

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 [3]: 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']:
    # for col in ['Price', 'Area (SQFT)', 'Remaining Lease Years']:
        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)

scaler = StandardScaler()
X_numerical_scaled = scaler.fit_transform(X[numerical_cols])
X_numerical_df = pd.DataFrame(X_numerical_scaled, columns=numerical_cols)
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) # split all categorical data into several columns with 1/0 for each column
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'))
```

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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
2774/2774 [=====] - 25s 9ms/step - loss: 880695115776.0000
Epoch 20/200
2774/2774 [=====] - 24s 9ms/step - loss: 811646255104.0000
Epoch 21/200
2774/2774 [=====] - 26s 9ms/step - loss: 834768601088.0000
Epoch 22/200
2774/2774 [=====] - 28s 10ms/step - loss: 847597666304.0000
Epoch 23/200
2774/2774 [=====] - 21s 8ms/step - loss: 834325053440.0000
Epoch 24/200
2774/2774 [=====] - 26s 10ms/step - loss: 812631654400.0000
Epoch 25/200
2774/2774 [=====] - 27s 10ms/step - loss: 754191106048.0000
Epoch 26/200
2774/2774 [=====] - 30s 11ms/step - loss: 906597236736.0000
Epoch 27/200
2774/2774 [=====] - 30s 11ms/step - loss: 791592173568.0000
Epoch 28/200
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Epoch 29/200
2774/2774 [=====] - 26s 9ms/step - loss: 757165457408.0000
Epoch 30/200
2774/2774 [=====] - 27s 10ms/step - loss: 761216761856.0000
Epoch 31/200
2774/2774 [=====] - 29s 10ms/step - loss: 807524892672.0000
Epoch 32/200
