

# ANN

While the ANN model can be further fine-tuned to increase its accuracy, such as increasing the number of layers and dropout, it is time-consuming and inefficient. Therefore, we focused more on other ML models.

```
In [1]: import pandas as pd
import numpy as np
import math
import tensorflow as tf
import tensorflow.keras as keras
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Activation
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import MinMaxScaler
from sklearn.inspection import permutation_importance
from sklearn.preprocessing import StandardScaler
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import r2_score
from collections import defaultdict
from sklearn import metrics
from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score
```

```
In [2]: # Import dataset
private_data = "../datasets/cleaned/cleaned_private.csv"

df = pd.read_csv(private_data, quotechar='\"', escapechar='\\', thousands=',')
df['Sale Month-Year'] = pd.to_datetime(df['Sale Date']).dt.to_period('M').astype(str)
```

```
In [ ]: def remove_outliers_iqr(df, column):
    q1 = df[column].quantile(0.25)
    q3 = df[column].quantile(0.75)
    iqr = q3 - q1
    lower_bound = q1 - 1.5 * iqr
    upper_bound = q3 + 1.5 * iqr
    return df[(df[column] >= lower_bound) & (df[column] <= upper_bound)]

# Apply outlier removal to Price only
for col in ['Price']:
    df = remove_outliers_iqr(df, col)
```

```
In [ ]: df_features=['Area (SQFT)', 'Lease_Category', 'Property Type', 'Postal District',
                    'Type of Sale', 'Floor Level', 'Sale Month-Year', 'Distance to MRT (km)']
X = df[df_features]
y = df['Price'].values

categorical_cols = ['Property Type', 'Postal District', 'Type of Sale', 'Floor Level', 'Sale Month-Year', 'Lease_Cat']
numerical_cols = ['Area (SQFT)', 'Distance to MRT (km)']

X_encoded = pd.get_dummies(X[categorical_cols], drop_first=True) # split all categorical data into several columns w

# scaling numerical features
scaler = StandardScaler()
X_numerical_scaled = scaler.fit_transform(X[numerical_cols])
X_numerical_df = pd.DataFrame(X_numerical_scaled, columns=numerical_cols)

# combining all the features into 1 dataframe
X_final = pd.concat([X_numerical_df, X_encoded], axis=1)
X_final = pd.concat([X[numerical_cols], X_encoded], axis=1)

x_train, x_test, y_train, y_test = train_test_split(X_final, y, test_size=0.25, random_state=40)
x_train = x_train.values.astype(np.float32)
x_test = x_test.values.astype(np.float32)
x_test = x_test.astype('float32') if isinstance(x_test, pd.DataFrame) else x_test.astype('float32')
y_test = y_test.astype('float32') if isinstance(y_test, pd.DataFrame) else y_test.astype('float32')
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)

(88744, 41)
(29582, 41)
(88744,)
(29582,)
```

```
In [ ]: # Define model builder
model=Sequential()
model.add(Dense(64, activation='relu'))
```

```

model.add(Dense(64, activation='relu'))
model.add(Dense(64, activation='relu'))
model.add(Dense(64, activation='relu'))
model.add(Dense(1))
model.compile(optimizer='adam', loss='mse')
model_result = model.fit(x_train, y_train, epochs=200)

class KerasModelWrapper:
    def __init__(self, model):
        self.model = model

    def fit(self, X, y):
        pass # already trained outside

    def predict(self, X):
        return self.model.predict(X).flatten()

    def score(self, X, y):
        y_pred = self.predict(X)
        return r2_score(y, y_pred)

wrapped_model = KerasModelWrapper(model)

# Use sklearn's permutation importance
result = permutation_importance(wrapped_model, x_test, y_test, n_repeats=10, random_state=42)

# Sort and print feature importances
sorted_idx = result.importances_mean.argsort()[::-1]

print("Feature importances (descending):\n")
feature_names = X_final.columns

# Map each one-hot encoded column back to its base category
grouped_importances = defaultdict(float)

for i, col in enumerate(feature_names):
    # Split by underscore only if it's one-hot encoded
    if '_' in col:
        base_feature = col.split('_')[0]
    else:
        base_feature = col # numerical feature (not one-hot encoded)

    grouped_importances[base_feature] += result.importances_mean[i]

# Sort and print the aggregated importances
sorted_importances = sorted(grouped_importances.items(), key=lambda x: x[1], reverse=True)

print("Aggregated Feature Importances:\n")
for feature, importance in sorted_importances:
    print(f"{feature}: {importance:.6f}")

features, importances = zip(*sorted_importances)

# Plot
plt.figure(figsize=(10, 6))
bars = plt.barh(features, importances, color='skyblue')
plt.xlabel("Importance")
plt.title("Aggregated Feature Importances")
plt.gca().invert_yaxis() # Highest importance on top

for bar in bars:
    width = bar.get_width()
    plt.text(width + 0.001,
             bar.get_y() + bar.get_height() / 2,
             f"{width:.4f}",
             va='center')

plt.tight_layout()
plt.show()

