

ENEE698P: Multi-Modal Semantic Communication Through Transformer-Aided Compression

Yoonkyo Jung | ECE Ph.D. Student

April 9, 2025

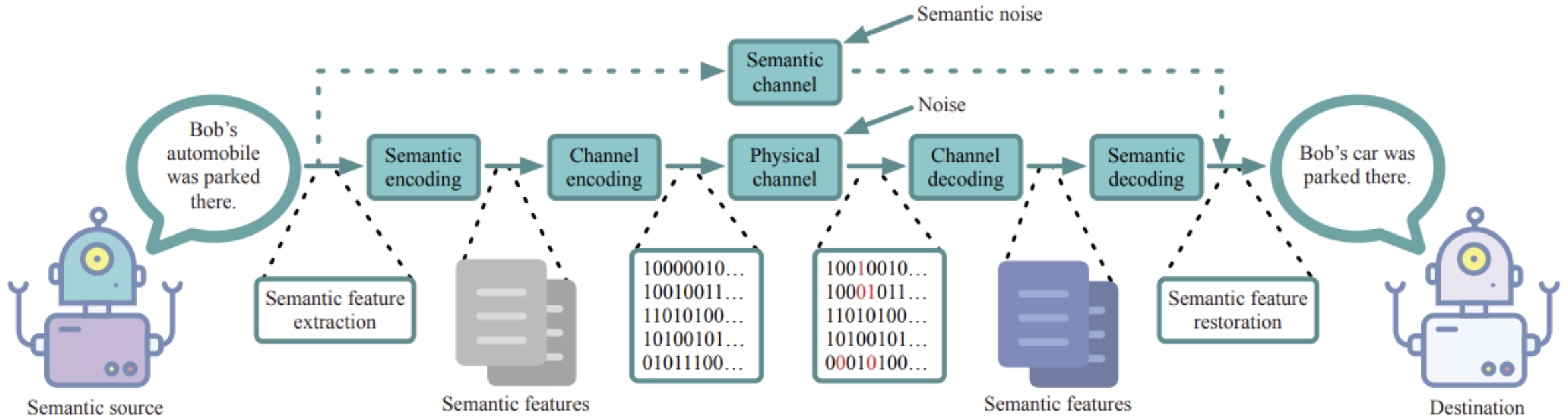
Contents

- Introduction
- Background
- System Model
- Results
- Conclusion

Introduction

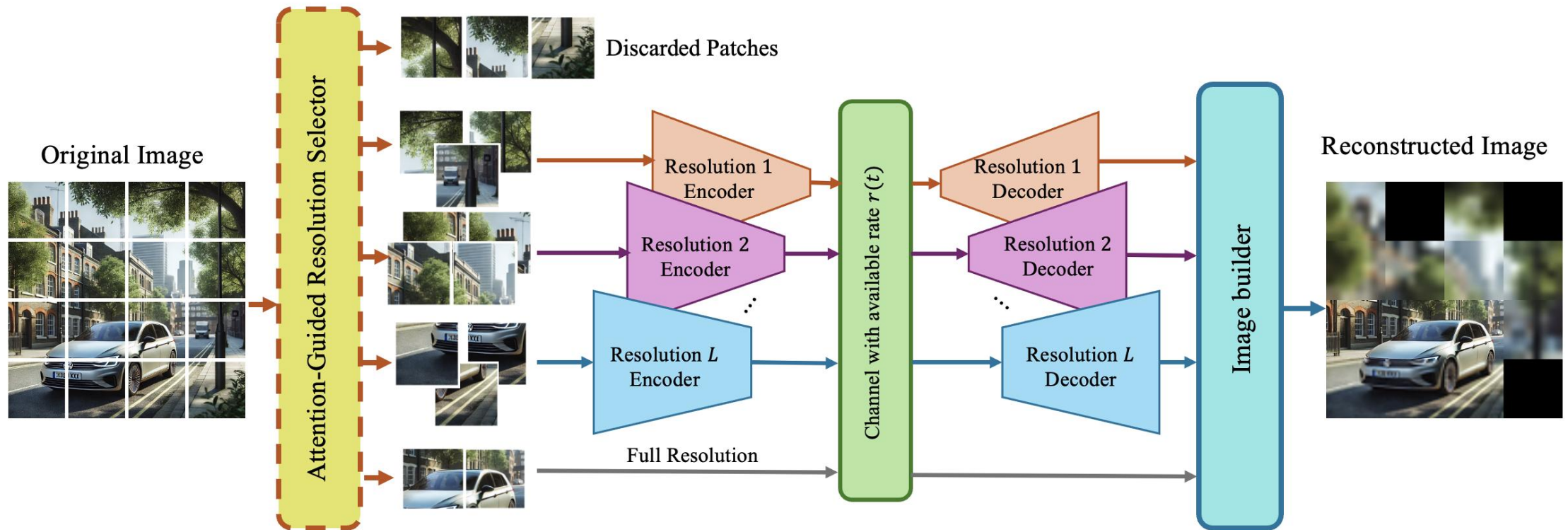
Introduction

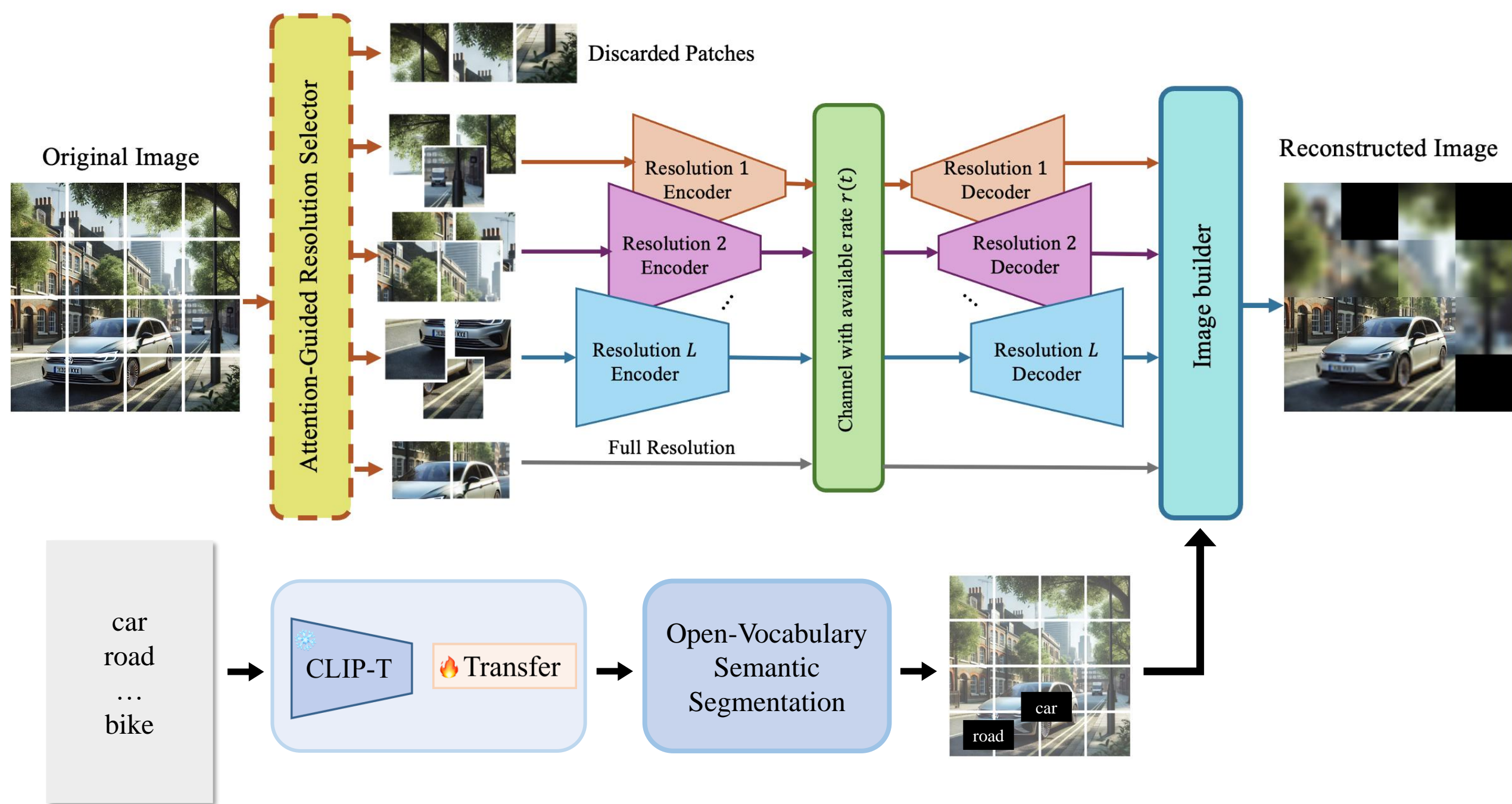
- Semantic communications is to extract the “meanings” or “features” of sent information from a source and “interpret” the semantic information at a destination.



Introduction

- Goal: Transmit multi-resolution data in limited bandwidth conditions.
- Developed transformer-based framework for channel-adaptive communication.

