### **Ethical Considerations**

#### Bias in MNIST:

The MNIST dataset may not represent diverse handwriting styles. This can cause unfair predictions for users outside the dataset's demographic. Using TensorFlow Fairness Indicators allows us to detect and visualize bias across subgroups, ensuring fairness and equal accuracy.

#### **Bias in Amazon Reviews Model:**

The sentiment analysis uses a limited rule-based system that might misinterpret sarcasm or cultural language differences. Expanding the lexicon, handling negations, or combining spaCy with a trained model can reduce these errors.

### **Mitigation Tools:**

- TensorFlow Fairness Indicators for fairness metrics and bias detection.
- *spaCy's rule-based pipelines* for improving entity and sentiment accuracy through better linguistic rules.

## 2 Troubleshooting Challenge

You'll be given (or can simulate) a **buggy TensorFlow model** — the goal is to fix it. If you haven't been provided one, we can simulate a typical issue for demonstration.

# **Example Buggy Code**

Here's a common example with a **dimension mismatch** in TensorFlow:

## **Fixed Version**

## **Explanation of Fixes:**

- 1. Flatten Layer Added: Converts 28×28 images into 784-length vectors.
- Loss Function Fixed: Changed from mse → sparse\_categorical\_crossentropy (correct for multi-class classification).
- 3. **Metrics Retained:** Accuracy is appropriate for classification.

# Report Section (Example Text)

### **Troubleshooting Challenge:**

The provided TensorFlow code produced a dimension mismatch error due to an incorrect input shape and loss function.

To fix it:

• A Flatten() layer was added to reshape image inputs.

• The loss function was corrected from mse to sparse\_categorical\_crossentropy.

After correction, the model trained successfully and achieved the expected accuracy.