Level : M2-IA (2023/2024) Subject : Deep Learning Dr. F.BOUGAMOUZA

## FIRST PARTIAL EXAM

## CHOOSE THE CORRECT OPTION FROM THE CHOICES PROVIDE (Select all that apply)

Question1: Select the applications that commonly use deep learning techniques:

- Image Recognition
- Speech Processing
- Sentiment Analysis
- Drug Discovery

Question2: Which activation functions are commonly used in deep neural networks?

- Sigmoid
- **ReLU (Rectified Linear Unit)**
- Softmax
- Linear Activation

Question3: Choose the common deep learning frameworks:

- TensorFlow
- PyTorch
- Keras

Question 4: Which tasks fall under supervised learning in deep learning?

- Image Classification
- ⊖ Clustering

← Structure Discovery

• Regression

Question 5: Choose steps that do not play a part in training a neural network in the context of deep learning.

- ← Forward Propagation
- ⊖ Backward Propagation
- Feature Engineering
- $\odot$  Optimization

Question 6: Select the elements affecting the computational demands in deep learning models:

- Model complexity
- Data size
- Learning rate
- Hardware resources

Question 7: Select the characteristics of the Rectified Linear Unit (ReLU) activation function:

- Aids in mitigating the vanishing gradient problem
  - Fast computation due to linearity
  - ← Provides output between -1 and 1
  - Efficiently handles negative input values

**Question 8:** Choose the activation functions suitable for output in binary classification tasks:

- ${\scriptstyle \ominus Softmax}$
- $\circ$  Sigmoid
- → Hyperbolic Tangent (tanh)
- $\odot$  Leaky ReLU

**Question 9:** Select the purpose of the Hyperbolic Tangent (tanh) activation function:

•—Similar to ReLU but computationally faster

- Mitigates the vanishing gradient problem
- → Provides output between 0 and 1
- Maps input values between -1 and 1

Question 10: Choose the characteristic of the Softmax activation function:

• Ideal for output in multi-class classification tasks

o-Introduces linearity in neural networks

• Mitigates the vanishing gradient problem

•---Suitable for regression tasks

Question11: Choose the activation functions effective in handling the vanishing gradient problem:

- Leaky ReLU
- $\circ \quad \textbf{Sigmoid}$
- o Tanh
- → Softmax

Question 12: Sources of Bias in Machine Learning:

- Biased Labels
- Dataset Imbalance
- Model Architecture
- Computational resources
- Question 13: Bias in Machine Learning refers to:
  - An algorithm's inherent prejudice towards certain outcomes
  - The variance between training and testing data
  - An error in the model's predictions
  - The model's inability to learn from the data
- Question 14: Common Types of Bias in Datasets:
  - o Group Bias
  - Data Sparsity
  - **Overfitting**
  - Precision Bias
- Question 15: Ethical Consideration in Addressing Bias involves:
  - Preprocessing techniques to balance data
  - o-Ignoring biased data to avoid model inaccuracies
  - o-Limiting the dataset to avoid controversial information
  - •---- Increasing bias for fair representation
- Question 16: Bias mitigation techniques involve:
  - Increasing bias to improve model accuracy
  - **Removing biased attributes from the dataset**

  - Using biased data exclusively for testing
- Question 17: Unintended consequences of biased models include:
  - Increased fairness in predictions
  - Reinforcement of societal prejudices
  - → Overfitting issues
  - --- Reduced model complexity
- Question 18: Bias in Machine Learning affects:
  - Model predictions and outcomes
    - Data collection only

- Training time
- Legal and ethical concerns

Question 19: Underrepresentation in training data can lead to:

- Increased bias in predictions
- $\odot$  Improved model generalization
- → Reduced computational complexity
- Decreased model performance

Question 20: Addressing Bias in Machine Learning involves:

---- Introducing more bias for comprehensive analysis

• Regular auditing and continuous improvement

•----Restricting model access to biased data only

Question 21: Data Types in Datasets:

- $\circ$  Numerical
- Categorical
- Text
- Audio
- Image

Question 22: Sources of Bias in Machine Learning Models:

- Feature Selection
- Algorithm Design
- Training Data
- Evaluation Metrics
- Model Interpretation
- Question 23: Ways to Mitigate Bias in Datasets:
  - Data Augmentation
  - Bias Correction Algorithms

  - ← Fairness aware Model Training
- Question 24: Impact of Bias in AI Systems:
  - Unfair Predictions
  - Reinforcing Stereotypes
  - ← Reduced Model Complexity
  - Inequality in Outcomes
- Question 25: Types of Algorithmic Bias:
  - Racial Bias
  - Gender Bias
  - Age Bias
  - Socio-economic Bias
  - Cultural Bias
- Question 26: Bias Mitigation Strategies in AI Ethics:
  - Diversity in Data Collection
  - Explainability in Models
  - Regular Ethical Audits
  - Incorporating Ethical Frameworks
- **Question 27:** Factors Affecting Computational Demands:
  - Model Complexity
    - Dataset Size

- Algorithm Selection
- Hardware Infrastructure
- Hyperparameter Tuning

Question 28: Hardware Requirements for Deep Learning:

- GPU (Graphics Processing Unit)
- CPU (Central Processing Unit)
- TPU (Tensor Processing Unit)
- FPGA (Field-Programmable Gate Array)
- HPC (High-Performance Computing Clusters)

Question 29: Adapting Hardware to Model Size:

- Larger Models Require Higher Memory Bandwidth
- Smaller Models Are More Computationally Intensive
- ---- Model Size Does Not Affect Hardware Adaptation
- Hardware Adaptation Depends on Algorithm Choice
- Specialized Hardware Is Essential for Large Models

Question 30: Factors Influencing Model Interpretability:

- Model Complexity
  - Choice of Activation Function
  - Feature Importance
  - Explainable AI Techniques
  - Training Data Size
- **Question 31:** Interpreting Model Outputs:
  - Understanding Model Confidence Levels
  - Identifying Prediction Boundaries
  - Explaining False Positive Predictions
  - Analyzing Model Uncertainty
  - ← Examining Loss Function Values

Question 32: Explainability Techniques:

- SHAP (SHapley Additive exPlanations)
- ← ReLU (Rectified Linear Unit)
- LIME (Local Interpretable Model-agnostic Explanations)
- PCA (Principal Component Analysis)

## Question 33: Trade-offs in Model Explainability:

- Higher Model Explainability Reduces Performance
- ↔ Model Explainability Always Enhances Model Trust
- Explainable Models Might Sacrifice Complexity
- Increasing Model Explainability slows down Generalization