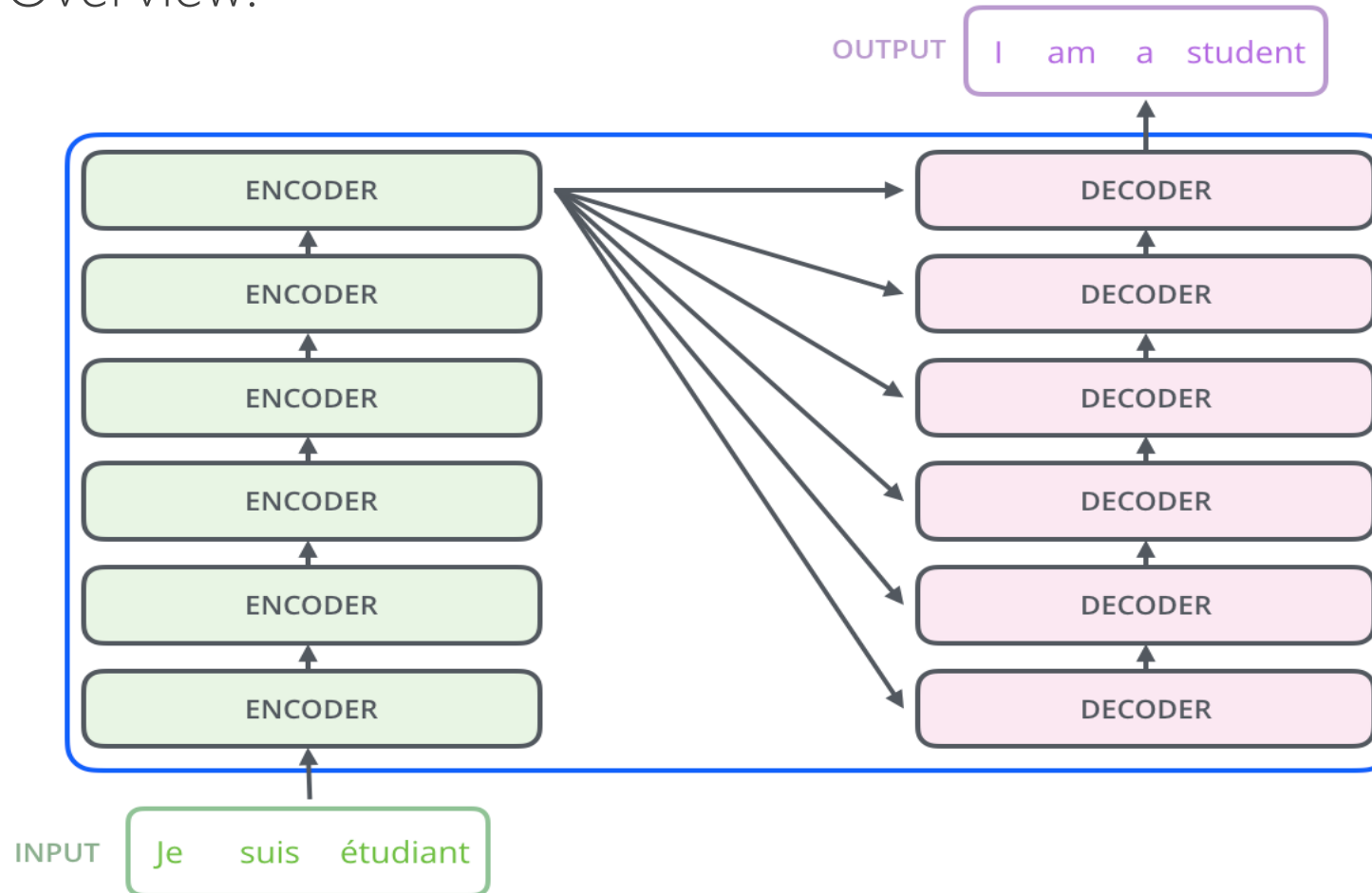




Background

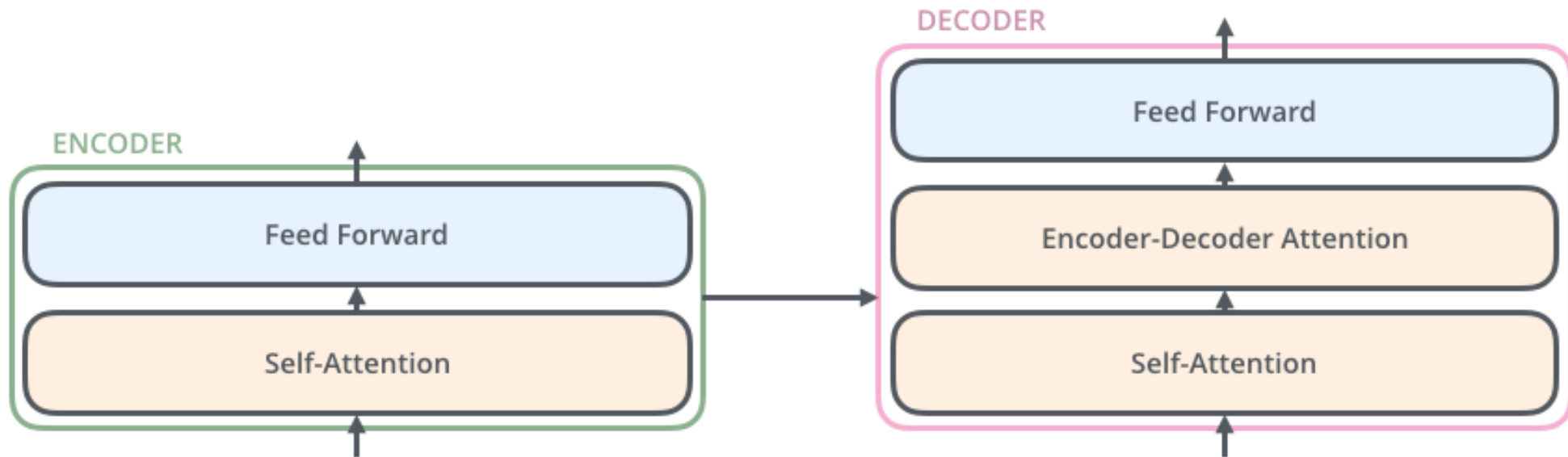
Background (Transformer)

