Assignment 4: Continuous Skip-Gram Model, Image Segmentation, and Continuous Bag-of-Words Model

Question 1: Befine Continuous Skip-Gram Model

- \* Definition: Neural network-based word embedding model in NXP, part of the Word2Vec framework by Mikolov et al. (2013).
- \* Objective: Predicts surrounding context words for a target word, capturing semantic relationships.
- \* Key Features:
- \* Loal: Maximize probability of context words given a target word.
- \* Architecture: Input layer (target word), hidden layer (word embeddings), output layer (context probability distribution).
- \* Training: Adjusts word vectors to capture semantic similarities (e.g., "king" and "queen").
- \* Applications:

- \* Text classification
- \* Information retrieval
- \* Sentiment analysis

Question 2: Explain Image Segmentation and Object Detection

- \* Image Segmentation: Partitions an image into segments to identify objects.
- \* Types:
- \* Semantic Segmentation: Classifies each pixel (e.g., road, car).
- \* Instance Segmentation: Differentiates between instances of the same class.
- \* Techniques:
- \* Thresholding
- \* Clustering (e.g., K-means)
- \* Deep Learning (e.g., Fully Convolutional Networks)
- \* Object Setection: Identifies and localizes objects with bounding boxes and class labels.
- \* Techniques:
- \* Traditional: HOL, Haar Cascades

- \* Deep Learning: YOXO (real-time detection), Faster R-CNN (efficient detection)
- \* Applications:
- \* Autonomous vehicles
- \* Video surveillance
- \* Robotics
- \* augmented reality

Question 3: Discuss Continuous Bag-of-Words (CBOW) Model

- \* Definition: Word embedding model in Word2Vec, predicts target word based on context words.
- \* Key Features:
- \* Objective: Maximize probability of target word given context words.
- \* Architecture: Input (context words), hidden layer (word embeddings), output (target word probability).
- \* Training: Adjusts embeddings to improve target word prediction.
- \* Applications: Similar to Skip-Gram-used

in sentiment analysis, text classification, recommendation systems.

\* Comparison with Skip-Gram:

\* Input: CBOW uses context words; Skip-

Gram uses target word.

\* Performance: CBOW is faster and better for smaller datasets; Skip-Gram performs well on larger datasets with rare words.