

## Class 10

## Subject: ARTIFICIAL INTELLIGENCE

## ANSWER KEY

### A. Multiple Choice Questions (1 mark each)

(5 × 1 = 5 marks)

1. **B)** To test and improve the model's performance using unseen data
2. **C)** Deep Learning
3. **B)** Object Detection
4. **B)** Transparency and accountability
5. **B)** Semantic Analysis

### B. Short-Answer Questions (2 marks each)

(2 × 2 = 4 marks)

#### 6. Precision vs Recall

- **Precision** measures how many of the predicted positive results are actually correct.

$$\text{Precision} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Positives}}$$

- **Recall** measures how many actual positive cases the model correctly identifies.

$$\text{Recall} = \frac{\text{True Positives}}{\text{True Positives} + \text{False Negatives}}$$

#### Example (Recall more critical):

In **medical diagnosis (e.g., detecting cancer)**, Recall is more critical than Precision because it is more important to identify *all* potential patients (even if some are false alarms) than to miss a real case.

#### 7. Importance of Hidden Layers in Neural Networks

Hidden layers help the network **learn complex patterns and representations** from data by performing **non-linear transformations**.

- They extract intermediate features between input and output.
- More hidden layers or neurons generally increase the **learning capacity**, allowing the model to capture deeper relationships — though too many can lead to overfitting.

## C. Application-Based / Analytical Questions (3 marks each)

### 8. Crop Yield Prediction

#### a) Type of AI Model:

- This is a **Regression** problem since the target (crop yield) is a **continuous value**.

#### b) Ensuring Reliability:

- **Train-test split:** Divide data (e.g., 80% train, 20% test) to ensure that model performance is evaluated on unseen data.
  - **Evaluation Metrics:** Use metrics such as **Mean Absolute Error (MAE)**, **Root Mean Squared Error (RMSE)**, or **R<sup>2</sup> score** to assess accuracy.
- These steps prevent overfitting and ensure the model generalizes well.

### 9. Comparison: Computer Vision vs NLP

Aspect	Computer Vision	Natural Language Processing (NLP)
<b>Type of Data</b>	Image or video data (pixels, colors, shapes)	Text or speech data (words, sentences, grammar)
<b>Tools / Libraries</b>	OpenCV, TensorFlow, Keras, PyTorch	NLTK, spaCy, Hugging Face Transformers
<b>Ethical Concerns (Examples)</b>	Facial recognition may violate privacy or exhibit racial bias	Resume screening or sentiment analysis may reinforce gender or language bias

## D. Long-Answer Case-Based Question (5 marks)

### 10. Case Study: Ethical and Technical Evaluation of AI Recruitment Model

Where bias could have been introduced:

- **Data Collection:** If training data included more male resumes or gendered language (e.g., “leadership” associated more with males).
- **Feature Selection:** If gender-related words were unintentionally included as predictors.

- **Model Training:** Biased weighting during optimization could amplify data imbalances.

### Evaluation Metrics to Reveal Bias:

- **Confusion Matrix** (split by gender) — reveals disproportionate false negatives/positives.
- **Fairness Metrics** such as **Demographic Parity**, **Equal Opportunity**, or **Disparate Impact Ratio** to measure bias across groups.
- **Precision/Recall per subgroup** — to check whether one gender consistently gets lower scores.

### Applying Ethical Principles:

Principle	Application in Model Redesign
<b>Autonomy</b>	Allow candidates to understand and challenge automated decisions; ensure transparency in criteria.
<b>Justice</b>	Ensure fair treatment — use balanced datasets and fairness-aware algorithms to remove gender bias.
<b>Non-maleficence</b>	Avoid harm by preventing discriminatory outcomes that could harm job seekers' opportunities.
<b>Beneficence</b>	Use AI to <i>benefit all</i> applicants — e.g., anonymized resumes or bias mitigation techniques.