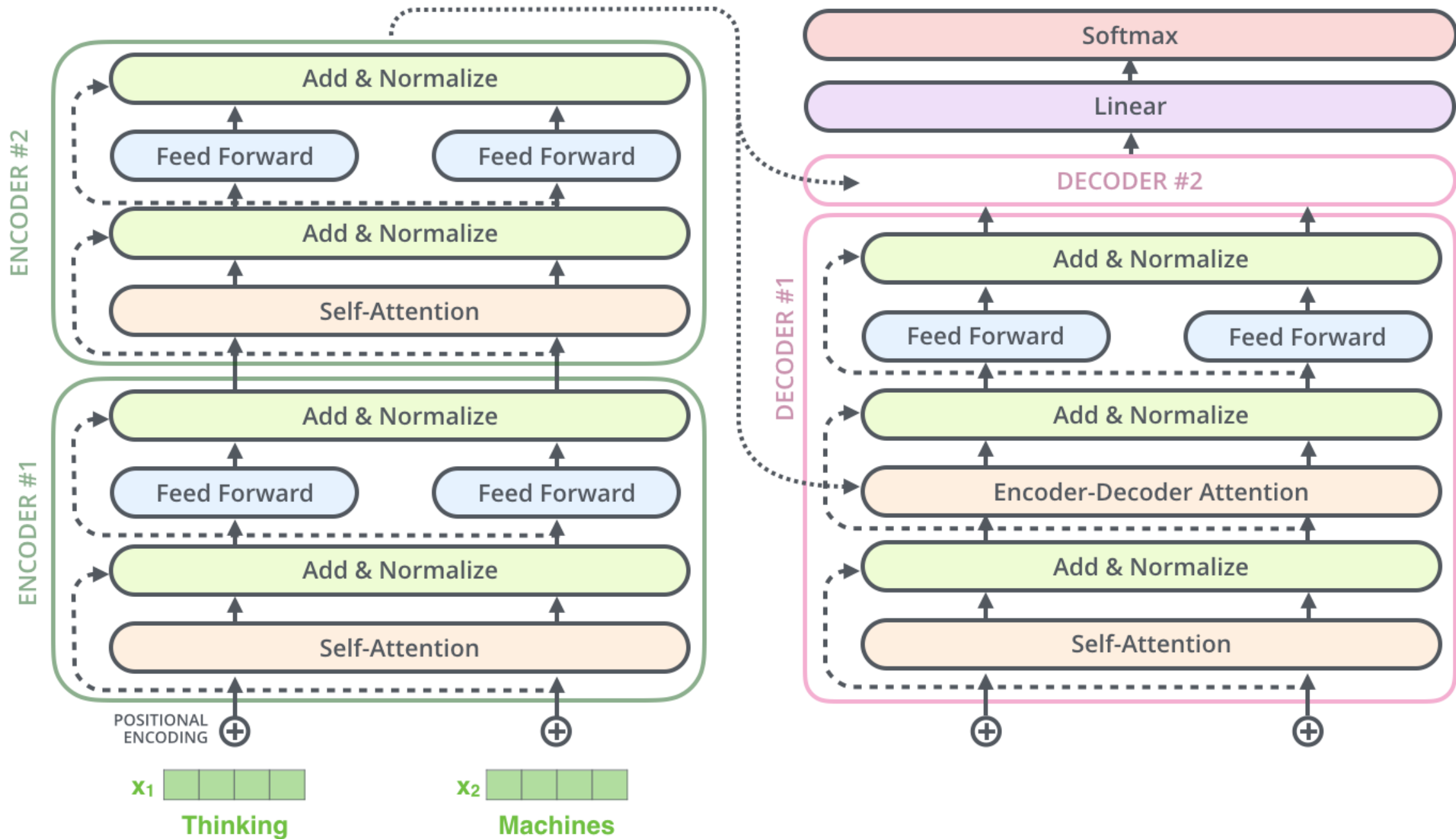
- Transformer Overview:



Background (Transformer)

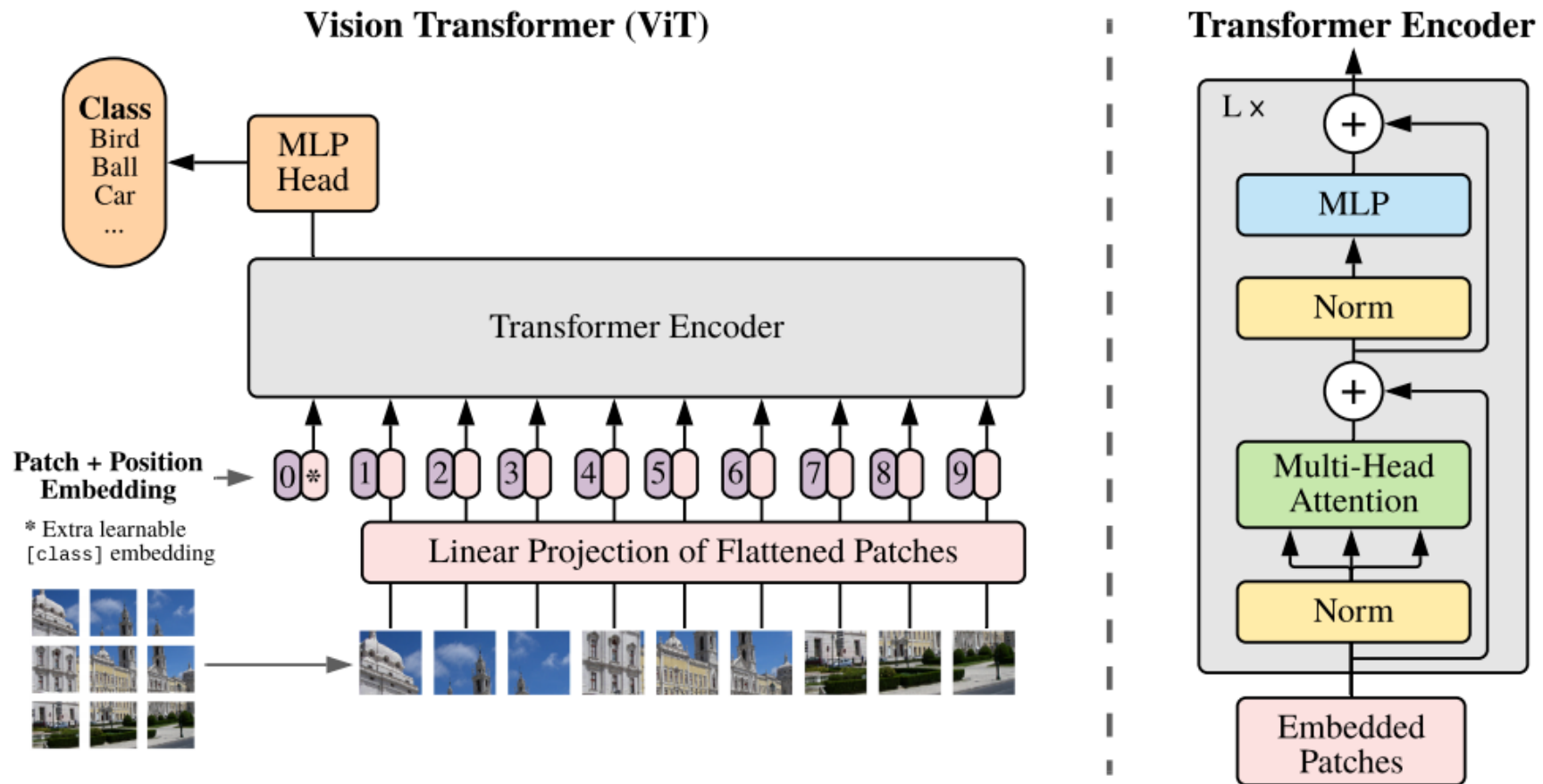
- Transformer Overview:
 - A deep learning architecture with self-attention mechanisms.
 - Enables focus on key elements in complex data.