```

Epoch 1/200  
2774/2774 [=====] - 14s 4ms/step - loss: 7974306709504.0000  
Epoch 2/200  
2774/2774 [=====] - 11s 4ms/step - loss: 3807742001152.0000  
Epoch 3/200  
2774/2774 [=====] - 11s 4ms/step - loss: 2851716726784.0000  
Epoch 4/200  
2774/2774 [=====] - 15s 5ms/step - loss: 2777113165824.0000  
Epoch 5/200  
2774/2774 [=====] - 14s 5ms/step - loss: 3047600422912.0000  
Epoch 6/200  
2774/2774 [=====] - 13s 5ms/step - loss: 2311380533248.0000  
Epoch 7/200  
2774/2774 [=====] - 15s 5ms/step - loss: 2068711342080.0000  
Epoch 8/200  
2774/2774 [=====] - 9s 3ms/step - loss: 1952607764480.0000  
Epoch 9/200  
2774/2774 [=====] - 11s 4ms/step - loss: 1543747796992.0000  
Epoch 10/200  
2774/2774 [=====] - 11s 4ms/step - loss: 1871928229888.0000  
Epoch 11/200  
2774/2774 [=====] - 11s 4ms/step - loss: 1533951213568.0000  
Epoch 12/200  
2774/2774 [=====] - 12s 4ms/step - loss: 1392954179584.0000  
Epoch 13/200  
2774/2774 [=====] - 20s 7ms/step - loss: 1111712858112.0000  
Epoch 14/200  
2774/2774 [=====] - 27s 10ms/step - loss: 1078364340224.0000  
Epoch 15/200  
2774/2774 [=====] - 23s 8ms/step - loss: 1062859964416.0000  
Epoch 16/200  
2774/2774 [=====] - 25s 9ms/step - loss: 913377787904.0000  
Epoch 17/200  
2774/2774 [=====] - 24s 9ms/step - loss: 969691758592.0000  
Epoch 18/200  
2774/2774 [=====] - 24s 9ms/step - loss: 899877896192.0000  
Epoch 19/200  
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Epoch 20/200  
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Epoch 21/200  
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Epoch 22/200  
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Epoch 23/200  
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Epoch 25/200  
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Epoch 26/200  
2774/2774 [=====] - 30s 11ms/step - loss: 906597236736.0000  
Epoch 27/200  
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Epoch 28/200  
2774/2774 [=====] - 27s 10ms/step - loss: 867282452480.0000  
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Epoch 30/200  
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Epoch 31/200  
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Epoch 32/200  
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Epoch 33/200  
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Epoch 34/200  
2774/2774 [=====] - 28s 10ms/step - loss: 718005796864.0000  
Epoch 35/200  
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Epoch 36/200  
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Epoch 37/200  
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Epoch 38/200  
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Epoch 39/200  
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Epoch 40/200  
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Epoch 43/200

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Epoch 54/200  
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Epoch 55/200  
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Epoch 56/200  
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Epoch 61/200  
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Epoch 64/200  
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Epoch 70/200  
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Epoch 73/200  
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Epoch 86/200  
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Epoch 158/200  
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2774/2774 [=====] - 6s 2ms/step - loss: 65322582016.0000  
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2774/2774 [=====] - 6s 2ms/step - loss: 65654599680.0000  
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Epoch 163/200  
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Epoch 166/200  
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2774/2774 [=====] - 3s 1ms/step - loss: 68502523904.0000