2774/2774 [=====] - 27s 10ms/step - loss: 685213220864.0000
Epoch 33/200
2774/2774 [=====] - 30s 11ms/step - loss: 848152297472.0000
Epoch 34/200
2774/2774 [=====] - 28s 10ms/step - loss: 718005796864.0000
Epoch 35/200
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Epoch 36/200
2774/2774 [=====] - 27s 10ms/step - loss: 696069521408.0000
Epoch 37/200
2774/2774 [=====] - 30s 11ms/step - loss: 639792513024.0000
Epoch 38/200
2774/2774 [=====] - 31s 11ms/step - loss: 656524771328.0000
Epoch 39/200
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Epoch 40/200
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Epoch 41/200
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Epoch 42/200
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Epoch 43/200

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Epoch 44/200
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Epoch 45/200
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Epoch 46/200
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Epoch 47/200
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Epoch 48/200
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Epoch 50/200
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Epoch 51/200
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Epoch 52/200
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Epoch 53/200
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Epoch 55/200
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Epoch 59/200
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Epoch 60/200
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Epoch 61/200
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Epoch 62/200
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Epoch 63/200
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Epoch 64/200
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Epoch 65/200
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Epoch 66/200
2774/2774 [=====] - 3s 931us/step - loss: 123514929152.0000
Epoch 67/200
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2774/2774 [=====] - 3s 911us/step - loss: 108336922624.0000
Epoch 70/200
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Epoch 73/200
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2774/2774 [=====] - 3s 920us/step - loss: 95675981824.0000
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Epoch 86/200
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2774/2774 [=====] - 6s 2ms/step - loss: 65322582016.0000
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Epoch 166/200
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Epoch 170/200
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Epoch 171/200
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Epoch 184/200
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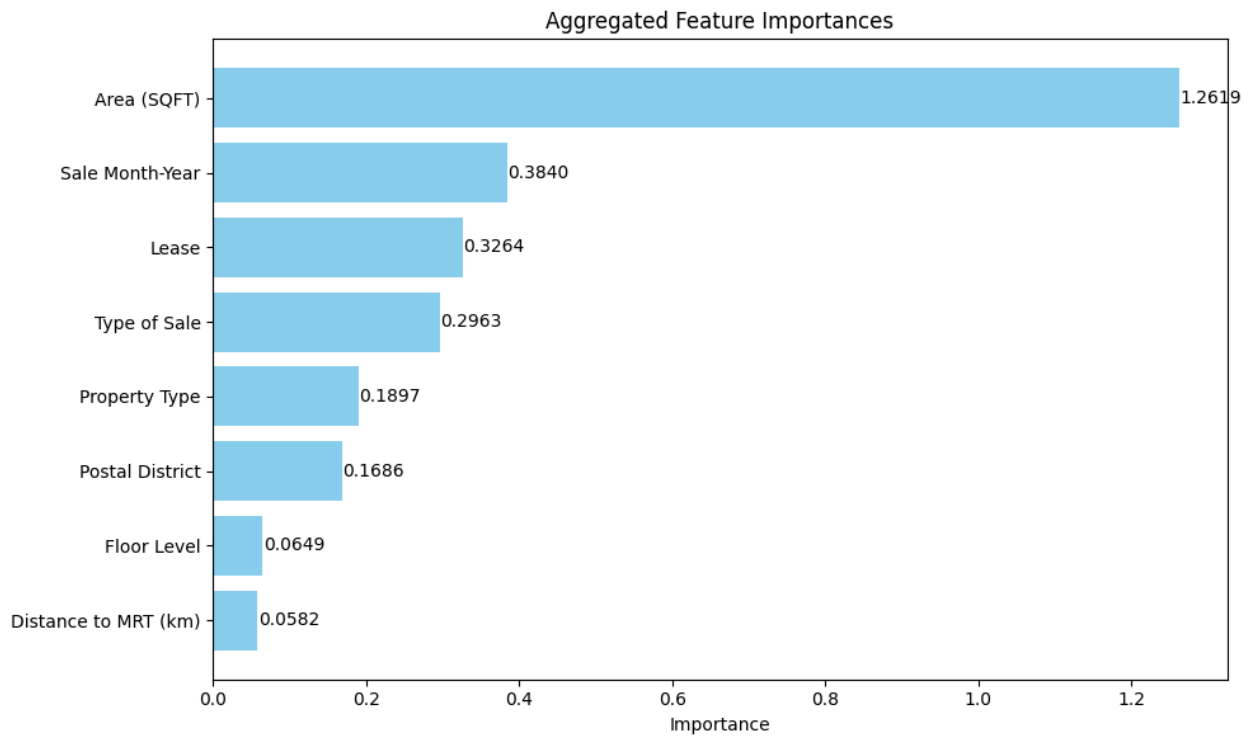
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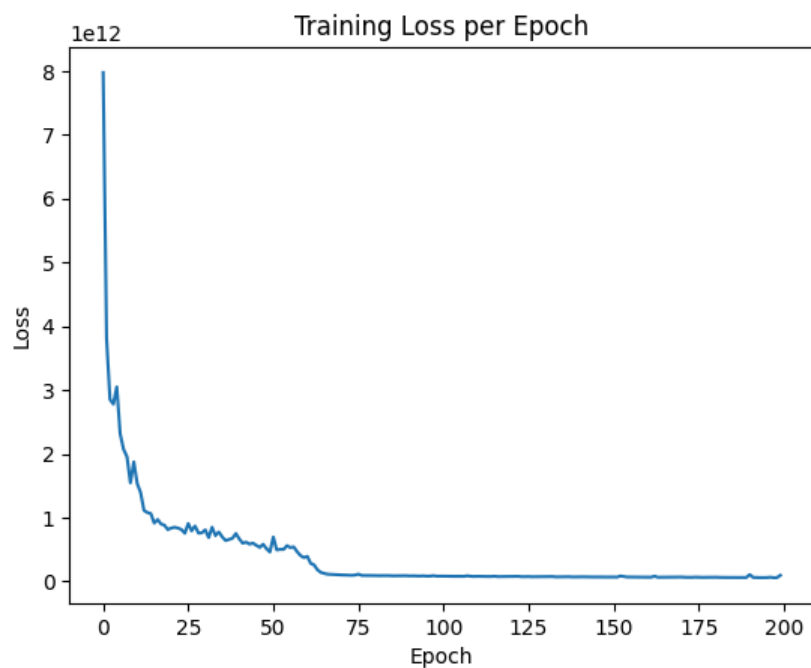
Feature importances (descending):

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

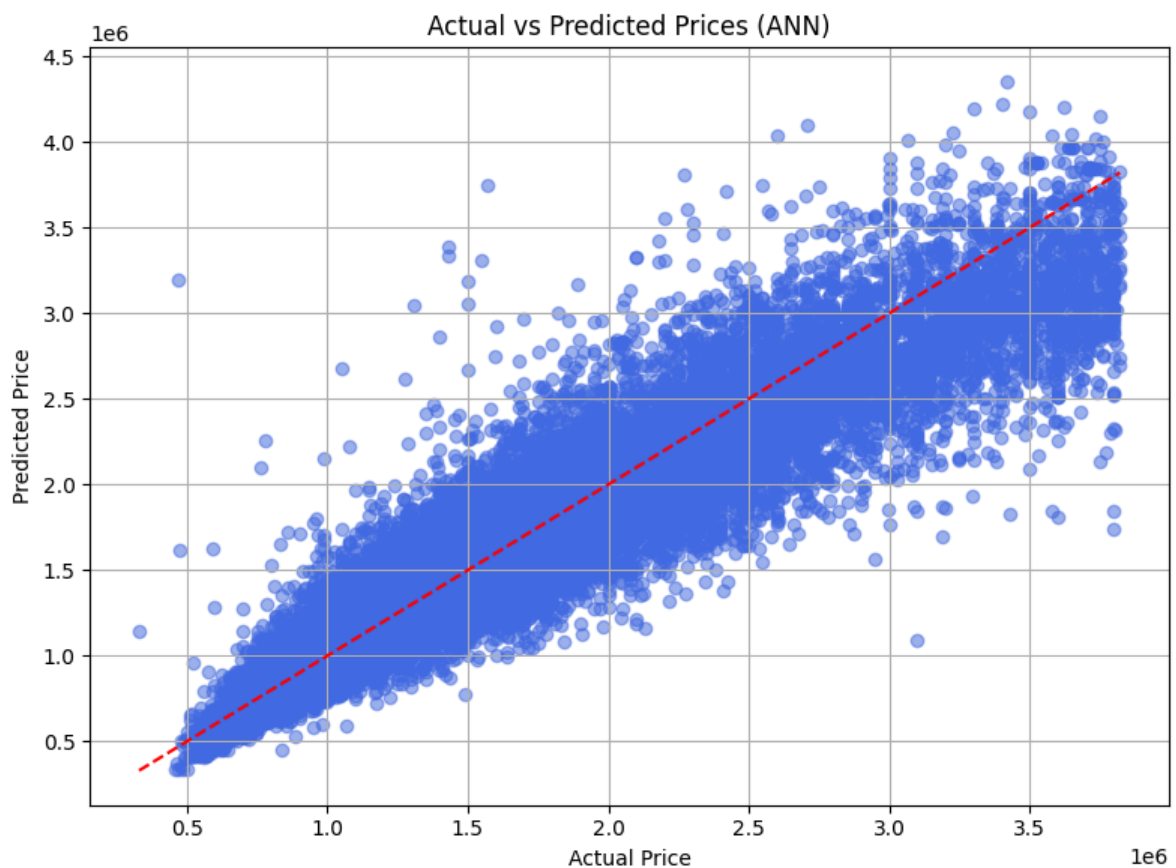
```
Out[17]:
```

	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()
```

925/925 [=====] - 1s 610us/step



```
In [18]: #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 []: