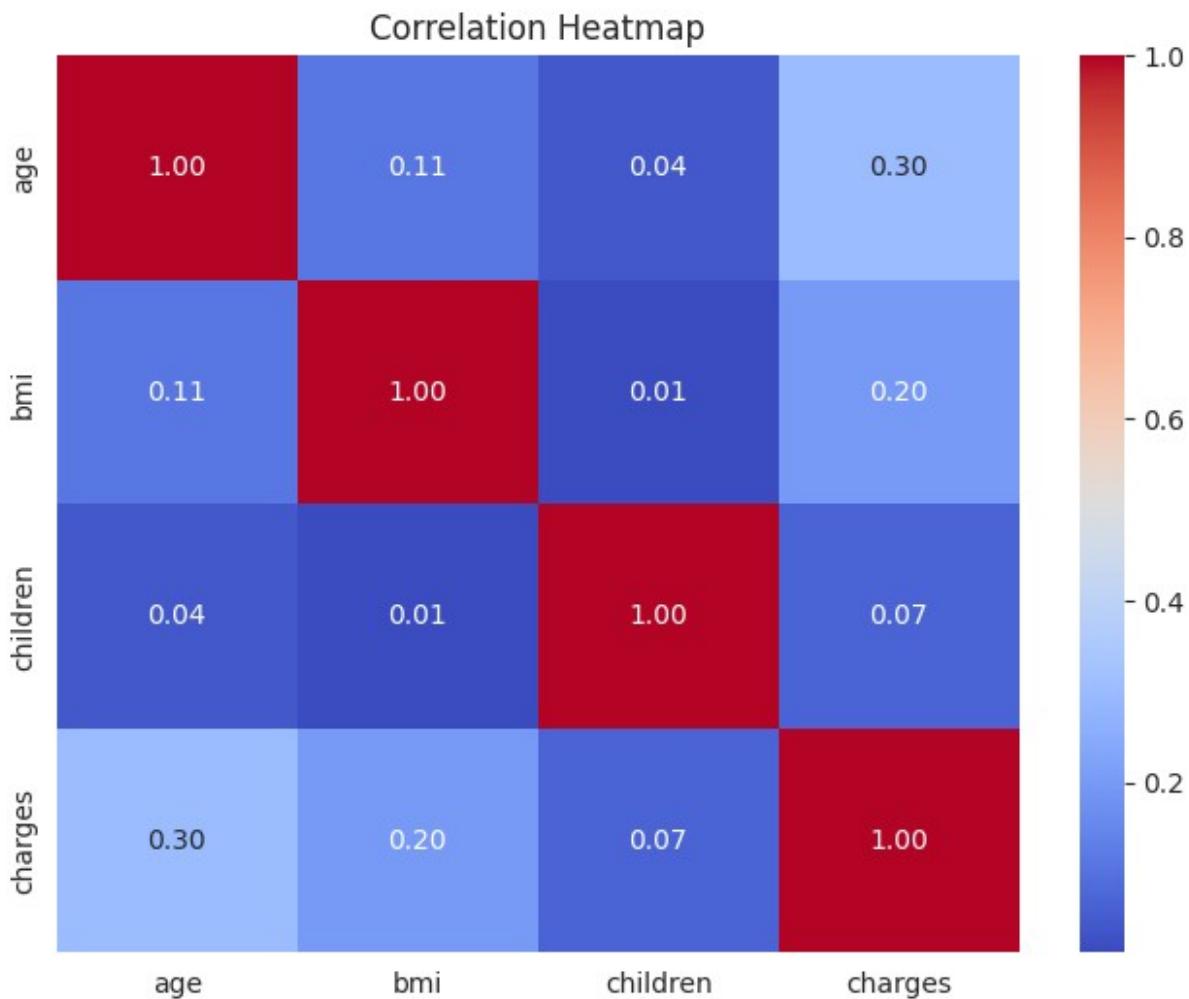




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# 5. Check for Missing Values
print("\n--- Missing Values ---")
print(df.isnull().sum())

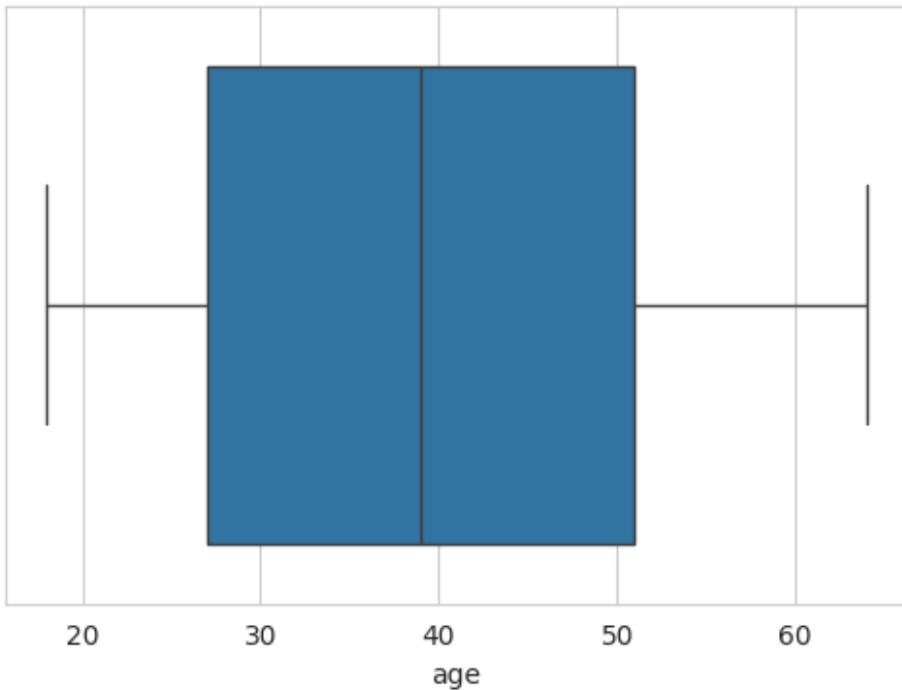
--- Missing Values ---
age        0
sex        0
bmi        0
children   0
smoker     0
region     0
charges    0
dtype: int64

