

Introduction to Machine Learning

Week 4 Assignment

Part 1:

1. Identify which type of machine learning (supervised, unsupervised, or re-inforcement learning) would be most appropriate for developing the smart thermostat.

Answer: Re-inforcement learning

2. Explain why this type of learning is suitable for this application

Re-inforcement learning is suitable because the smart thermostat algorithm will need to learn to make sequential decisions by interacting with the environment and receiving feedback.

Part 2

1.

Student Number	Actual Response	Predicted Value
1	1	1
2	0	0
3	1	1
4	1	0
5	0	0
6	1	1
7	0	1
8	0	0
9	1	1
10	0	0
11	1	0
12	1	1
13	0	0
14	0	1
15	1	1

- True Positive (TP) occurs when Actual = 1 and Predicted = 1
- True Negative (TN) occurs when Actual = 0 and Predicted = 0
- False Positive (FP) occurs when Actual = 0 and Predicted = 1
- False Negative (FN) occurs when Actual = 1 and Predicted = 0

	Positive (Predicted)	Negative(Predicted)
Positive (Actual)	6 (TP)	2 (FN)
Negative (Actual)	2(FP)	5 (TN)

2.

$$1. \text{ Accuracy} = \frac{\text{True Positives} + \text{True Negatives}}{\text{Total}}$$

$$\text{Accuracy} = \frac{6 + 5}{15} = \frac{11}{15} = \mathbf{0.73 (73\%)}$$

$$2. \text{ Precision} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Positive}}$$

$$\text{Precision} = \frac{6}{6+2} = \frac{6}{8} = \mathbf{0.75 (75\%)}$$

$$3. \text{ Recall} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Negative}}$$

$$\text{Recall} = \frac{6}{6+2} = \frac{6}{8} = \mathbf{0.75 (75\%)}$$

$$4. \text{ F1 Score} = 2 \cdot \frac{\text{Precision} \cdot \text{Recall}}{\text{Precision} + \text{Recall}}$$

$$\text{F1 Score} = 2 \cdot \frac{0.75 \cdot 0.75}{0.75 + 0.75} = 2 \cdot \frac{0.5625}{1.5} = 2 \cdot 0.375 = \mathbf{0.75 (75\%)}$$

3. Yes. It is a fairly good model because:

1. In term of Accuracy the model correctly predicts about 73% of the instances.
2. In term of Precision the model correctly identified the positive class 75% of the time.
3. In term of Recall the model correctly identifies 75% of all actual "Yes" instances.
4. In term of F1 Score there is a balanced performance between precision and recall.

4. To improve the model, we can:

1. Increase Data Quality and Quantity
2. Cross-Validation
3. Hyperparameter Tuning
4. Handle Class Imbalance
5. Use Ensemble Methods

```
#IMPORT LIBRARIES
import numpy as np
import matplotlib.pyplot as plt
```

You are given a list of inputs x and outputs y and a linear regression model. The aim of this exercise is to validate the given model by calculating the mean square error (MSE) and the coefficient of determination (R-squared).

```
x=[-2.0,-1.6,-1.1,-0.67,-0.22,0.22,0.67,1.11,1.6,2.0]
y=[-0.46,1.3,0.27,1.4,2.6,2.9,3.1,2.3,3.8,3.6]
```

1. What is the design matrix X ?

```
# CONSTRUCT THE DESIGN MATRIX
X= np.vstack((np.ones(len(x)), x)).T
```

2. The linear regression model is given by $y_{pre} = \theta_0 + x\theta_1$, in matrix form we have $y_{pre} = X * \theta$ where $\theta_0 = 2.08$ and $\theta_1 = 0.94$.
Compute the predicted output y_{pre}

```
# CREATE A LIST FOR YOUR PARAMETERS
Theta = np.array([2.08, 0.94])

#COMPUTE THE PREDICTED OUTPUT
y_pre= X @ Theta
```

3. Compute the MSE and the R-squared

$$MSE = \frac{1}{n} \sum_{i=0}^n (y_i - y_{pred,i})^2$$

$$R^2 = 1 - \frac{SSR}{TSS}$$

$$SSR = \sum_{i=0}^n (y_i - y_{pred,i})^2$$

$$TSS = \sum_{i=0}^n (y_i - \bar{y}_i)^2$$

```
MSE = np.mean((y - y_pre) ** 2)
print("Mean Squared Error:",MSE)
SSR = np.sum((y - y_pre) ** 2)
y_mean = np.mean(y)
TSS = np.sum((y - y_mean) ** 2)
r2 = 1 - (SSR / TSS)
print("R-squared:",r2)
```

➞ Mean Squared Error: 0.34752241200000006
R-squared: 0.8079868920138196

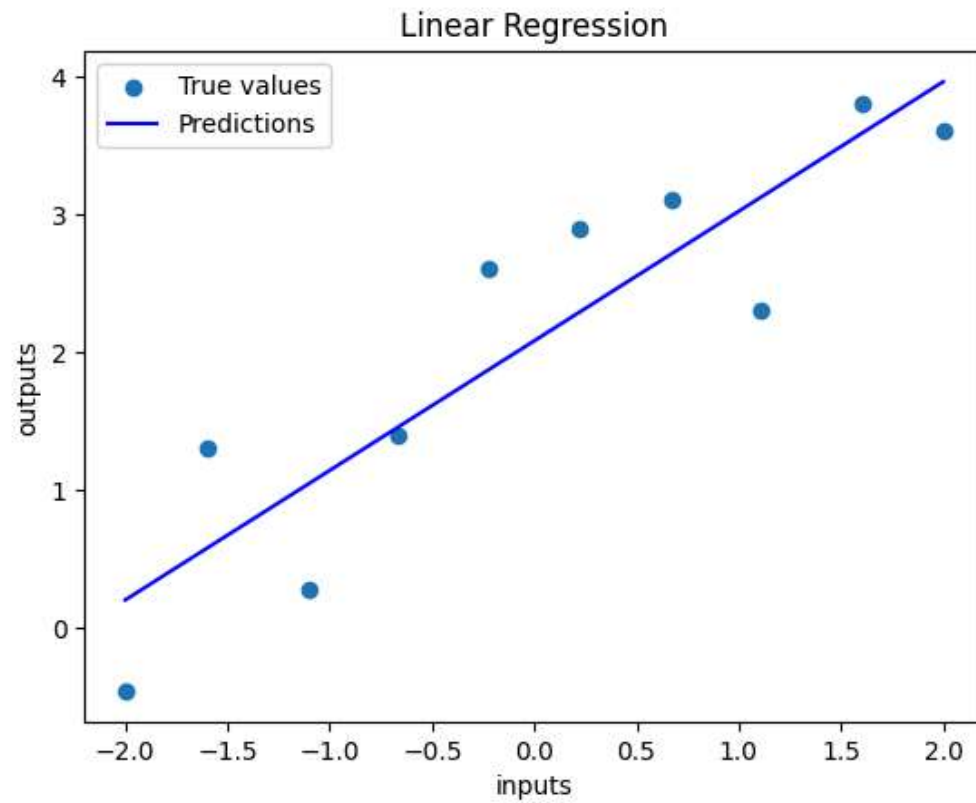
4. Is this model good?

Yes

5. Plot the true outputs and the predicted against the inputs on the same figure

```
#=====
#Visualize the model
#=====

plt.figure(1)
plt.scatter(x, y,label='True values')
plt.plot(x,y_pre, 'b',label='Predictions')
plt.xlabel('inputs')
plt.ylabel('outputs')
plt.title('Linear Regression')
plt.legend()
plt.show()
```



Start coding or [generate](#) with AI.

