



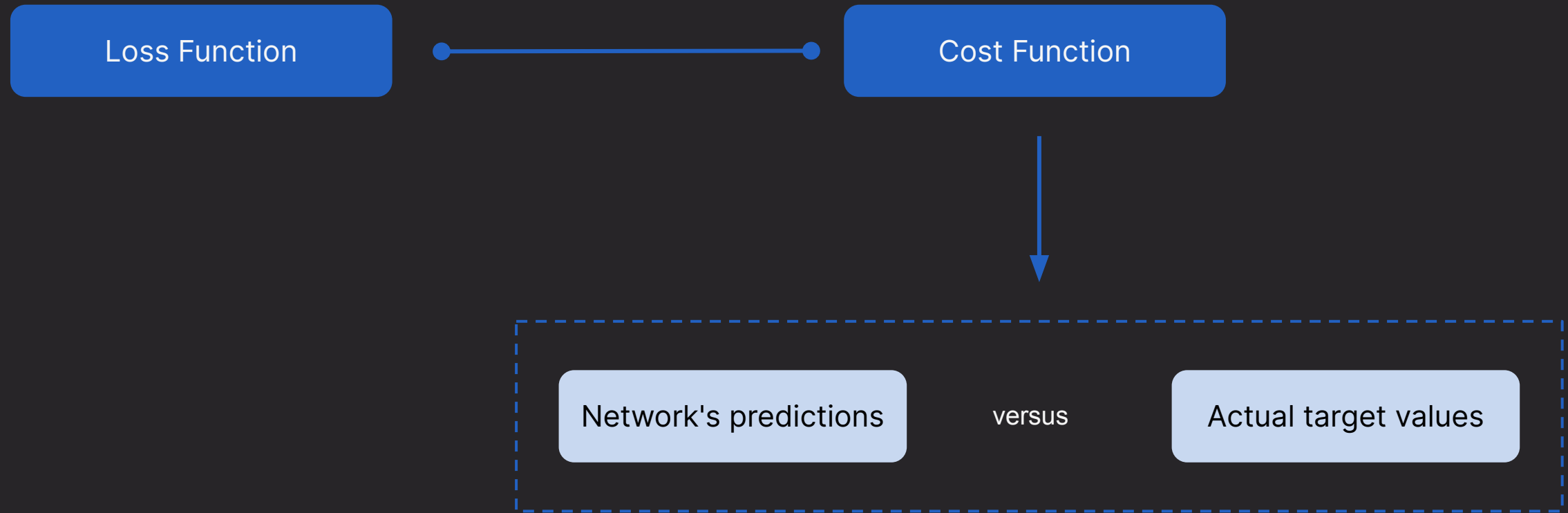
Understanding workings of Neural Networks

