# **Regression Loss Functions:**

#### 1. Mean Squared Error (MSE):

- $\circ$  Formula:  $(\frac{1}{N}\sum (y_{\mathrm{true}} y_{\mathrm{pred}})^2)$
- o Example:

```
loss = keras.losses.MeanSquaredError()
```

• Use Case: Useful for regression problems where the goal is to minimize the difference between predicted and true values.

## 2. Mean Absolute Error (MAE):

- $\circ$  Formula:  $(rac{1}{N}\sum |y_{
  m true}-y_{
  m pred}|)$
- Example:

```
loss = keras.losses.MeanAbsoluteError()
```

• Use Case: Useful for regression problems, especially when the dataset contains outliers.

#### **Classification Loss Functions:**

## 1. Binary Crossentropy:

- $\circ$  Formula:  $(-\sum (y_{ ext{true}}\log(y_{ ext{pred}}) + (1-y_{ ext{true}})\log(1-y_{ ext{pred}})))$
- Example:

• Use Case: Ideal for binary classification problems.

# 2. Categorical Crossentropy:

- $\circ$  Formula:  $(-\sum y_{ ext{true}} \log(y_{ ext{pred}}))$
- o Example:

```
loss = keras.losses.CategoricalCrossentropy()
```

o Use Case: Best suited for multi-class classification problems where the labels are one-hot encoded.

# 3. Sparse Categorical Crossentropy:

o Example:

```
loss = keras.losses.SparseCategoricalCrossentropy()
```

• Use Case: Suitable for multi-class classification problems where the labels are integers instead of one-hot encoded.

## 4. Hinge Loss:

- $\circ$  Formula:  $(\max(0, 1 y_{\text{true}} \cdot y_{\text{pred}}))$
- o Example:

• Use Case: Generally used in binary classification problems and often used in SVMs.

# **Clustering Loss Functions:**

#### 1. Kullback-Leibler Divergence (KL Divergence):

- $\circ$  Formula:  $(\sum y_{ ext{true}} \log(rac{y_{ ext{true}}}{y_{ ext{pred}}}))$
- o Example:

• Use Case: Useful when comparing two probability distributions, often used in tasks like clustering or in models like Variational Autoencoders (VAEs).

# **Summary Table**

| Loss<br>Function                           | Formula   | Example  | Use Case   | Commonly<br>Used Models                                      |
|--|---|--|--|--|
| Mean<br>Squared<br>Error (MSE)             | $(rac{1}{N}\sum (y_{ m true}-y_{ m pred})^2)$  | loss = keras.losses.MeanSquaredError()                         | Regression<br>problems   | Linear<br>Regression,<br>Neural<br>Networks                  |
| Mean<br>Absolute<br>Error (MAE)            | $(rac{1}{N}\sum  y_{ m true} - y_{ m pred} )$  | <pre>loss = keras.losses.MeanAbsoluteError()</pre>             | Regression<br>with outliers  | Linear<br>Regression,<br>Neural<br>Networks                  |
| Binary<br>Crossentropy                     | $egin{aligned} (-\sum(y_{	ext{true}}\log(y_{	ext{pred}}) + \ (1-y_{	ext{true}})\log(1-y_{	ext{pred}})) \end{aligned}$ | <pre>loss = keras.losses.BinaryCrossentropy()</pre>            | Binary<br>classification   | Logistic<br>Regression,<br>Neural<br>Networks                |
| Categorical<br>Crossentropy                | $(-\sum y_{	ext{true}} \log(y_{	ext{pred}}))$   | <pre>loss = keras.losses.CategoricalCrossentropy()</pre>       | Multi-class<br>classification<br>with one-<br>hot<br>encoded<br>labels | Neural<br>Networks   |
| Sparse<br>Categorical<br>Crossentropy      | Similar to Categorical<br>Crossentropy  | <pre>loss = keras.losses.SparseCategoricalCrossentropy()</pre> | Multi-class<br>classification<br>with integer<br>labels                | Neural<br>Networks   |
| Hinge Loss                                 | $(\max(0, 1 - y_{	ext{true}} \cdot y_{	ext{pred}}))$  | <pre>loss = keras.losses.Hinge()</pre>                         | Binary<br>classification   | SVM, Neural<br>Networks                                      |
| Kullback-<br>Leibler<br>Divergence<br>(KL) | $(\sum y_{	ext{true}} \log(rac{y_{	ext{true}}}{y_{	ext{pred}}}))$  | <pre>loss = keras.losses.KLDivergence()</pre>                  | Comparing<br>two<br>probability<br>distributions                       | Variational<br>Autoencoders<br>(VAEs),<br>Neural<br>Networks |