



```
2    Newspaper   200 non-null      float64
3    Sales      200 non-null      float64
dtypes: float64(4)
memory usage: 6.4 KB

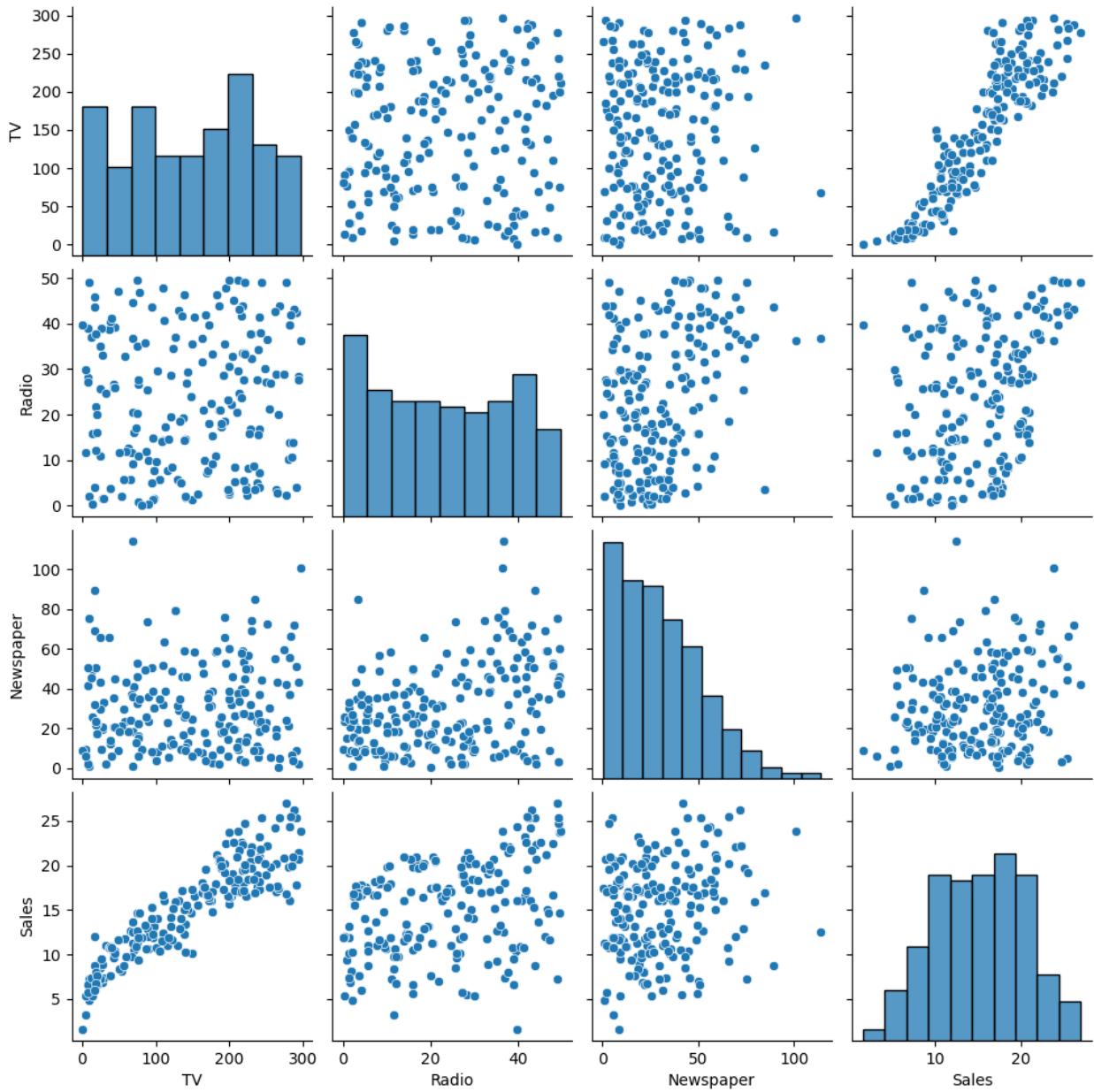
df.describe()

{"summary": {"name": "df", "rows": 8, "fields": [{"column": "TV", "properties": {"dtype": "number", "std": 93.12930693433862, "min": 0.7, "max": 296.4, "num_unique_values": 8, "samples": [147.0425, 149.75, 200.0], "semantic_type": "\\", "description": "\\n        }, "column": "Radio", "properties": {"dtype": "number", "std": 64.62946191825954, "min": 0.0, "max": 200.0, "num_unique_values": 8, "samples": [23.264000000000006, 22.9, 200.0], "semantic_type": "\\", "description": "\\n        }, "column": "Newspaper", "properties": {"dtype": "number", "std": 67.53295876114069, "min": 0.3, "max": 200.0, "num_unique_values": 8, "samples": [30.553999999999995, 25.75, 200.0], "semantic_type": "\\", "description": "\\n        }, "column": "Sales", "properties": {"dtype": "number", "std": 66.381408327359, "min": 1.6, "max": 200.0, "num_unique_values": 8, "samples": [15.13050000000001, 16.0, 200.0], "semantic_type": "\\", "description": "\\n        }}, "type": "dataframe"}}