Background (Vision Transformer)

- Vision Transformer (ViT)



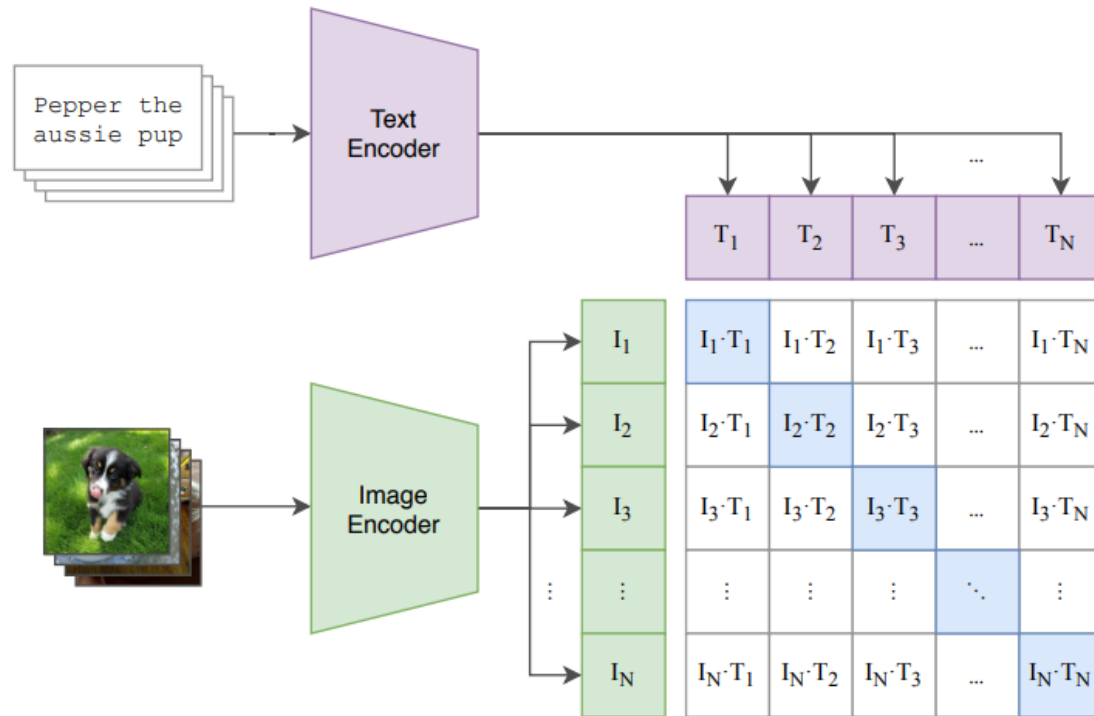
Background (Transformer & ViT)

- Key Features:
 - Multi-head attention
 - Positional encodings
 - Encoder-decoder structure
- Application for the paper:
 - Image patches encoded and compressed based on semantic content.

