



```
import pandas as pd
import numpy as np
from sklearn.model_selection import cross_val_score
from sklearn.linear_model import LinearRegression
from sklearn.tree import DecisionTreeRegressor
from sklearn.ensemble import RandomForestRegressor
from sklearn.svm import SVR
from sklearn.neighbors import KNeighborsRegressor
from sklearn.datasets import make_regression
```

```
df= pd.read_csv("/content/tvmarketing.csv")
```

```
df.head()
```




	TV	Sales
0	230.1	22.1
1	44.5	10.4
2	17.2	9.3
3	151.5	18.5
4	180.8	12.9




Next steps:

[Generate code with df](#)
[View recommended plots](#)


```
df.tail()
```




	TV	Sales
195	38.2	7.6
196	94.2	9.7
197	177.0	12.8
198	283.6	25.5
199	232.1	13.4




```
df.shape
```

 (200, 2)


```
df.describe().T
```



	count	mean	std	min	25%	50%	75%	max
TV	200.0	147.0425	85.854236	0.7	74.375	149.75	218.825	296.4
Sales	200.0	14.0225	5.217457	1.6	10.375	12.90	17.400	27.0



```
df.info()
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 2 columns):
#   Column  Non-Null Count  Dtype
---  -
0    TV      200 non-null     float64
1   Sales  200 non-null     float64
dtypes: float64(2)
memory usage: 3.2 KB
```

```
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.metrics import r2_score
from sklearn.datasets import make_regression
from sklearn.linear_model import LinearRegression
from sklearn.tree import DecisionTreeRegressor
from sklearn.ensemble import RandomForestRegressor
from sklearn.svm import SVR
from sklearn.neighbors import KNeighborsRegressor
from sklearn.ensemble import GradientBoostingRegressor
from sklearn.neural_network import MLPRegressor

X, y = make_regression(n_samples=100, n_features=5, noise=0.1, random_state=42)

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

regressors = {
    "Linear Regression": LinearRegression(),
    "Decision Tree": DecisionTreeRegressor(),
    "Random Forest": RandomForestRegressor(),
    "Support Vector Regressor": SVR(),
    "K-Neighbors Regressor": KNeighborsRegressor(),
    "Gradient Boosting Regressor": GradientBoostingRegressor(),
    "Neural Network" : MLPRegressor(max_iter=1000)
}

import warnings

for name, model in regressors.items():
    model.fit(X_train, y_train)
    scores = cross_val_score(model, X, y, cv=50, scoring='neg_mean_squared_error')

    mse_scores = -scores
    mean_mse = np.mean(mse_scores)
    std_mse = np.std(mse_scores)
    def adjusted_r2_score(r2, n, p):
        return 1 - (1 - r2) * ((n - 1) / (n - p - 1))
```



[illegible]

```
from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score
```

```
r2 = r2_score(y_test, y_pred)
```

```
n=X_test.shape[0]
n
```

→ 20

```
p = X_test.shape[1]
p
```

5

```
results = []
```

```
for name,model in regressors.items():
    model.fit(X_train, y_train)
    y_pred = model.predict(X_test)
    r2 = r2_score(y_test, y_pred)
    n = X_test.shape[0]
    p = X_test.shape[1]
    adj_r2 = adjusted_r2_score(r2, n, p)
    mse = mean_squared_error(y_test,y_pred)
    mae = mean_absolute_error(y_test,y_pred)
    # Append results as dictionaries to the list
    results.append({"Regressor": name, "R2 Score": r2, "Adjusted R2 Score": adj_r2,"mean_squared_error":mse,"mean_absolute_error":mae})
```

```
↳ /usr/local/lib/python3.10/dist-packages/sklearn/neural_network/_multilayer_perceptron.py:686: ConvergenceWarning: Stochastic Optimizer:
  warnings.warn(
```

```
for name, model in regressors.items():
    print(f"--- {name} ---")

    metrics = next((item for item in results if item["Regressor"] == name), None)
    if metrics:
        for metric_name, value in metrics.items():
            if metric_name != 'Regressor':
                print(f"{metric_name}: {value:.4f}")
        print()

best_model_name = max(results, key=lambda k: k['Adjusted R2 Score'])
print(f"Best Model based on Adjusted R-squared: {best_model_name}")
```

```

--- Linear Regression ---
R2 Score: 1.0000
Adjusted R2 Score: 1.0000
mean_squared_error: 0.0113
mean_absolute_error: 0.0844

--- Decision Tree ---
R2 Score: 0.5186
Adjusted R2 Score: 0.3467
mean_squared_error: 9667.8080
mean_absolute_error: 77.4563

--- Random Forest ---
R2 Score: 0.7787
Adjusted R2 Score: 0.6996
mean_squared_error: 4444.9632
mean_absolute_error: 44.4607

--- Support Vector Regressor ---
R2 Score: 0.0405
Adjusted R2 Score: -0.3022
mean_squared_error: 19269.5792
mean_absolute_error: 104.6839

--- K-Neighbors Regressor ---
R2 Score: 0.7146
Adjusted R2 Score: 0.6127
mean_squared_error: 5730.6648
mean_absolute_error: 49.6629

--- Gradient Boosting Regressor ---
R2 Score: 0.8464
Adjusted R2 Score: 0.7916
mean_squared_error: 3084.1109
mean_absolute_error: 41.4867

--- Neural Network ---
R2 Score: 0.9696
Adjusted R2 Score: 0.9587
mean_squared_error: 611.2466
mean_absolute_error: 18.4604

Best Model based on Adjusted R-squared: {'Regressor': 'Linear Regression', 'R2 Score': 0.999994350808352, 'Adjusted R2 Score': 0.99999

```

```

from pandas import DataFrame
results_df = pd.DataFrame(results)

```

```
print(results_df)
```

```

Regressor R2 Score Adjusted R2 Score \
0 Linear Regression 0.999999 0.999999
1 Decision Tree 0.518587 0.346654
2 Random Forest 0.778661 0.699612
3 Support Vector Regressor 0.040463 -0.302229
4 K-Neighbors Regressor 0.714639 0.612725
5 Gradient Boosting Regressor 0.846425 0.791577
6 Neural Network 0.969563 0.958692

mean_squared_error mean_absolute_error
0 0.011345 0.084376
1 9667.807992 77.456259
2 4444.963239 44.460679
3 19269.579169 104.683863
4 5730.664764 49.662876
5 3084.110912 41.486681
6 611.246629 18.460399

```

```

def adjusted_r2_score(r2, n, p):
    return 1 - (1 - r2) * ((n - 1) / (n - p - 1))

```

```
print(results_df)
```

```

Regressor R2 Score Adjusted R2 Score \
0 Linear Regression 0.999999 0.999999
1 Decision Tree 0.518587 0.346654
2 Random Forest 0.778661 0.699612
3 Support Vector Regressor 0.040463 -0.302229
4 K-Neighbors Regressor 0.714639 0.612725
5 Gradient Boosting Regressor 0.846425 0.791577

```

6	Neural Network	0.969563	0.958692
	mean_squared_error	mean_absolute_error	
0	0.011345	0.084376	
1	9667.807992	77.456259	
2	4444.963239	44.460679	
3	19269.579169	104.683863	
4	5730.664764	49.662876	
5	3084.110912	41.486681	
6	611.246629	18.460399	

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