Epoch 171/200  
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Epoch 172/200  
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2774/2774 [=====] - 3s 1ms/step - loss: 66163216384.0000  
Epoch 176/200  
2774/2774 [=====] - 3s 1ms/step - loss: 63833214976.0000  
Epoch 177/200  
2774/2774 [=====] - 3s 1ms/step - loss: 63053008896.0000  
Epoch 178/200  
2774/2774 [=====] - 3s 1ms/step - loss: 63650988032.0000  
Epoch 179/200  
2774/2774 [=====] - 3s 1ms/step - loss: 62971072512.0000  
Epoch 180/200  
2774/2774 [=====] - 3s 1ms/step - loss: 62928986112.0000  
Epoch 181/200  
2774/2774 [=====] - 3s 1ms/step - loss: 64751808512.0000  
Epoch 182/200  
2774/2774 [=====] - 3s 1ms/step - loss: 62252748800.0000  
Epoch 183/200  
2774/2774 [=====] - 3s 1ms/step - loss: 61849423872.0000  
Epoch 184/200  
2774/2774 [=====] - 3s 1ms/step - loss: 62092681216.0000  
Epoch 185/200  
2774/2774 [=====] - 3s 1ms/step - loss: 61685399552.0000  
Epoch 186/200  
2774/2774 [=====] - 3s 1ms/step - loss: 62243696640.0000  
Epoch 187/200  
2774/2774 [=====] - 3s 1ms/step - loss: 60905996288.0000  
Epoch 188/200  