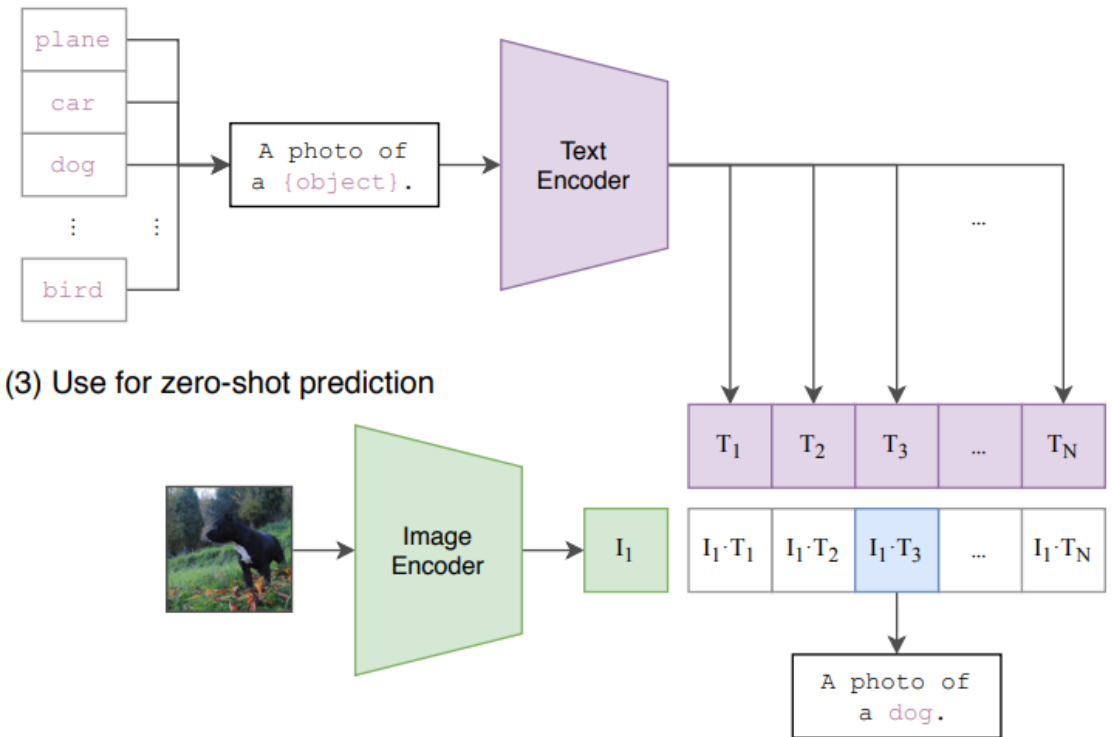
Background (CLIP)

- Contrastive Language–Image Pretraining (CLIP)
 - A model that predicts image-text similarity