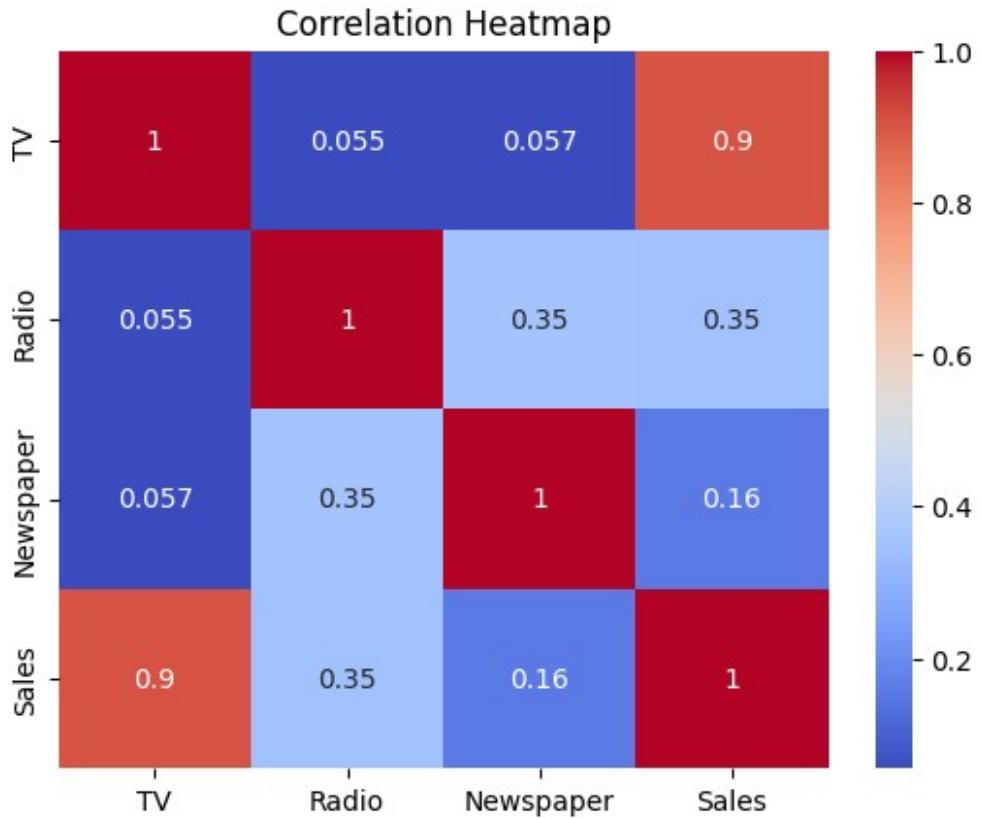
df.isnull().sum()

TV      0
Radio   0
Newspaper 0
Sales   0
dtype: int64

# @title Exploratory Data Analysis (EDA)
sns.pairplot(df)
plt.show()
```



```
sns.heatmap(df.corr(), annot=True, cmap='coolwarm')
plt.title("Correlation Heatmap")
plt.show()
```



```

# @title Feature Selection
X = df[['TV', 'Radio', 'Newspaper']]
y = df['Sales']

# @title Train-Test Split
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42
)

# @title Build Regression Model
model = LinearRegression()
model.fit(X_train, y_train)

LinearRegression()

# @title Make Predictions
y_pred = model.predict(X_test)

# @title Model Evaluation
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)

Mean Squared Error: 2.9077569102710896

```

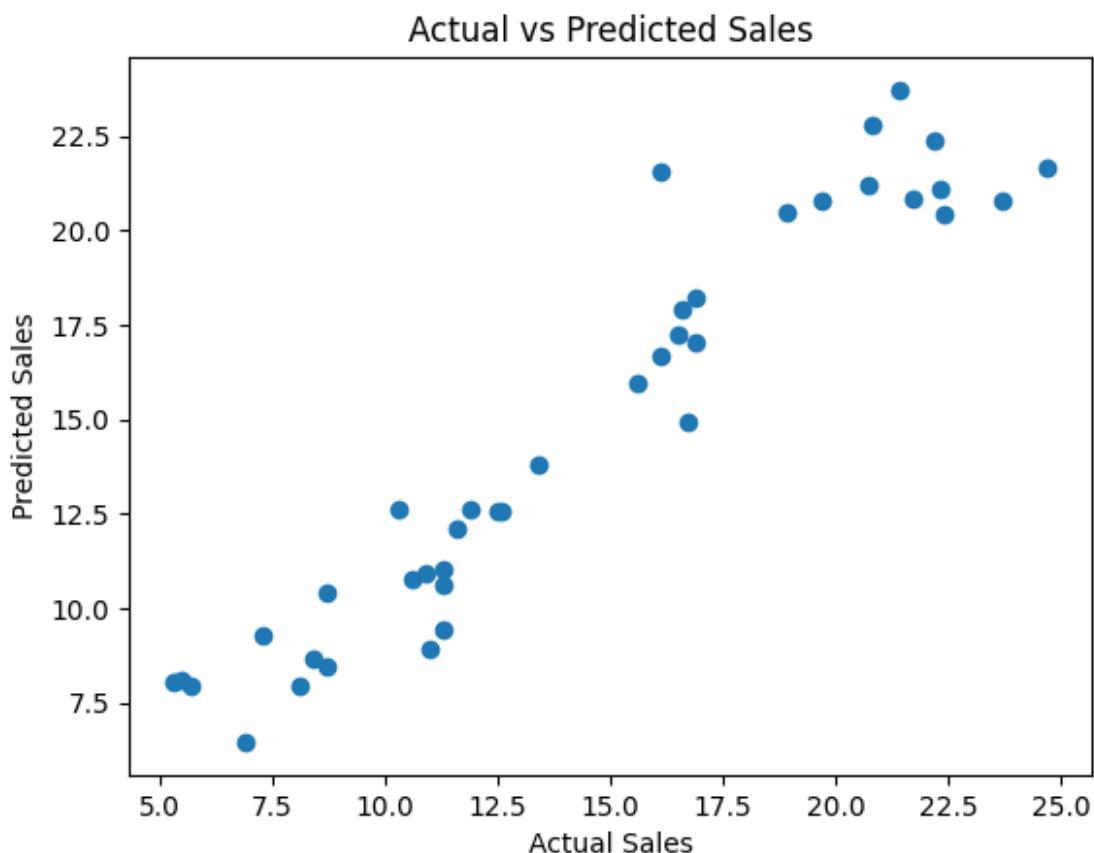
```

r2 = r2_score(y_test, y_pred)
print("R2 Score:", r2)

R2 Score: 0.9059011844150826

# @title Actual vs Predicted Visualization
plt.scatter(y_test, y_pred)
plt.xlabel("Actual Sales")
plt.ylabel("Predicted Sales")
plt.title("Actual vs Predicted Sales")
plt.show()

```



```

