

Student Performance Prediction

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Introduction

Predicting how students will do in school is becoming a big deal in education. By looking at things like how often they show up to class, their past grades, how they study, and even how they interact with others, schools can spot patterns that help them for future performance. This kind of analysis gives teachers, schools, and parents useful information to provide the right kind of guidance for their future . It makes education more easier and efficient so every student has a better chance to reach their desire goal by using their full potential .

Methodology

The process of predicting student performance involves a few key steps, explained simply:

- 1. **Collect Data:** Taking information about students, like attendance, past grades, study habits, and even how they interact with other student ,teachers and parents .*
- 2. **Analyze Patterns:** By data analysis we identify the pattern od attendance ,marks and which student is continuously growing and which does not*
- 3. **Build a Model:** Create a system that can learn from the data and make predictions. it might predict if a student is likely to struggle in a subject based on their past performance and results.*
- 4. **Test and Improve:** Check the predictions are accurate by comparing the model with real student data .And improve the model as needed.*
- 5. **Take Action:** Use the predictions to help students. For example, if the student need help so by our system we can analyse on time and give the required guidance .*

Code

```
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
import matplotlib.pyplot as plt

# Sample dataset (you can modify or load your own)
data = {
    'Study_Hours': [2, 3, 4, 5, 6, 7, 8, 9, 10, 11],
    'Attendance': [60, 65, 70, 75, 80, 85, 90, 92, 95, 98],
    'Previous_Score': [50, 55, 58, 60, 65, 68, 72, 75, 78, 80],
    'Final_Score': [55, 60, 62, 66, 70, 73, 76, 78, 81, 85]
}
```

Convert to DataFrame

```
df = pd.DataFrame(data)
```

Split features and target

```
X = df[['Study_Hours', 'Attendance', 'Previous_Score']]
```

```
y = df['Final_Score']
```

Split into training and test sets (80% train, 20% test)

```
X_train, X_test, y_train, y_test = train_test_split(X, y,  
test_size=0.2, random_state=42)
```

Scale the features (important for neural networks)

```
scaler = StandardScaler()
```

```
X_train_scaled = scaler.fit_transform(X_train)
```

```
X_test_scaled = scaler.transform(X_test)
```

Build the Neural Network Model

```
model = Sequential()
```

```
model.add(Dense(16, input_dim=X_train.shape[1],  
activation='relu'))
```

```
model.add(Dense(8, activation='relu'))
```

```
model.add(Dense(1)) # Output layer for regression
```

Compile the model

```
model.compile(optimizer='adam', loss='mse', metrics=['mae'])
```

```
# Train the model
```

```
history = model.fit(X_train_scaled, y_train, epochs=100,  
batch_size=4, validation_data=(X_test_scaled, y_test),  
verbose=0)
```

```
# Evaluate the model
```

```
loss, mae = model.evaluate(X_test_scaled, y_test, verbose=0)
```

```
print(f'Mean Absolute Error: {mae:.2f}')
```

```
# Predict on test data
```

```
predictions = model.predict(X_test_scaled)
```

```
print("\nPredictions vs Actual Values:")
```

```
print(pd.DataFrame({'Predicted': predictions.flatten(), 'Actual':  
y_test.values}))
```

```
# Plot training history
```

```
plt.plot(history.history['mae'], label='Training MAE')
```

```
plt.plot(history.history['val_mae'], label='Validation MAE')
```

```
plt.title('Model Performance')
```

```
plt.xlabel('Epochs')
```

```
plt.ylabel('Mean Absolute Error')
```

```
plt.legend()
```

```
plt.show()
```

Screenshot of output