(1) Contrastive pre-training



(2) Create dataset classifier from label text

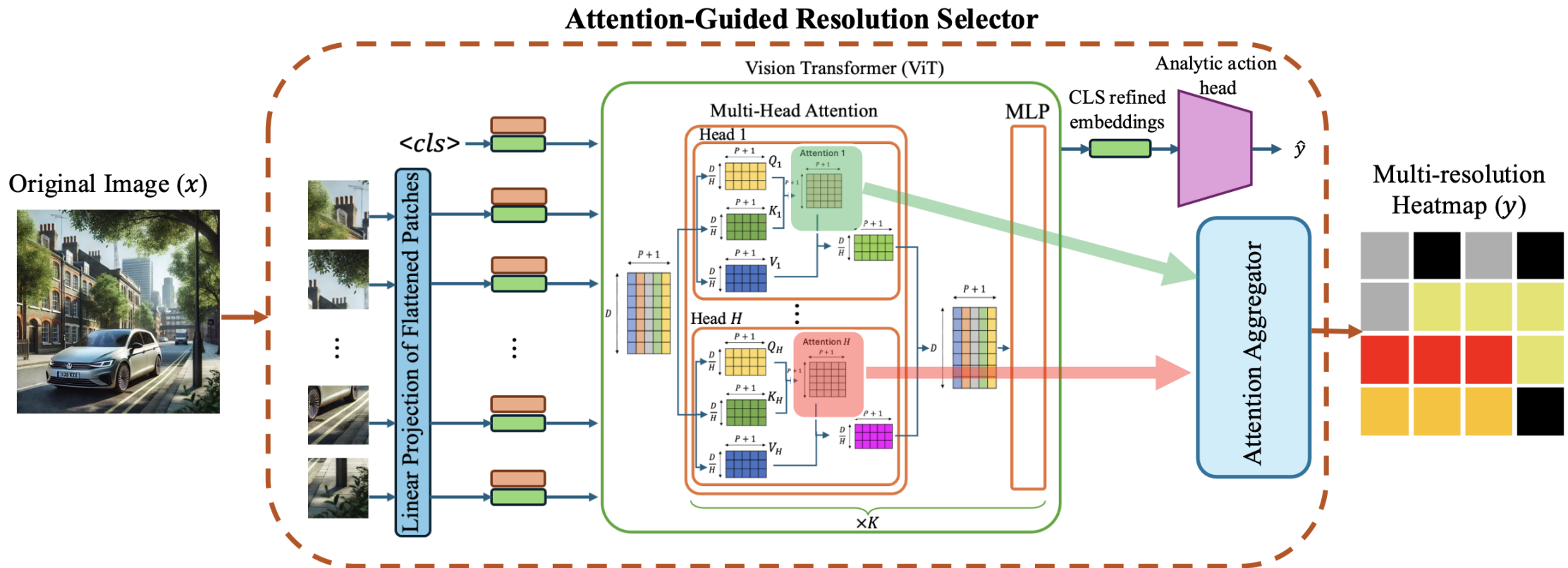


(3) Use for zero-shot prediction

System Model

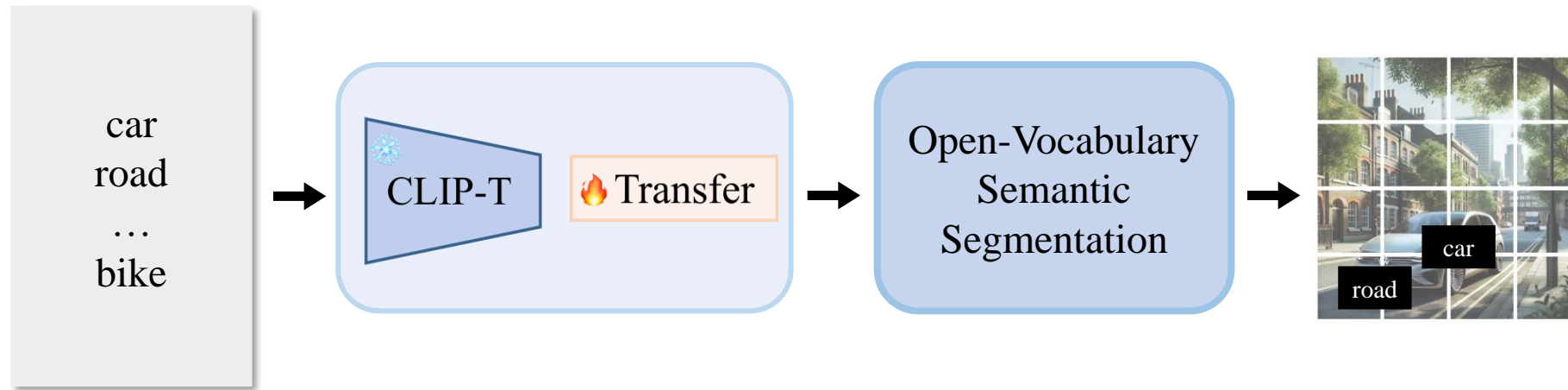
System Model

- Attention-Guided Resolution Selector determines the encoding resolution for each patch based on its semantic importance and available channel rate.



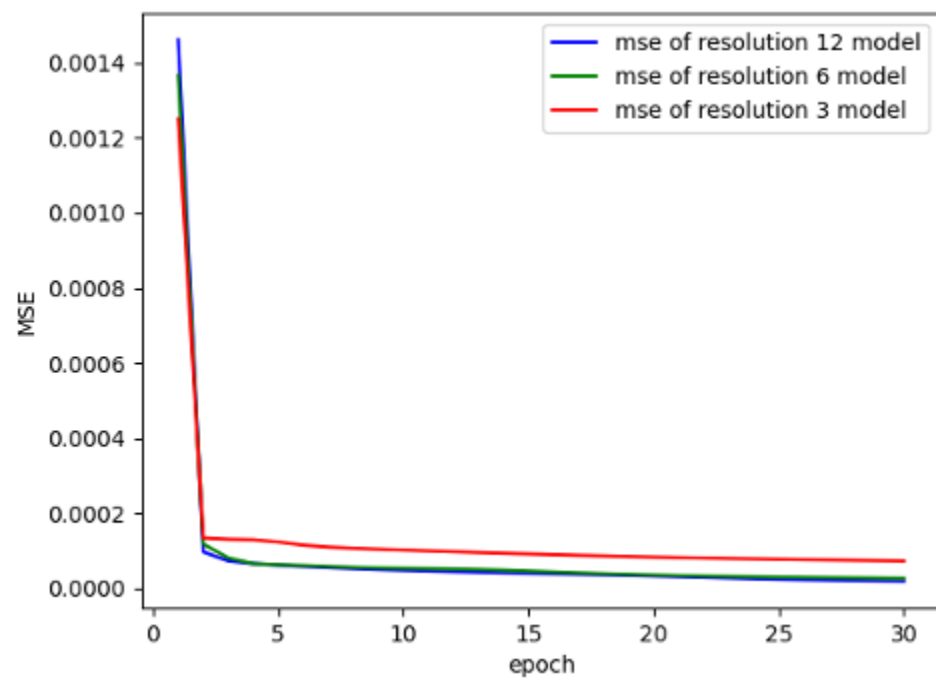
System Model

- Open-vocabulary segmentation framework find an attention score of images based on the text input