# 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()
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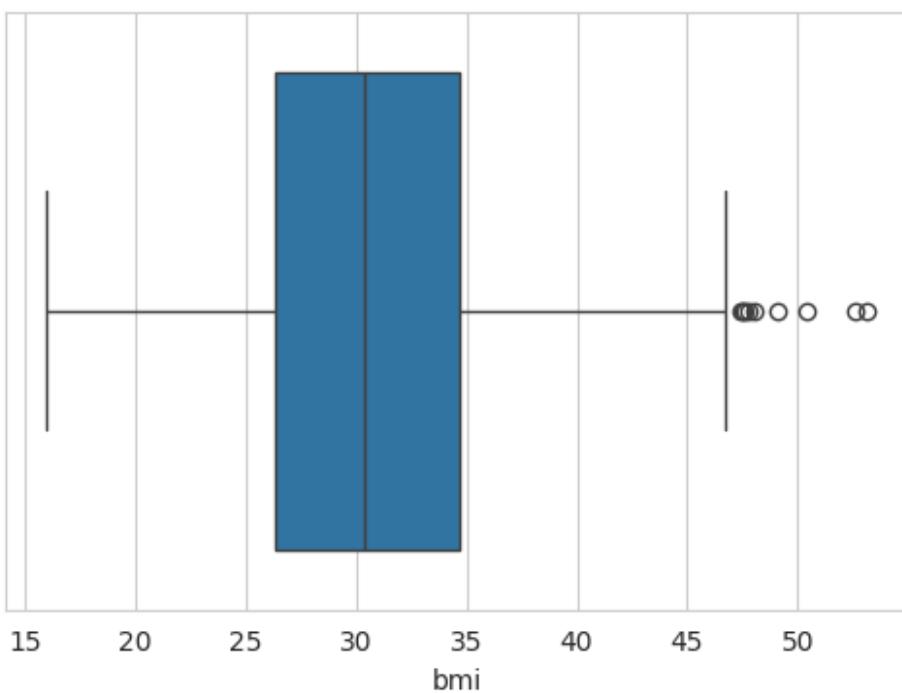


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# 7. Detect Outliers
# Using Boxplots
for col in num_cols:
    plt.figure(figsize=(6,4))
    sns.boxplot(x=df[col])
    plt.title(f"Outlier Detection for {col}")
    plt.show()
```

