# TF-IDF: Term Frequency-Inverse Document Frequency

Your Name

March 24, 2025

## Conceptual Overview

- ► TF-IDF aims to reflect how important a word is to a document in a collection (corpus).
- It combines two metrics:
  - Term Frequency (TF): How often a term appears in a document.
  - ► Inverse Document Frequency (IDF): How rare a term is across the entire corpus.

### **Example Corpus**

#### Consider a small corpus of three call transcripts:

- ▶ Document 1 (D1): "Congratulations! You've won a prize. Please call to claim your reward."
- ▶ Document 2 (D2): "Urgent notice! Your account has been compromised. Call immediately to secure it."
- ▶ Document 3 (D3): "Hello, this is a standard call for account verification. Please verify your information."

# Step 1: Term Frequency (TF)

TF measures how often each term appears in each document (raw count):

Term	D1	D2	D3
Congratulations	1	0	0
Won	1	0	0
Prize	1	0	0
Please	1	0	1
Call	1	1	1
Account	0	1	1
Secure	0	1	0
Verification	0	0	1

Table: Raw Term Frequency Count in Each Document

## Step 2: Compute Term Frequency (TF)

#### Formula:

$$TF(t,d) = \frac{\text{Number of times term } t \text{ appears in document } d}{\text{Total number of terms in } d}$$

Example for "Call" in D1 (assuming D1 has 10 words):

$$TF(Call, D1) = \frac{1}{10} = 0.1$$

# Step 3: Compute Inverse Document Frequency (IDF)

#### Formula:

$$IDF(t) = \log\left(\frac{N}{df(t)}\right)$$

where:

- $\triangleright$  N = Total number of documents in the corpus.
- ightharpoonup df(t) =Number of documents containing term t.

Example for "Call":

$$IDF(Call) = \log\left(\frac{3}{3}\right) = \log(1) = 0$$

Example for "Congratulations":

$$IDF(Congratulations) = log\left(\frac{3}{1}\right) = log(3) \approx 0.477$$

## Step 4: Compute TF-IDF

#### Formula:

$$TF$$
- $IDF(t, d) = TF(t, d) \times IDF(t)$ 

Example for "Congratulations" in D1:

$$TF$$
- $IDF$ (Congratulations,  $D1$ ) =  $0.1 \times 0.477 = 0.0477$ 

Example for "Call" (appears in all documents, IDF = 0):

$$TF$$
- $IDF$ (Call,  $D1$ ) =  $0.1 \times 0 = 0$ 

#### Final TF-IDF Table

Term	D1	D2	D3
Congratulations	0.0477	0	0
Won	0.0477	0	0
Prize	0.0477	0	0
Please	0.0477	0	0.0477
Call	0	0	0
Account	0	0.0477	0.0477
Secure	0	0.477	0
Verification	0	0	0.477

Table: TF-IDF Scores for Terms in Each Document

#### Conclusion

- TF-IDF helps identify important terms in a document.
- Words common in all documents (like "Call") get \*\*low\*\*
  TF-IDF scores.
- ▶ Rare words (like "Secure" and "Verification") get \*\*high\*\* TF-IDF scores, highlighting their importance.
- ► This numerical representation of text is then used as features for machine learning models.