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# Lab 4: Linear Regression
# 0. Install and Import Libraries

# Install ucimlrepo if not done
!pip install ucimlrepo

from ucimlrepo import fetch_ucirepo
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
Collecting ucimlrepo
  Downloading ucimlrepo-0.0.7-py3-none-any.whl.metadata (5.5 kB)
Requirement already satisfied: pandas>=1.0.0 in /usr/local/lib/python3.12/dist-packages (from ucimlrepo) (2.2.2)
Requirement already satisfied: certifi>=2020.12.5 in /usr/local/lib/python3.12/dist-packages (from ucimlrepo) (2025.11.11)
Requirement already satisfied: numpy>=1.26.0 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2.0.2)
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2.9.0)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.12/dist-packages (from pandas>=1.0.0->ucimlrepo) (2025.1)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.12/dist-packages (from python-dateutil>=2.8.2->pandas) (1.17.0)
Downloading ucimlrepo-0.0.7-py3-none-any.whl (8.0 kB)
Installing collected packages: ucimlrepo
Successfully installed ucimlrepo-0.0.7
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# Lab 4: Linear Regression
# 1. Load Dataset and Inspect

from ucimlrepo import fetch_ucirepo

chess_king_rook_vs_king_pawn = fetch_ucirepo(id=22)

# Features and target extraction
X = chess_king_rook_vs_king_pawn.data.features
y = chess_king_rook_vs_king_pawn.data.targets

# Metadata and variable information
print("Metadata:\n", chess_king_rook_vs_king_pawn.metadata)
print("\nVariables:\n", chess_king_rook_vs_king_pawn.variables)

# Display sample features and targets
print("Sample features:\n", X.head())
print("Sample targets:\n", y.head())
```

| | | | | | | | |
|----|-------|---------|-------------|------|------|------|----|
| 2 | bkon8 | Feature | Categorical | None | None | None | no |
| 3 | bkona | Feature | Categorical | None | None | None | no |
| 4 | bkspr | Feature | Categorical | None | None | None | no |
| 5 | bkxbq | Feature | Categorical | None | None | None | no |
| 6 | bkxcr | Feature | Categorical | None | None | None | no |
| 7 | bkxwp | Feature | Categorical | None | None | None | no |
| 8 | blxwp | Feature | Categorical | None | None | None | no |
| 9 | bxqsq | Feature | Categorical | None | None | None | no |
| 10 | cntxt | Feature | Categorical | None | None | None | no |
| 11 | dsopp | Feature | Categorical | None | None | None | no |
| 12 | dwipd | Feature | Categorical | None | None | None | no |
| 13 | hdchk | Feature | Categorical | None | None | None | no |
| 14 | katri | Feature | Categorical | None | None | None | no |
| 15 | mulch | Feature | Categorical | None | None | None | no |
| 16 | qxmsq | Feature | Categorical | None | None | None | no |
| 17 | r2ar8 | Feature | Categorical | None | None | None | no |
| 18 | reskd | Feature | Categorical | None | None | None | no |
| 19 | reskr | Feature | Categorical | None | None | None | no |
| 20 | rimmx | Feature | Categorical | None | None | None | no |
| 21 | rkxwp | Feature | Categorical | None | None | None | no |
| 22 | rxmsq | Feature | Categorical | None | None | None | no |
| 23 | simpl | Feature | Categorical | None | None | None | no |
| 24 | skach | Feature | Categorical | None | None | None | no |
| 25 | skewr | Feature | Categorical | None | None | None | no |
| 26 | skrxp | Feature | Categorical | None | None | None | no |
| 27 | spcop | Feature | Categorical | None | None | None | no |
| 28 | stlmt | Feature | Categorical | None | None | None | no |
| 29 | thrsk | Feature | Categorical | None | None | None | no |
| 30 | wkcti | Feature | Categorical | None | None | None | no |
| 31 | wkna8 | Feature | Categorical | None | None | None | no |
| 32 | wknck | Feature | Categorical | None | None | None | no |

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f      f      f      f      t      f      t      f      f      f      f      ...      f
f      f      f      f      f      f      f      f      t      f      f      ...      f
f      f      f      f      f      f      f      f      f      f      f      ...      f

      skrxp  spcop  stlmt  thrsk  wkcti  wkna8  wknck  wkovl  wkpos
f      f      f      f      f      f      f      t      t      n
f      f      f      f      f      f      f      t      t      n
f      f      f      f      f      f      f      t      t      n
f      f      f      f      f      f      f      t      t      n
f      f      f      f      f      f      f      t      t      n

[5 rows x 35 columns]
Sample targets:
      wtoeg
f      won
f      won
f      won
f      won
f      won