Video 2: Understanding Loss Functions

In Air

Loss Function

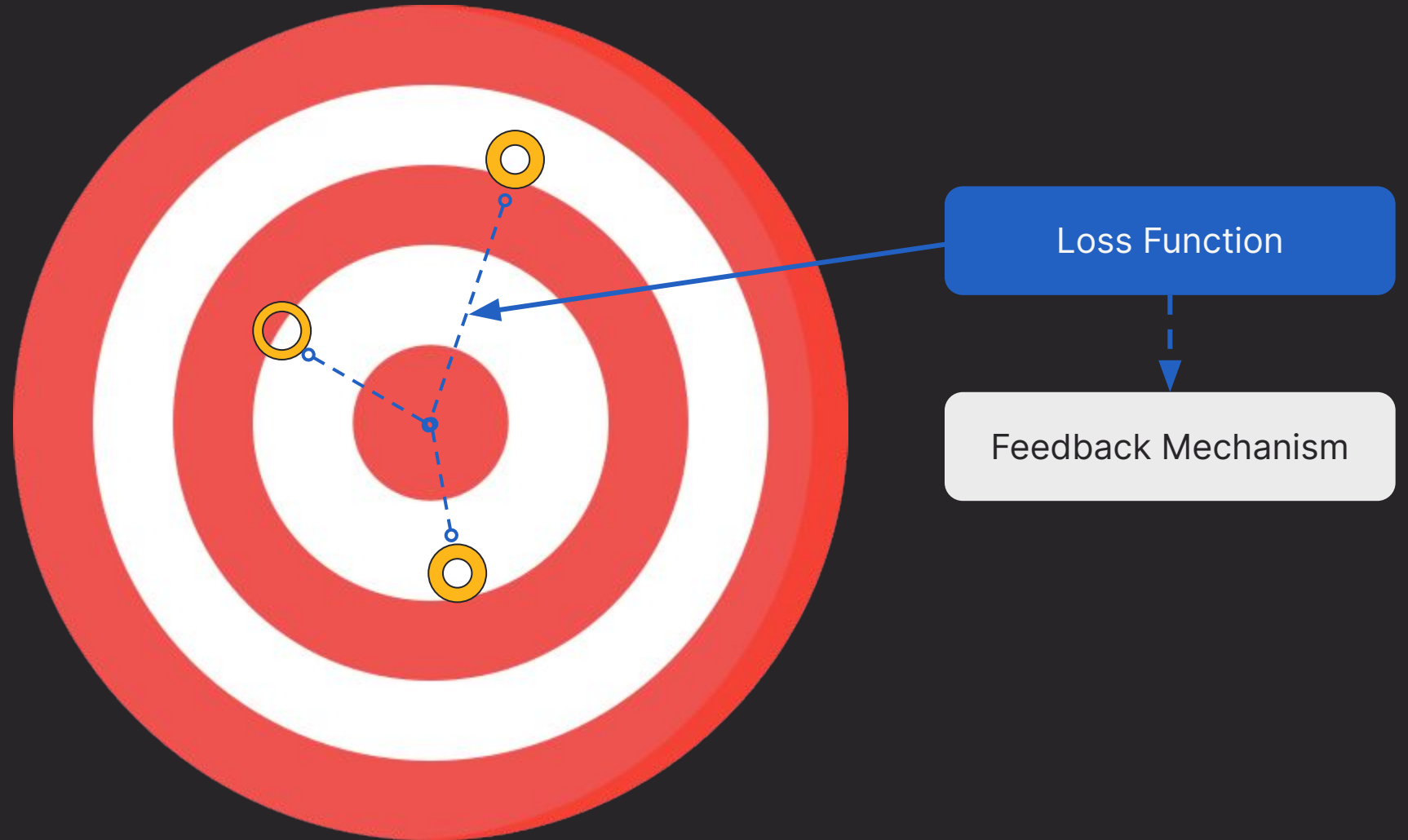
Method to measure the error of network's predictions