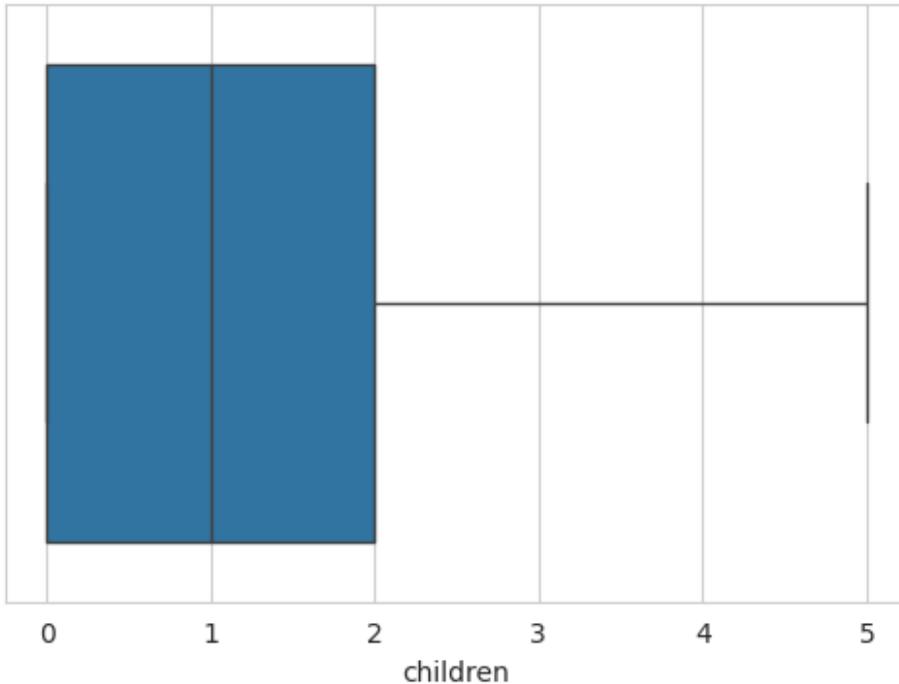
Outlier Detection for age



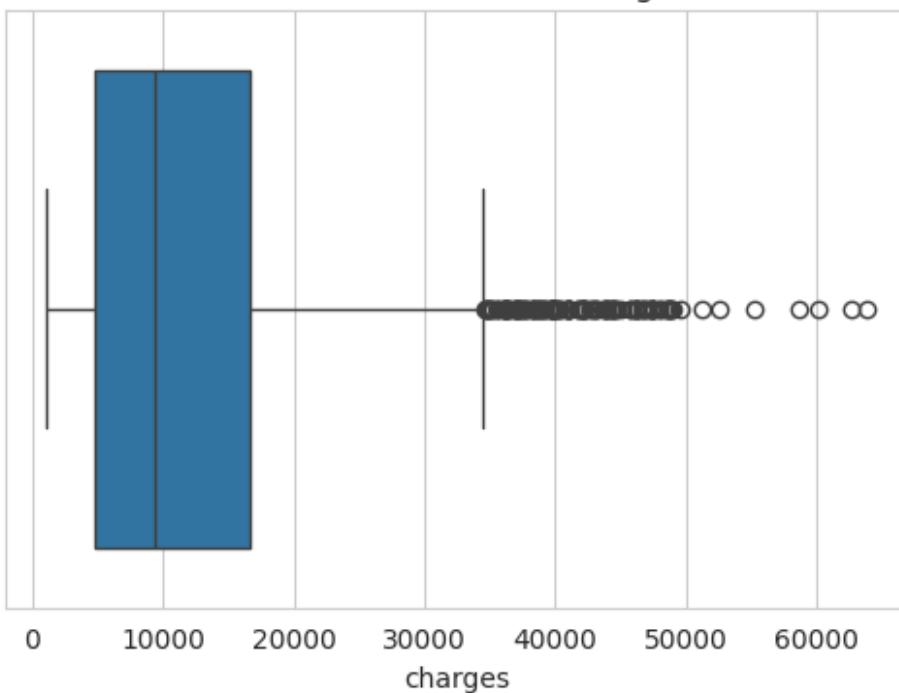
Outlier Detection for bmi



Outlier Detection for children



Outlier Detection for charges



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# 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

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Final Dataset Shape: (1338, 8)