2774/2774 [=====] - 3s 1ms/step - loss: 60919500800.0000  
Epoch 189/200  
2774/2774 [=====] - 3s 1ms/step - loss: 60728647680.0000  
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2774/2774 [=====] - 3s 1ms/step - loss: 60717846528.0000  
Epoch 191/200  
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Epoch 192/200  
2774/2774 [=====] - 3s 1ms/step - loss: 61991444480.0000  
Epoch 193/200  
2774/2774 [=====] - 3s 1ms/step - loss: 59417665536.0000  
Epoch 194/200  
2774/2774 [=====] - 3s 1ms/step - loss: 59146919936.0000  
Epoch 195/200  
2774/2774 [=====] - 3s 1ms/step - loss: 59975979008.0000  
Epoch 196/200  
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Epoch 197/200  
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Epoch 198/200  
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Epoch 199/200  
2774/2774 [=====] - 3s 1ms/step - loss: 58482339840.0000  
Epoch 200/200  
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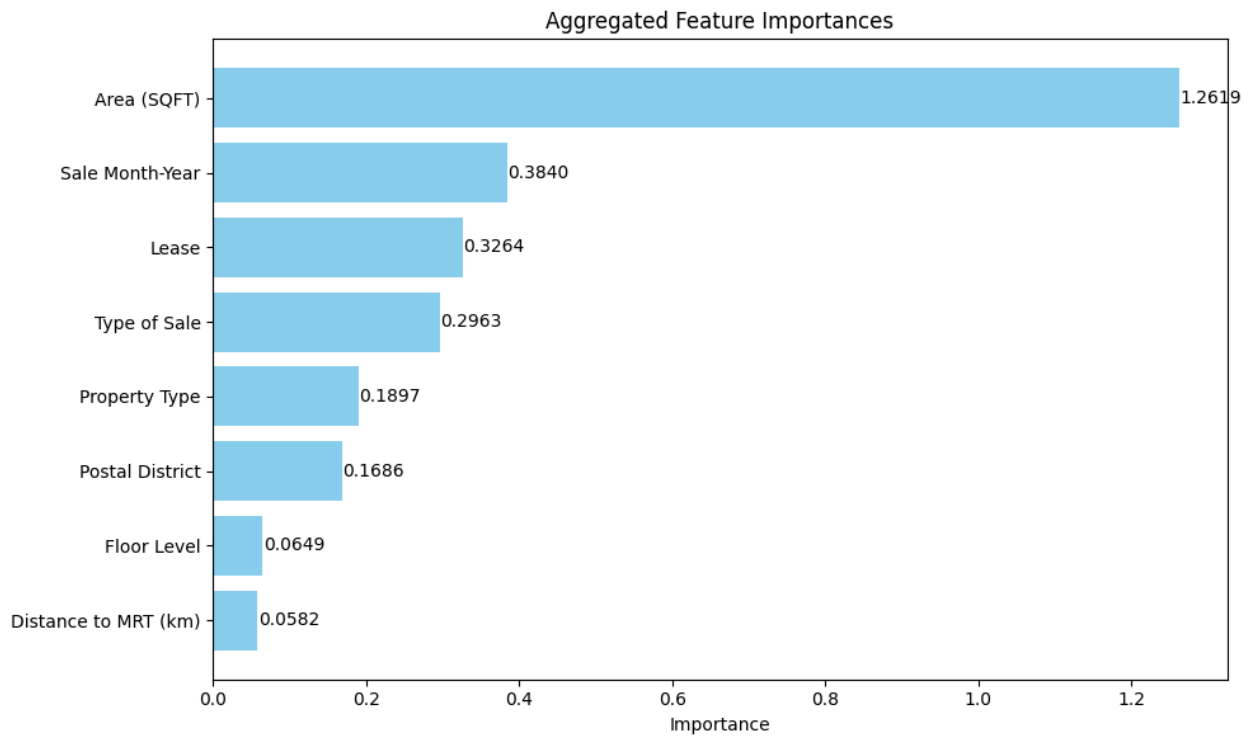
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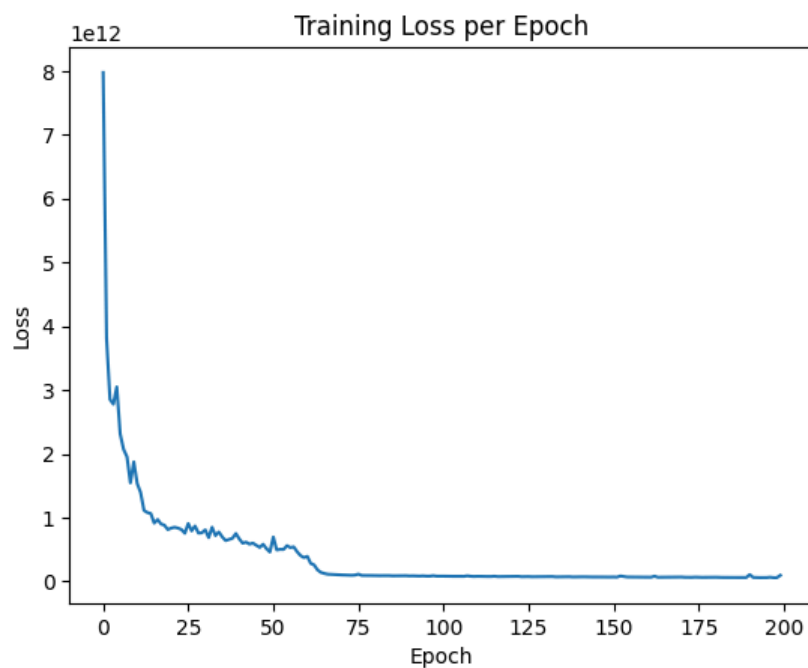
Feature importances (descending):