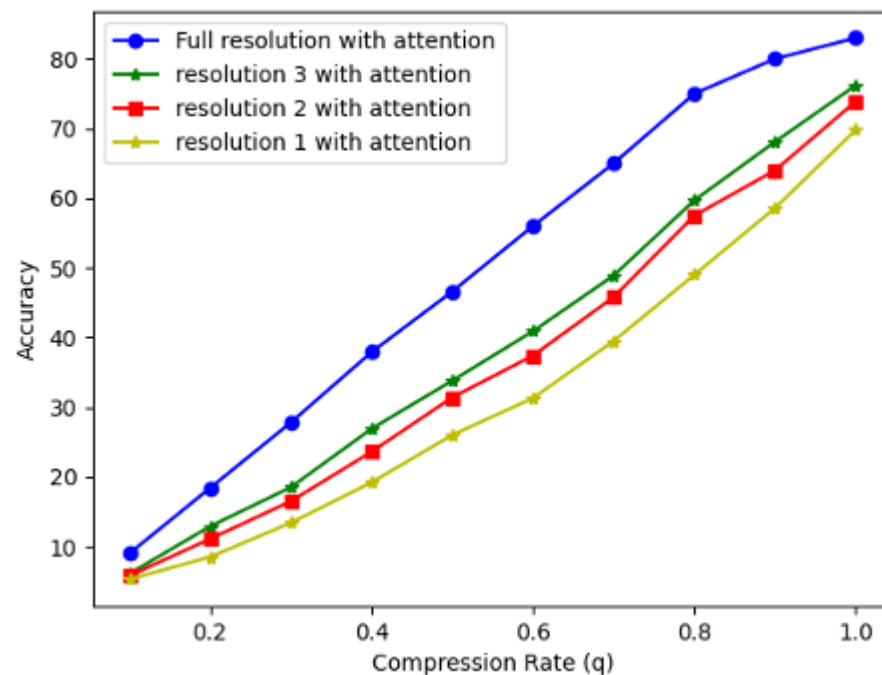


Results

Results

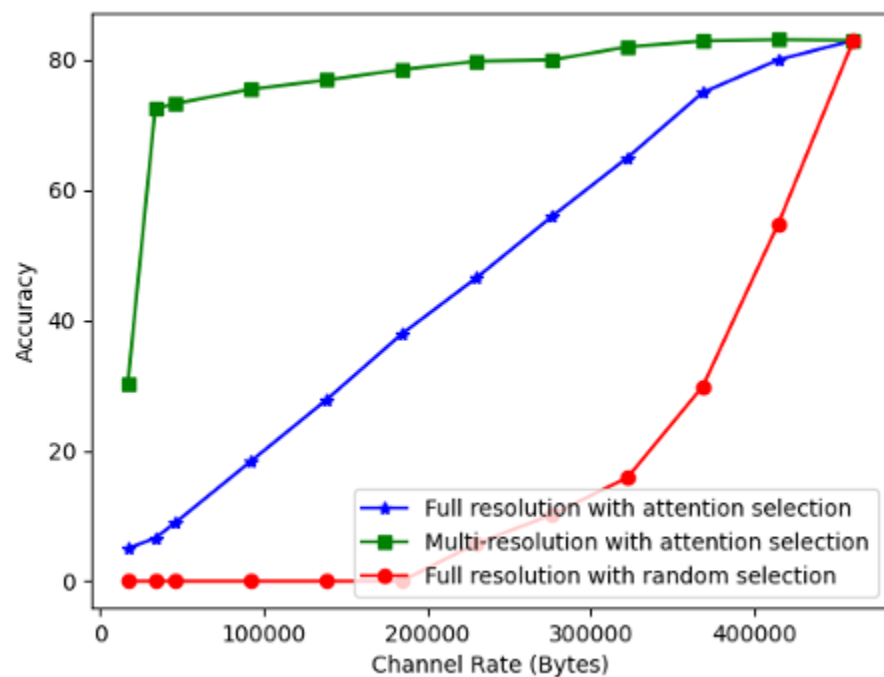


<Reconstruction result for three medium resolutions.>

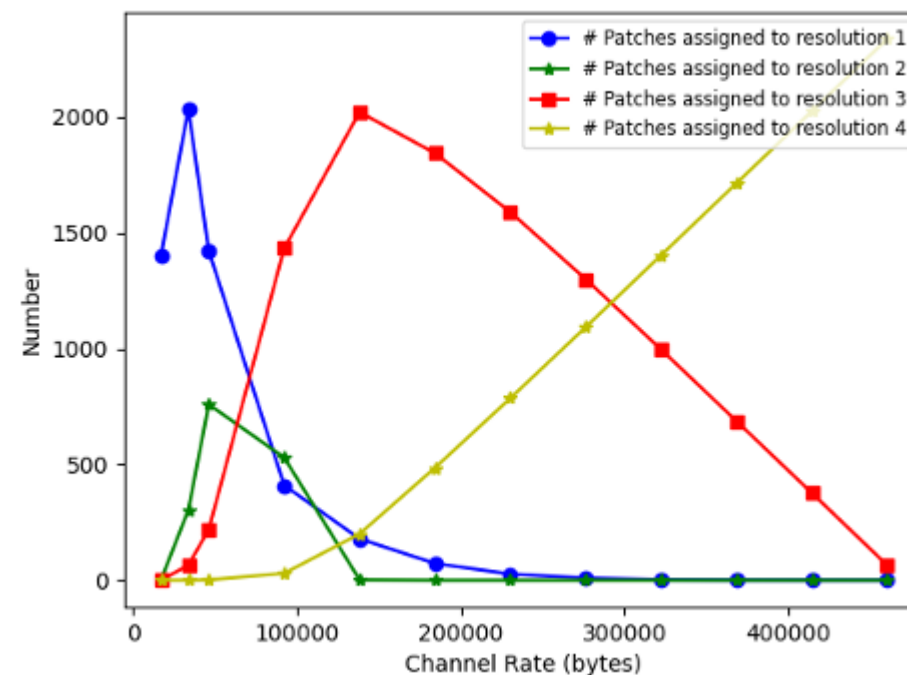


<Accuracy result for three medium resolutions.>

Results



<Accuracy result for adaptive multi-resolution semantic communication framework in various channel rates.>



<Resolution assignment to patches in different channel rate.>

Conclusion

Conclusion

- This work proposes a novel semantic communication framework that fuses open-vocabulary vision–language segmentation with transformer-based compression.
- We build on recent advances in open-vocabulary segmentation, which leverage large pre-trained vision–language models to break away from fixed label sets and segment arbitrary categories described by text prompts.
- But merging these models with the same scale can be a challenge.

Conclusion

- Potential Directions:
 - Extend to other multimodal data (e.g., eye-tracking and images together).
 - Extend to video data with recent techniques in computer vision fields
 - Develop task-specific optimization techniques for diverse domains (e.g., image anomaly detection).
- Key Challenges:
 - Efficient training of adaptive encoders.
 - Addressing latency in dynamic channel conditions.