

## WEEK 3 ASSIGNMENT

### 1. Short Answer Questions

- **Q1: Explain the primary differences between TensorFlow and PyTorch. When would you choose one over the other?**

TensorFlow is a framework for large-scale deep learning with strong deployment support, while PyTorch is more flexible and Pythonic, great for research and experimentation.

I would choose TensorFlow for production models and PyTorch for fast prototyping and academic work.

- **Q2: Describe two use cases for Jupyter Notebooks in AI development.**
  - (i) Used for interactive coding and visualization of data and models
  - (ii) Used for experiment tracking, where codes output and notes are kept together.
- **Q3: How does spaCy enhance NLP tasks compared to basic Python string operations?**

spaCy provides pre-trained NLP models and efficient tokenization, tagging and parsing, unlike basic string operations that handle text only at the character or word level.

### 2. Comparative Analysis

- **Compare Scikit-learn and TensorFlow in terms of:**
  - **Target applications** (e.g., classical ML vs. deep learning).
  - **Ease of use for beginners.**
  - **Community support.**

| Feature             | Scikit-learn  | TensorFlow   |
|---------------------|---|--|
| Target applications | Classical ML (regression, Classification, Clustering) | Deep learning (neural networks, large-scale AI models) |
| Ease of use         | Easier for beginners; simple API                      | Steeper learning curve; more complex setup             |
| Community support   | Large for ML tasks                                    | Huge for AI/ deep learning, supported by google        |

## **ETHICS & OPTIMIZATION**

### **Potential biases:**

- If trained on only certain handwriting styles, the model may fail on unfamiliar styles.
- Digit misclassification can affect applications like digit-based forms, banking, or accessibility tools.

### **Mitigation:**

- Use **TensorFlow Fairness Indicators** to analyse prediction fairness across subgroups (e.g., handwriting styles)
- Use **data augmentation** to diversify training data (rotations, scaling, slant variations).

### Amazon Reviews (NLP) example:

#### Potential biases:

- If reviews contain biased language, sentiment analysis may unfairly favor/penalize certain brands or products.
- Rule-based sentiment might misinterpret sarcasm or context.

#### Mitigation:

- Use spaCy rule-based systems to detect biased words or entities.
- Combine with lexicons or pre-trained models for sentiment correction.

## **Troubleshooting Challenge**

- **Buggy Code:** A provided TensorFlow script has errors (e.g., dimension mismatches, incorrect loss functions). Debug and fix the code.

Dimension mismatch (The input data wasn't reshaped correctly for the CNN)  
to fix, reshaped the images to include the channel dimension using:

```
x_train = x_train.reshape(-1, 28, 28, 1)
x_test = x_test.reshape(-1, 28, 28, 1)
```

Incorrect loss function (The model used a regression loss instead of a classification one)  
to fix, changed the loss to `sparse_categorical_crossentropy` for multi-class classification:

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