Loss Function

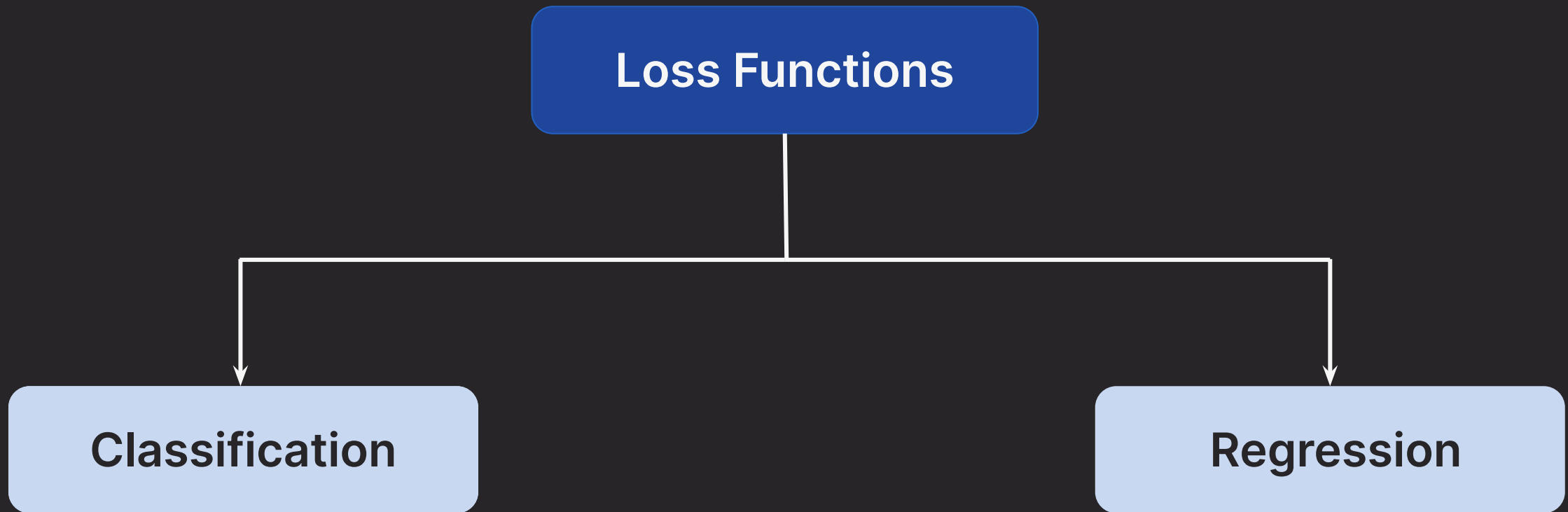


Loss Function



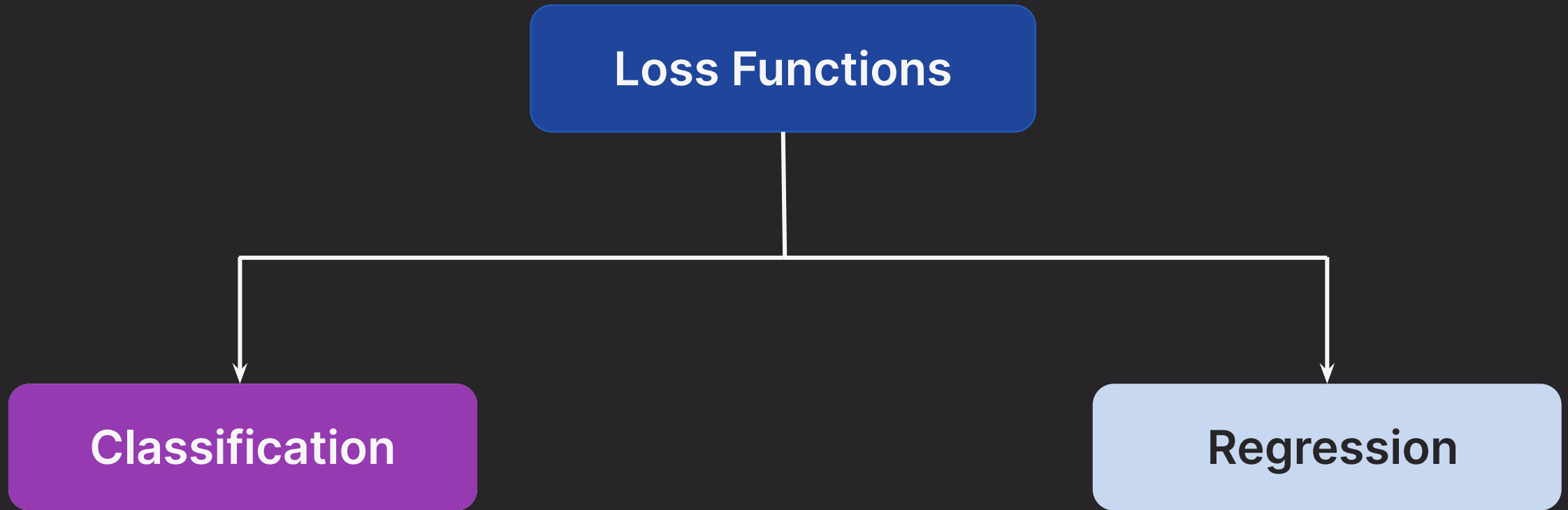
Loss Function

Categorized based on the machine learning tasks



Loss Function

Categorized based on the machine learning tasks



Binary Cross-Entropy Loss or Log Loss

Binary Cross Entropy / Log Loss

Evaluates the accuracy of binary classification predictions against true outcomes

Difference Calculation:

y_i (actual targets) - p_i (predicted probabilities)

Logarithmic Transformation:

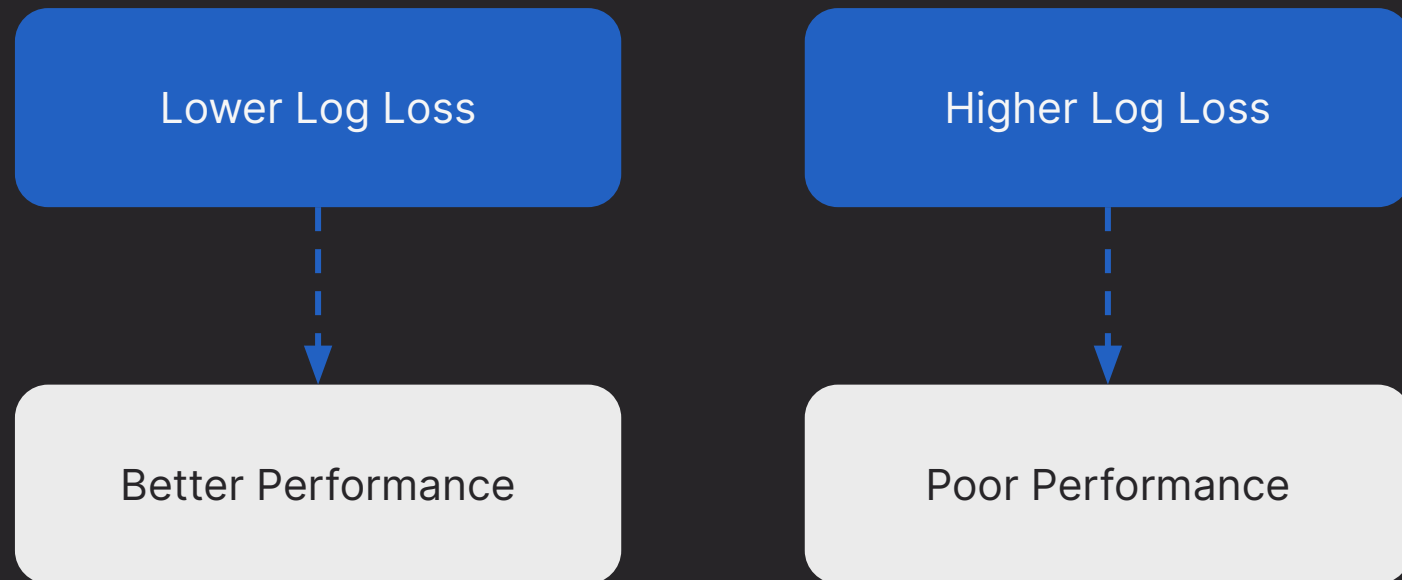
$-\log(p_i)$ for positive class and $-\log(1 - p_i)$ for negative class

Average Loss:

$$L(y, f(x)) = -\sum [y_i * \log(p_i) + (1 - y_i) * \log(1 - (p_i))]$$

 **Minimize the loss value during training**

Binary Cross Entropy / Log Loss



Category Cross Entropy Loss

A variation of BCE loss applicable for multi-classification problems.

$$\text{CCE} = -\sum y_i \text{Log}(p)$$

Category Cross Entropy Loss: $-\sum y_i \text{Log}(p)$

Consider a model to guess the type of vehicle in an image from 5 options.



Two-wheeler



Autorickshaw



Car



Bus



Truck

Category Cross Entropy Loss: $-\sum y_i \text{Log} (p)$

Consider a model to guess the type of vehicle in an image from 5 options.



Two-wheeler



Autorickshaw



Car



Bus



Truck



Given Input

Category Cross Entropy Loss: $-\sum y_i \text{Log}(p)$

Consider a model to guess the type of vehicle in an image from 5 options.



Two-wheeler

Probability: 0.05



Autorickshaw

Probability: 0.02



Car

Probability: 0.05



Bus

Probability: 0.85



Truck

Probability: 0.03



Given Input

Category Cross Entropy Loss: $-\sum y_i \text{Log} (p)$

Consider a model to guess the type of vehicle in an image from 5 options.