Aggregated Feature Importances:

Area (SQFT): 1.261851  
Sale Month-Year: 0.384004  
Lease: 0.326418  
Type of Sale: 0.296333  
Property Type: 0.189708  
Postal District: 0.168648  
Floor Level: 0.064866  
Distance to MRT (km): 0.058170



```
In [ ]: # training loss
loss = model_result.history['loss']
sns.lineplot(x=range(len(loss)),y=loss)
plt.title("Training Loss per Epoch")
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.show()
```



```
In [17]: test_predictions = wrapped_model.predict(x_test)
df_pred=pd.DataFrame({'test_actual': y_test})
df_pred['test_pred']=test_predictions
df_pred.head()
```

925/925 [=====] - 1s 607us/step

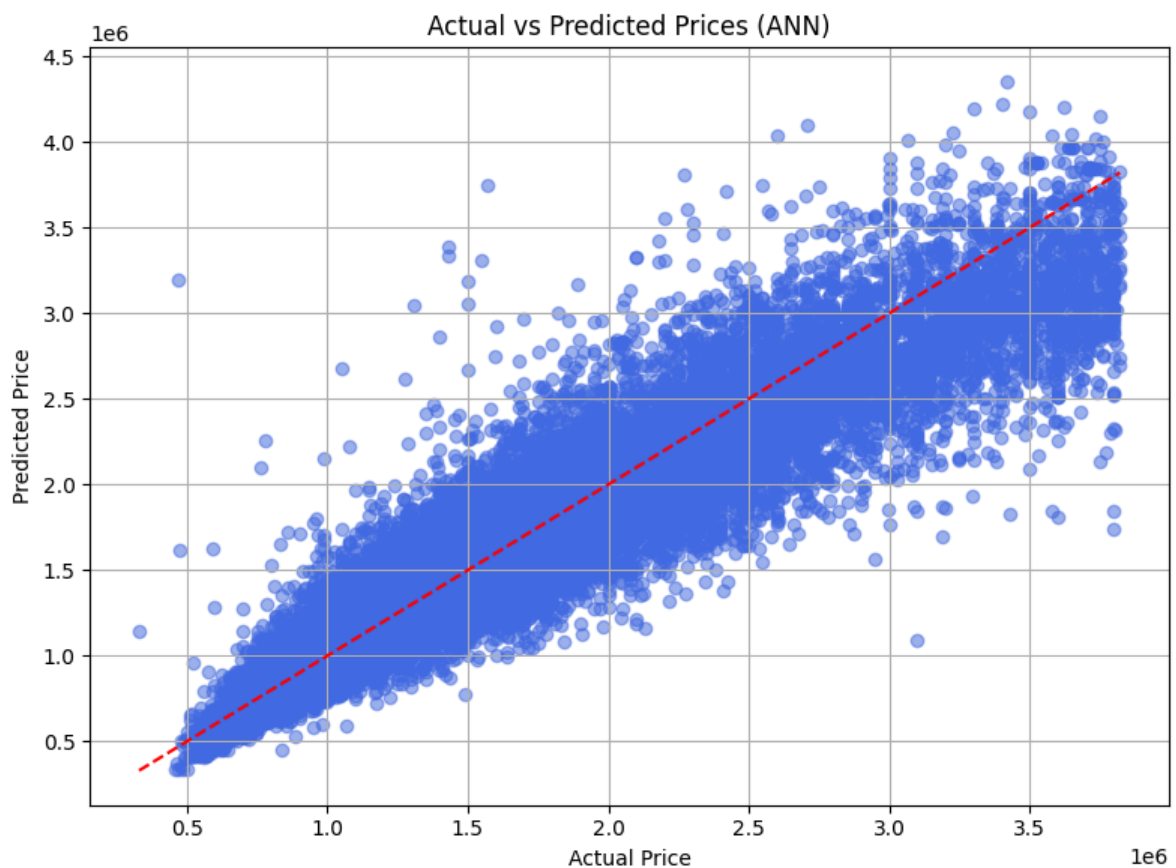
```
Out[17]:
```

	test_actual	test_pred
0	3675000.0	3.333120e+06
1	1447300.0	1.330068e+06
2	960000.0	1.014468e+06
3	1060000.0	1.261895e+06
4	1206000.0	1.267639e+06

```
In [19]: # Predict using the ANN model
y_pred = model.predict(x_test).flatten()

# Scatter Plot: Actual vs Predicted Prices
plt.figure(figsize=(8, 6))
plt.scatter(y_test, y_pred, alpha=0.5, color='royalblue')
plt.plot([y_test.min(), y_test.max()], [y_test.min(), y_test.max()], 'r--') # perfect prediction line
plt.xlabel('Actual Price')
plt.ylabel('Predicted Price')
plt.title('Actual vs Predicted Prices (ANN)')
plt.grid(True)
plt.tight_layout()
plt.show()
```

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```
In [ ]: #find rmse score
mse = mean_squared_error(df_pred['test_actual'], df_pred['test_pred'])
rmse = math.sqrt(mse)
print(f"RMSE: {rmse}")

mse = mean_squared_error(df_pred['test_actual'], df_pred['test_pred'])
mae = mean_absolute_error(df_pred['test_actual'], df_pred['test_pred'])
r2 = r2_score(df_pred['test_actual'], df_pred['test_pred'])

print(f"Mean Squared Error (MSE): {mse:.2f}")
print(f"Mean Absolute Error (MAE): {mae:.2f}")
print(f"R^2 Score: {r2:.4f}")
```

```
RMSE: 249481.61294973223
Mean Squared Error (MSE): 62241075200.00
Mean Absolute Error (MAE): 173008.09
R^2 Score: 0.8737
```

```
In [ ]:
```