```

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# Lab 4: Linear Regression
# 2. Train-Test Split

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# For simple regression, select one independent variable (e.g., first column)
X_single = X.iloc[:, [0]] # Replace 0 with appropriate feature index for regression relevance

# Apply one-hot encoding to the selected feature
X_single_encoded = pd.get_dummies(X_single, drop_first=True)

# Encode the target variable
y_encoded = y['wtoeg'].apply(lambda x: 1 if x == 'won' else 0)

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X_train, X_test, y_train, y_test = train_test_split(X_single_encoded, y_encoded, test_size=0.2, random_state=42)

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# Lab 4: Linear Regression
# 3. Model Fitting

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reg = LinearRegression()
reg.fit(X_train, y_train)

print("Intercept (B0):", reg.intercept_)
print("Slope (B1):", reg.coef_)

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Intercept (B0): 0.5200506542845083
Slope (B1): [-0.00133408]

```

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# Lab 4: Linear Regression
# 4. Prediction, Visualization, and Residuals

```

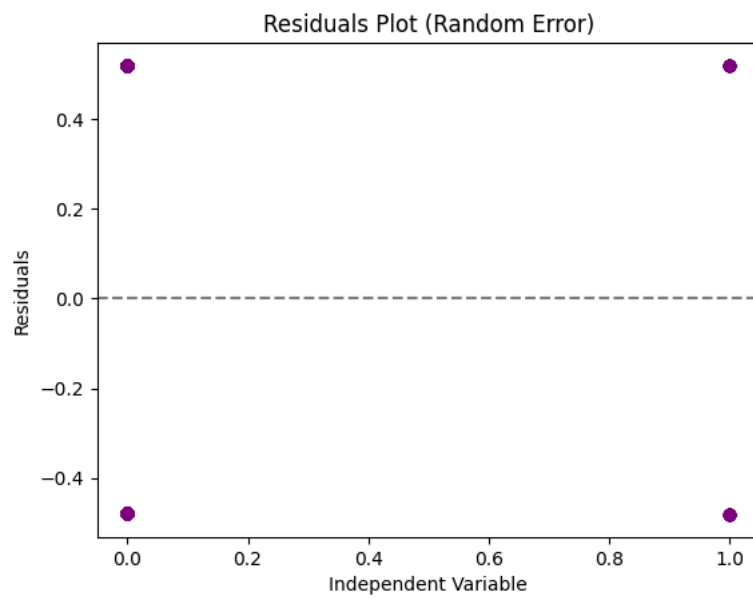
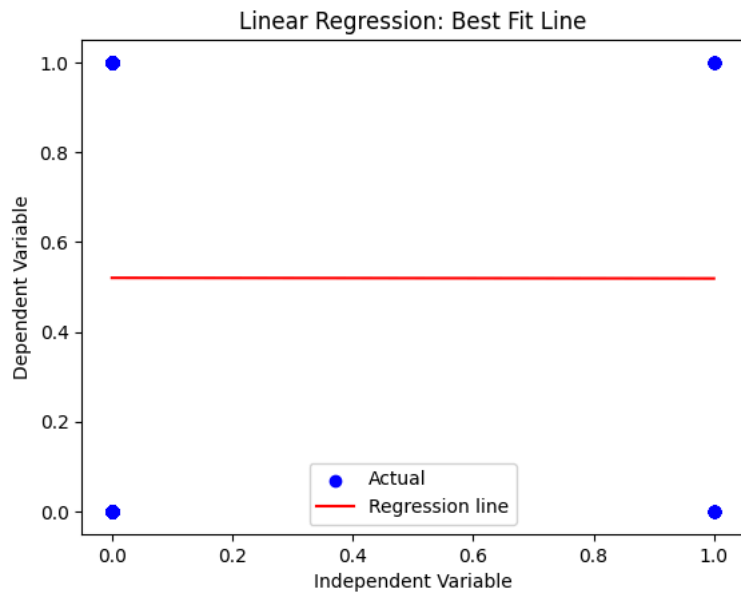
```

# Predict on train and test set
y_pred = reg.predict(X_test)

# Scatter plot predicted vs actual
plt.scatter(X_test, y_test, color='blue', label='Actual')
plt.plot(X_test, y_pred, color='red', label='Regression line')
plt.xlabel('Independent Variable')
plt.ylabel('Dependent Variable')
plt.title('Linear Regression: Best Fit Line')
plt.legend()
plt.show()

# Residuals
residuals = y_pred - y_test
plt.scatter(X_test, residuals, color='purple')
plt.axhline(y=0, linestyle='--', color='gray')
plt.xlabel('Independent Variable')
plt.ylabel('Residuals')
plt.title('Residuals Plot (Random Error)')
plt.show()

```



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