

# trabajo-final

May 23, 2025

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score

from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
```

```
[27]: from google.colab import files
uploaded = files.upload()
```

<IPython.core.display.HTML object>

Saving real\_estate\_clean.csv to real\_estate\_clean (1).csv

```
[28]: # Cargar datos
df = pd.read_csv("real_estate_clean.csv")

# Mostrar las primeras filas
df.head()
```

```
[28]:
```

	No	X1 transaction date	X2 house age	\
0	1	2012.917	32.0	
1	2	2012.917	19.5	
2	3	2013.583	13.3	
3	4	2013.500	13.3	
4	5	2012.833	5.0	

  

	X3 distance to the nearest MRT station	X4 number of convenience stores	\
0	84.87882	10	
1	306.59470	9	
2	561.98450	5	
3	561.98450	5	
4	390.56840	5	

	X5 latitude	X6 longitude	Y house price of unit area
0	24.98298	121.54024	37.9
1	24.98034	121.53951	42.2
2	24.98746	121.54391	47.3
3	24.98746	121.54391	54.8
4	24.97937	121.54245	43.1

```
[29]: df.columns
```

```
[29]: Index(['No', 'X1 transaction date', 'X2 house age',
        'X3 distance to the nearest MRT station',
        'X4 number of convenience stores', 'X5 latitude', 'X6 longitude',
        'Y house price of unit area'],
        dtype='object')
```

```
[30]: # Renombrar columnas para simplificar
df.columns = ['No', 'TransactionDate', 'HouseAge', 'DistanceToMRT',
             ↪ 'ConvenienceStores', 'Latitude', 'Longitude', 'PricePerUnit']

# Eliminar columna "No" que no aporta
df.drop(columns=['No'], inplace=True)
```

```
[31]: X = df[['TransactionDate', 'HouseAge', 'DistanceToMRT', 'ConvenienceStores',
             ↪ 'Latitude', 'Longitude']]
y = df['PricePerUnit']

# Dividir datos en entrenamiento y prueba
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
             ↪ random_state=42)
```

## Modelo de Regresión Lineal Múltiple

```
[32]: # Crear modelo
lr_model = LinearRegression()

# Entrenar modelo
lr_model.fit(X_train, y_train)

# Predicciones
y_pred_lr = lr_model.predict(X_test)

# Evaluación
mse_lr = mean_squared_error(y_test, y_pred_lr)
r2_lr = r2_score(y_test, y_pred_lr)

print(" Regresión Lineal")
```

```
print("MSE:", mse_lr)
print("R²:", r2_lr)
```

Regresión Lineal  
MSE: 60.97637965746742  
R²: 0.6309164822754011

### Modelo Red Neuronal(RNA)

```
[9]: # Crear modelo secuencial
nn_model = Sequential()
nn_model.add(Dense(64, activation='relu', input_shape=(X_train.shape[1],)))
nn_model.add(Dense(32, activation='relu'))
nn_model.add(Dense(1)) # Capa de salida

# Compilar modelo
nn_model.compile(optimizer='adam', loss='mean_squared_error')

# Entrenar modelo
history = nn_model.fit(X_train, y_train, epochs=100, validation_split=0.2,
    verbose=0)

# Predicciones
y_pred_nn = nn_model.predict(X_test).flatten()

# Evaluación
mse_nn = mean_squared_error(y_test, y_pred_nn)
r2_nn = r2_score(y_test, y_pred_nn)

print(" Red Neuronal Artificial")
print("MSE:", mse_nn)
print("R²:", r2_nn)
```

/usr/local/lib/python3.11/dist-packages/keras/src/layers/core/dense.py:87:  
UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. When  
using Sequential models, prefer using an `Input(shape)` object as the first  
layer in the model instead.

```
super().__init__(activity_regularizer=activity_regularizer, **kwargs)

1/1          0s 69ms/step
Red Neuronal Artificial
MSE: 30.181989968604327
R²: 0.05430079998106485
```

### predicciones nuevas

```
[33]: # Tomamos 5 muestras aleatorias del conjunto de prueba
X_new = X_test.sample(5, random_state=1)
```

```

# Predicciones con ambos modelos
preds_lr = lr_model.predict(X_new)
preds_nn = nn_model.predict(X_new).flatten()

# Mostrar resultados
pred_df = pd.DataFrame({
    'Real': y_test.loc[X_new.index],
    'Predicción_LR': preds_lr,
    'Predicción_RNA': preds_nn
})
print(pred_df)

```

```

1/1          0s 87ms/step
   Real  Predicción_LR  Predicción_RNA
118  48.6      45.724767      35.138149
33   22.9      30.325355      31.107468
332  30.8      33.661584      35.358959
140  45.1      14.986237       1.767540
94   38.2      44.327459      76.860878

```

### Comparar modelos

```

[34]: print(" Comparación de Modelos")
      print(f"Regresión Lineal - MSE: {mse_lr:.2f}, R²: {r2_lr:.2f}")
      print(f"Red Neuronal - MSE: {mse_nn:.2f}, R²: {r2_nn:.2f}")

```

```

Comparación de Modelos
Regresión Lineal - MSE: 60.98, R²: 0.63
Red Neuronal - MSE: 30.18, R²: 0.05

```

```

[35]: plt.plot(history.history['loss'], label='Entrenamiento')
      plt.plot(history.history['val_loss'], label='Validación')
      plt.title('Pérdida durante el entrenamiento')
      plt.xlabel('Época')
      plt.ylabel('MSE')
      plt.legend()
      plt.show()

```

