# **Training Day-74 Report:**

Simple Neural Network with TensorFlow, Word Embedding, and understanding CBOW and Skip-gram models.

### 1. Simple Neural Network with TensorFlow

#### Goal:

Build a neural network using TensorFlow to solve a binary classification problem (e.g., XOR).

## **Implementation:**

```
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
```

```
# XOR dataset

X = [[0, 0], [0, 1], [1, 0], [1, 1]]

y = [0, 1, 1, 0]

# Define a sequential model

model = Sequential([

Dense(8, activation='relu', input_shape=(2,)), # Hidden layer with 8 neurons

Dense(1, activation='sigmoid') # Output layer for binary classification
])

# Compile the model

model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])

# Train the model

model.fit(X, y, epochs=100, verbose=1)
```

```
# Evaluate the model
print("Evaluation Results:", model.evaluate(X, y))
```

# Predictions

```
print("Predictions:", model.predict(X))
```

This simple NN demonstrates how to use TensorFlow to solve small problems with minimal code.

#### 2. Word Embedding

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Word embeddings convert words into dense vectors in a continuous vector space, capturing semantic meanings.

### **Using TensorFlow/Keras:**

```
from tensorflow.keras.models import Sequential
```

from tensorflow.keras.layers import Embedding, Flatten, Dense

```
# Sample vocabulary and corresponding tokenized sentences
vocab_size = 50
embedding dim = 8
max length = 10
# Example tokenized input
sentences = [[1, 2, 3, 4], [3, 4, 1, 2]]
padded sentences = tf.keras.preprocessing.sequence.pad sequences(sentences,
maxlen=max length)
# Define the embedding layer model
model = Sequential([
  Embedding(input dim=vocab size, output dim=embedding dim,
input length=max length),
  Flatten(),
  Dense(1, activation='sigmoid')
```

```
# Compile and summarize the model
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
model.summary()

# Dummy output labels
labels = [1, 0]

# Train the model
model.fit(padded_sentences, labels, epochs=10)
```

This creates word embeddings for a vocabulary and uses them in a classification task.

#### 3. CBOW (Continuous Bag of Words) & Skip-gram

CBOW and Skip-gram are two approaches used in Word2Vec for learning word embeddings.

#### **CBOW:**

- Predicts a target word from its surrounding context words.
- Suitable for frequent words in a corpus.

#### Skip-gram:

- Predicts context words from a target word.
- Suitable for infrequent words.

#### **TensorFlow Implementation of Skip-gram:**

```
import numpy as np
import tensorflow as tf
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, Embedding, Dot, Flatten
```

```
# Sample dataset

word_pairs = [(1, 2), (1, 3), (2, 3), (2, 4)] # (target, context)

vocab_size = 5 # Vocabulary size

embedding_dim = 8
```

```
# Prepare inputs and labels
targets, contexts = zip(*word_pairs)
targets = np.array(targets)
contexts = np.array(contexts)
# Define the model
input target = Input((1,))
input\_context = Input((1,))
embedding = Embedding(vocab size, embedding dim, input length=1)
target embedding = embedding(input target)
context embedding = embedding(input context)
dot product = Dot(axes=-1)([target embedding, context embedding])
output = Flatten()(dot product)
model = Model(inputs=[input target, input context], outputs=output)
model.compile(optimizer='adam', loss='mse')
# Train the model
model.fit([targets, contexts], np.ones(len(word pairs)), epochs=100)
This implementation focuses on Skip-gram, where the network learns word embeddings by
maximizing the similarity between target and context words.
```

#### **Summary:**

- Simple NN: Built a basic neural network using TensorFlow for XOR classification.
- **Word Embedding**: Showed how to use TensorFlow's Embedding layer for converting words into dense vectors.

CBOW & Skip-gram: Explained concepts and provided a Skip-gram implementation in TensorFlow.		