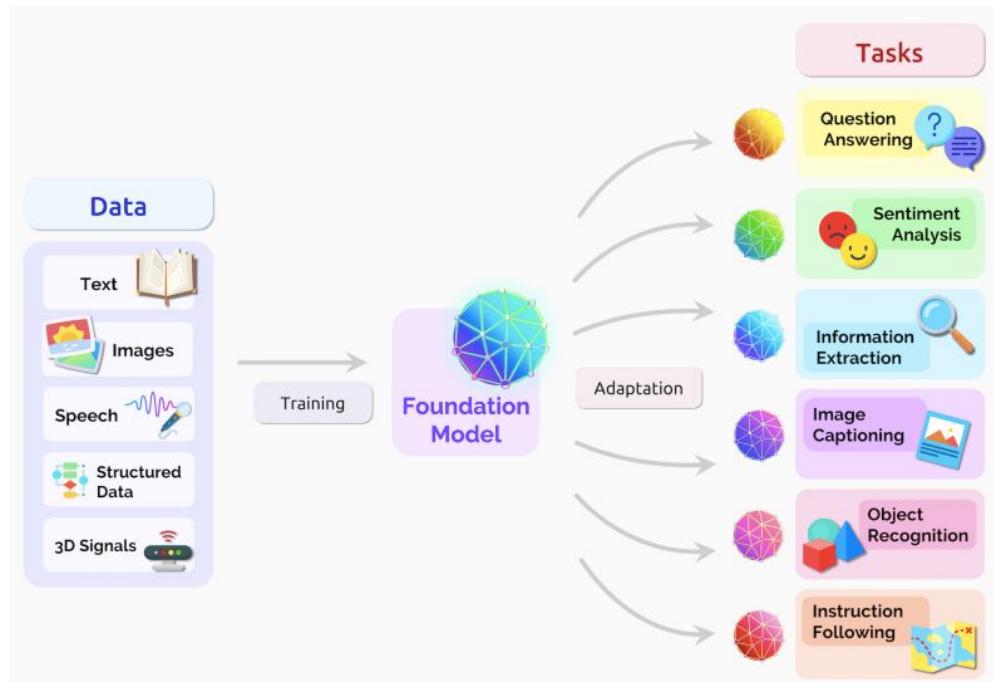


Foundation models and Segmentation

Luis Cossio

Master in Engineering sciences, mention
in Electrical Engineering

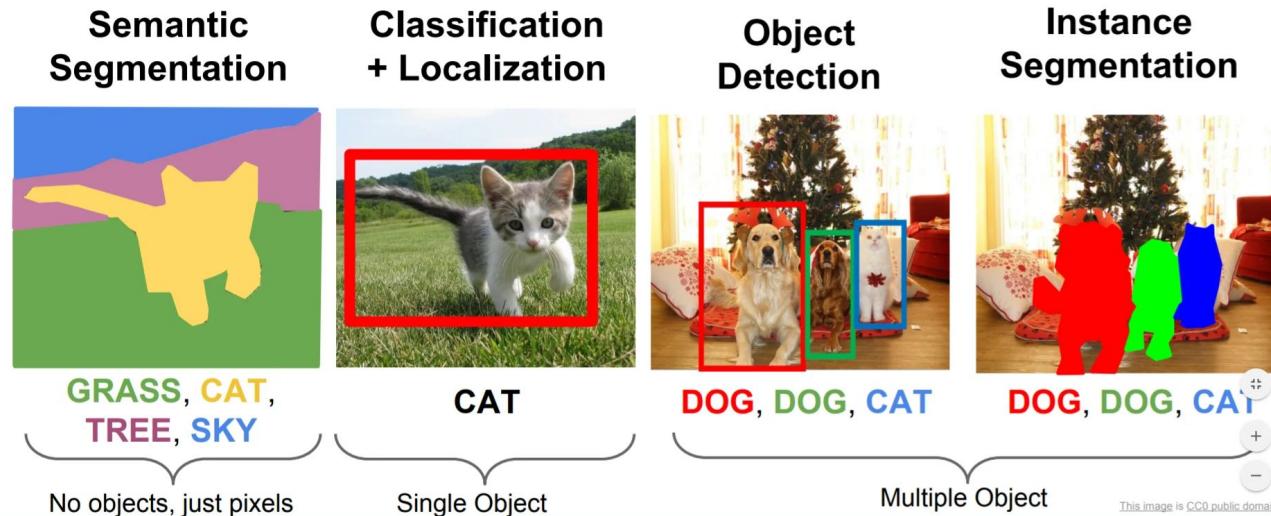
luis.cossio@uoh.cl



Segmentación de objetos

- Different type of tasks in image processing:

