

```
In [11]: import pandas as pd
import numpy as np
import seaborn as sb
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeRegressor
from sklearn.metrics import accuracy_score
from sklearn.metrics import mean_squared_error, r2_score
```

```
In [12]: df = pd.read_csv('Alumni Giving Regression (Edited).csv')
```

```
In [13]: df.head()
```

```
Out[13]:
```

	A	B	C	D	E	F
0	24	0.42	0.16	0.59	0.81	0.08
1	19	0.49	0.04	0.37	0.69	0.11
2	18	0.24	0.17	0.66	0.87	0.31
3	8	0.74	0.00	0.81	0.88	0.11
4	8	0.95	0.00	0.86	0.92	0.28

```
In [14]: df.describe()
```

```
Out[14]:
```

	A	B	C	D	E	F
count	123.000000	123.000000	123.000000	123.000000	123.000000	123.000000
mean	17.772358	0.403659	0.136260	0.645203	0.841138	0.141789
std	4.517385	0.133897	0.060101	0.169794	0.083942	0.080674
min	6.000000	0.140000	0.000000	0.260000	0.580000	0.020000
25%	16.000000	0.320000	0.095000	0.505000	0.780000	0.080000
50%	18.000000	0.380000	0.130000	0.640000	0.840000	0.130000
75%	20.000000	0.460000	0.180000	0.785000	0.910000	0.170000
max	31.000000	0.950000	0.310000	0.960000	0.980000	0.410000

```
In [15]: X = df[['A', 'B', 'C', 'D', 'E']]
y = df['F']
```

```
In [16]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
In [17]: model = DecisionTreeRegressor()
```

```
In [18]: model.fit(X_train, y_train)
```

```
Out[18]: DecisionTreeRegressor()
```

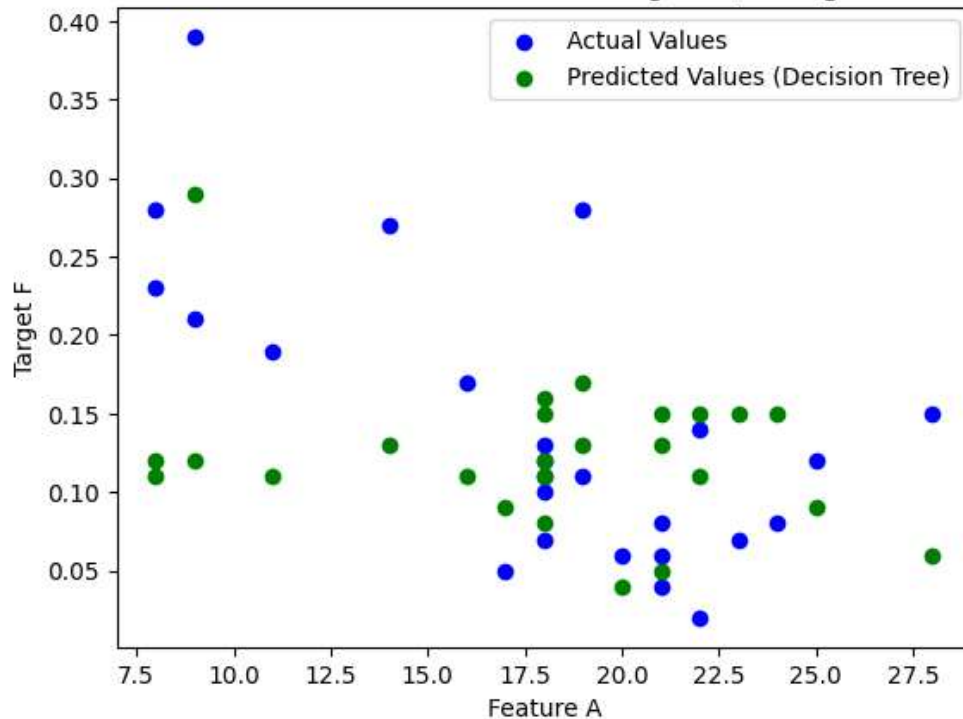
**In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.**

**On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.**

```
In [34]: y_pred = model.predict(X_test)
```

```
In [39]: plt.scatter(X_test['A'], y_test, color='blue', label='Actual Values')
plt.scatter(X_test['A'], y_pred, color='green', label='Predicted Values (Decision Tree)')
plt.xlabel("Feature A")
plt.ylabel("Target F")
plt.title("Actual vs Predicted Values for Feature A and Target F (Testing Set - Decision Tree)")
plt.legend()
plt.show()
```

Actual vs Predicted Values for Feature A and Target F (Testing Set - Decision Tree)



```
In [1]: import pandas as pd
import numpy as np
import seaborn as sb
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error
```

```
In [2]: df = pd.read_csv('Alumni Giving Regression (Edited).csv')
```

```
In [3]: df.head()
```

```
Out[3]:
```

	A	B	C	D	E	F
0	24	0.42	0.16	0.59	0.81	0.08
1	19	0.49	0.04	0.37	0.69	0.11
2	18	0.24	0.17	0.66	0.87	0.31
3	8	0.74	0.00	0.81	0.88	0.11
4	8	0.95	0.00	0.86	0.92	0.28

```
In [4]: df.describe()
```

```
Out[4]:
```

	A	B	C	D	E	F
count	123.000000	123.000000	123.000000	123.000000	123.000000	123.000000
mean	17.772358	0.403659	0.136260	0.645203	0.841138	0.141789
std	4.517385	0.133897	0.060101	0.169794	0.083942	0.080674
min	6.000000	0.140000	0.000000	0.260000	0.580000	0.020000
25%	16.000000	0.320000	0.095000	0.505000	0.780000	0.080000
50%	18.000000	0.380000	0.130000	0.640000	0.840000	0.130000
75%	20.000000	0.460000	0.180000	0.785000	0.910000	0.170000
max	31.000000	0.950000	0.310000	0.960000	0.980000	0.410000

```
In [5]: X = df[['A', 'B', 'C', 'D', 'E']]
y = df['F']
```

```
In [6]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random
```

```
In [7]: model = RandomForestRegressor(n_estimators=100, random_state=42)
```

```
In [8]: model.fit(X_train, y_train)
```

```
Out[8]: RandomForestRegressor(random_state=42)
```

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```
In [9]: y_pred = model.predict(X_test)
```

```
In [10]: mse = mean_squared_error(y_test, y_pred)
print(f"Mean Squared Error: {mse}")
```

Mean Squared Error: 0.0038105840000000005

```
In [12]: plt.scatter(X_test['A'], y_test, color='blue', label='Actual Values')
plt.scatter(X_test['A'], y_pred, color='red', label='Predicted Values')
plt.xlabel("Feature A")
plt.ylabel("Target F")
plt.title("Actual vs Predicted Values for Feature A and Target F (Testing Set)")
plt.legend()
plt.show()
```

