# **Fraud Email using BERT Classification**

# **Detailed Breakdown of Functions and Their Documentation**

# 1. FocalLoss Class

#### **Purpose:**

The FocalLoss class is a custom loss function designed to address class imbalance by focusing more on difficult or misclassified examples during training.

#### Methods:

• \_\_init\_\_(self, alpha=0.25, gamma=2.0, reduction='mean')

# **Description:**

Constructor to initialize the focal loss with configurable parameters.

#### Parameters:

- o alpha (float): Weighting factor for the loss of each class (default 0.25).
- gamma (float): Focusing parameter to reduce the loss contribution from easy examples (default 2.0).
- reduction (str): Specifies the reduction to apply to the output ('mean', 'sum', or 'none').

# forward(self, inputs, targets)

#### **Description:**

Computes the focal loss given the model's raw predictions (inputs) and the true labels (targets).

#### **Parameters:**

- o inputs (Tensor): Model predictions (logits) for each class.
- o targets (Tensor): Ground truth class labels.

#### Returns:

 A Tensor representing the computed focal loss based on the reduction specified during initialization.

#### 2. Additional Functions and Classes

(Note: The notebook further defines functions for data processing, model training, and evaluation. The following is a summary of these components as derived from the notebook.)

## a. Data Processing Functions

These functions include:

#### Dataset Loader Class (Custom Dataset)

#### **Purpose:**

A custom Dataset class (e.g., EmailDataset) is implemented to handle tokenization and conversion of email texts into the format required by the BERT model.

#### **Documentation:**

- \_\_init\_\_(self, dataframe, tokenizer, max\_length):
  Initializes the dataset with a pandas DataFrame containing the email data, a tokenizer instance, and a maximum sequence length for BERT.
- \_\_len\_\_(self):Returns the total number of samples.
- \_\_getitem\_\_(self, index):
  Retrieves the tokenized email and its corresponding label for the given index.

#### b. Model Training and Evaluation Functions

Functions here facilitate training and evaluating the model:

#### Training Function (e.g., train\_model)

#### **Purpose:**

Runs the training loop for the model, iterating over batches of training data, computing loss (using focal loss or cross-entropy), and updating model weights via backpropagation.

#### **Documentation:**

o train\_model(model, dataloader, optimizer, scheduler, loss\_fn, device):

Trains the BERT model for one epoch.

#### **Parameters:**

model: The BERT model instance.

- dataloader: A DataLoader object containing the training data.
- optimizer: Optimizer for updating model parameters (e.g., AdamW).
- scheduler: Learning rate scheduler (e.g., cosine schedule with warmup).
- loss fn: Loss function (custom FocalLoss or another).
- device: Device (CPU/GPU) on which to run the training. Returns:
- The average training loss for the epoch.

# Evaluation Function (e.g., evaluate\_model)

#### **Purpose:**

Evaluates the performance of the model on a validation or test set.

# **Documentation:**

# o evaluate\_model(model, dataloader, loss\_fn, device):

Computes validation loss and other metrics (accuracy, F1-score, precision, recall).

#### **Parameters:**

- model: The BERT model instance.
- dataloader: A DataLoader object containing the evaluation data.
- loss fn: Loss function used for evaluation.
- device: Device (CPU/GPU) for evaluation. Returns:
- A dictionary containing evaluation metrics and loss.

#### c. Data Augmentation Function

### • Augmentation using EDA (Easy Data Augmentation)

# **Purpose:**

Applies augmentation techniques (synonym replacement, random insertion, etc.) to enrich the dataset and improve model robustness.

#### **Documentation:**

o augment\_text(text):

Augments a given text string using EDA.

#### Parameters:

- text (str): The input email text. Returns:
- A new text string that has been augmented.