# @title Model Coefficients Interpretation
coeff_df = pd.DataFrame({
    'Feature': X.columns,
    'Coefficient': model.coef_
})

coeff_df

{
  "summary": {
    "name": "coeff_df",
    "rows": 3,
    "fields": [
      {
        "column": "Feature",
        "dtype": "string"
      }
    ],
    "properties": {}
  }
}

```

```
  "num_unique_values": 3,\n    "samples": [\n      "TV",\n      "Radio",\n      "Newspaper"],\n    "semantic_type": "",\n    "description": "",\n  },\n  {\n    "column": "Coefficient",\n    "properties": {\n      "dtype": "number",\n      "std": 0.04831639943091294,\n      "min": 0.0043366468220340446,\n      "max": 0.10094536239295579,\n      "num_unique_values": 3,\n      "samples": [\n        0.05450927083721978,\n        0.10094536239295579,\n        0.0043366468220340446\n      ],\n      "semantic_type": "",\n      "description": "",\n    }\n  }\n],\n"\n]\n}","type":"dataframe","variable_name":"coeff_df"}\n\n# @title Conclusion\nprint("Sales Prediction model built successfully.")\nprint("TV advertising has the highest impact on sales.")\n\nSales Prediction model built successfully.\nTV advertising has the highest impact on sales.
```