Two-wheeler

Probability: 0.05



Autorickshaw

Probability: 0.02



Car

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Truck

Probability: 0.03

Target (y_i) = [0 0 0 1 0]

Category Cross Entropy Loss: $-\sum y_i \text{Log} (p)$

Consider a model to guess the type of vehicle in an image from 5 options.



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Probability: 0.03

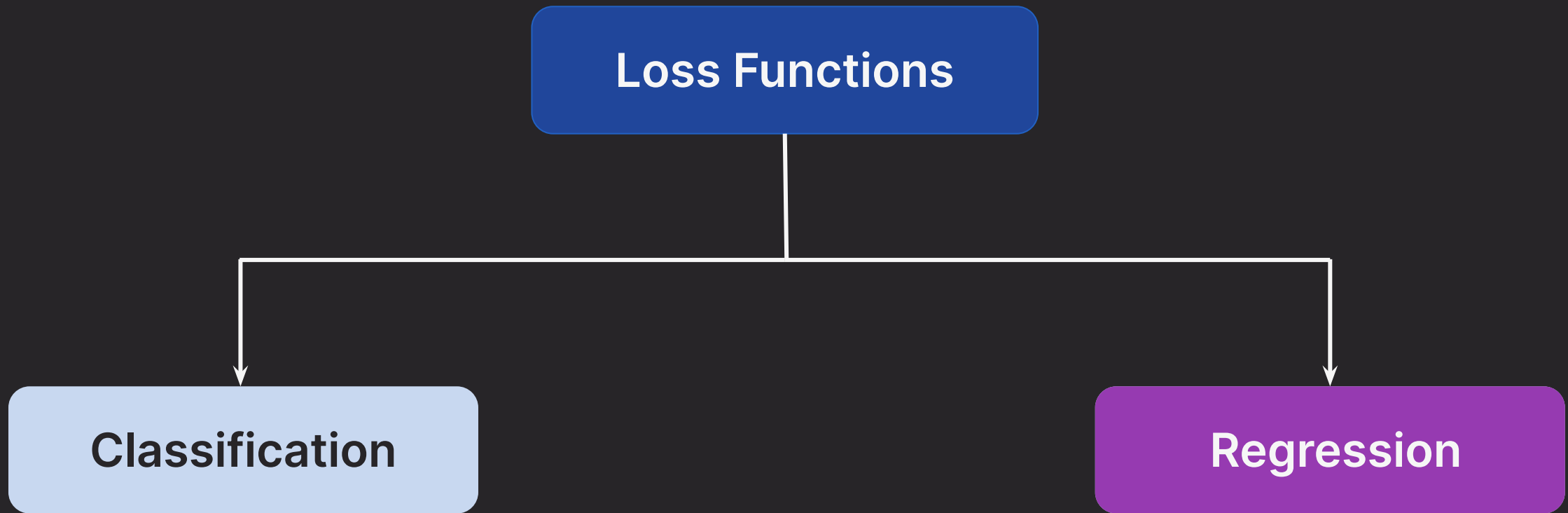
Target (y_i) = [0 0 0 1 0]

$$\text{CCE}_{\text{others}} = 0 * \text{Log} (p)$$

$$\text{CCE}_{\text{bus}} = - 1 * \text{Log} (0.85) = 0.07$$

Loss Function

Categorized based on the machine learning tasks



Loss functions for Regression Problems

Mean Square Error (MSE) measures the average squared difference between predictions and actual values.

$$\text{MSE} = (1/n) \sum (\hat{y} - y_i)^2$$

Number of samples in the dataset

Predicted value for the i^{th} sample

Target value for the i^{th} sample

Loss functions for Regression Problems



MSE not the best choice if data contains significant outliers





Mean Absolute Error

Loss functions for Regression Problems

Mean Absolute Error (MAE) averages the absolute differences between predicted and actual values.

$$\text{MAE} = (1/n) \sum |\hat{y} - y_i|$$

Diagram illustrating the components of the MAE formula:

- \hat{y} : Predicted value for the i^{th} sample
- n : Number of samples in the dataset
- y_i : Target value for the i^{th} sample

MAE: Preferred Scenarios

Predicting delivery times



Conclusion



- Binary Cross-Entropy Loss
- Mean Square Error (MSE)
- Mean Absolute Error (MAE)



Most commonly used loss functions

Hands-on