- Object detection
- Instance Segmentation
- Semantic Segmentation



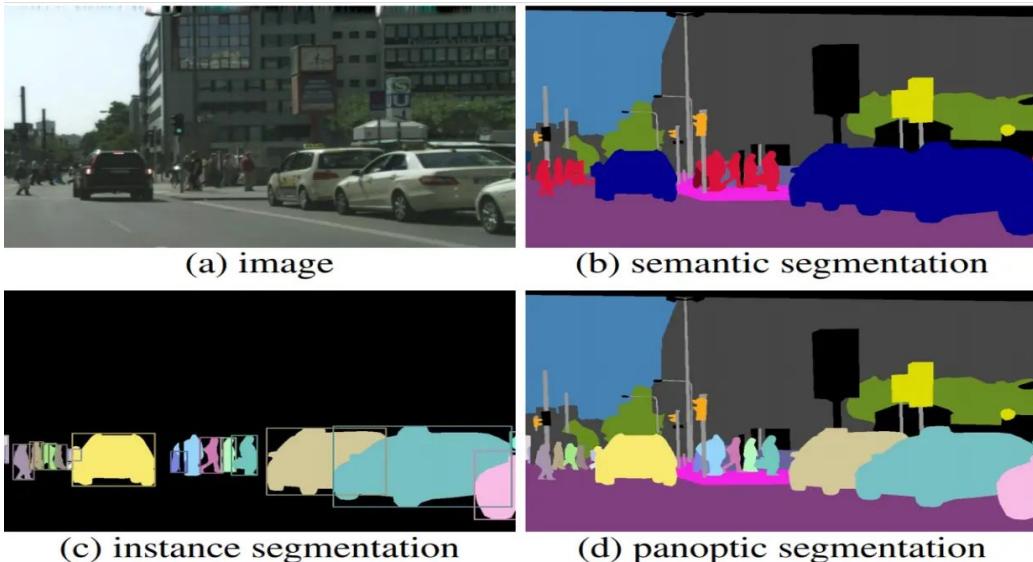
Segmentación de objetos

- Different type of tasks in image processing:
 - Object detection
 - Instance Segmentation
 - Semantic Segmentation
- Segmentation
 - Define in a pixel wise manner the class of instances or regions



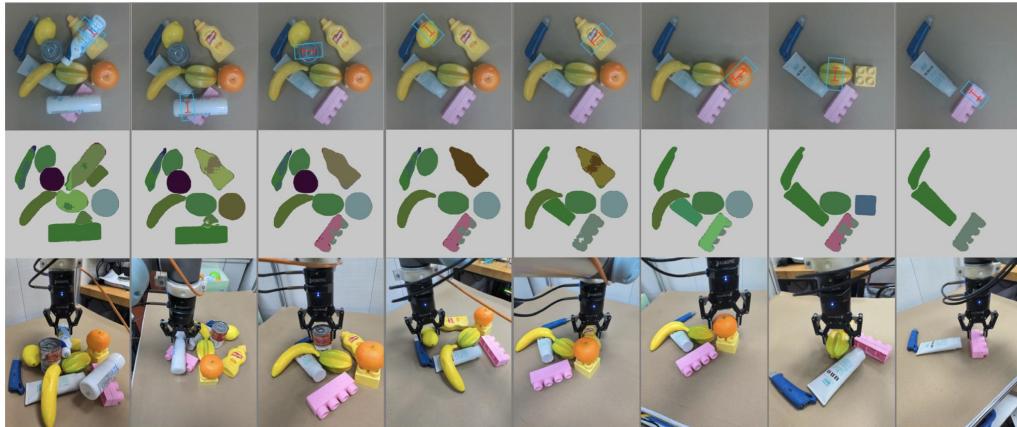
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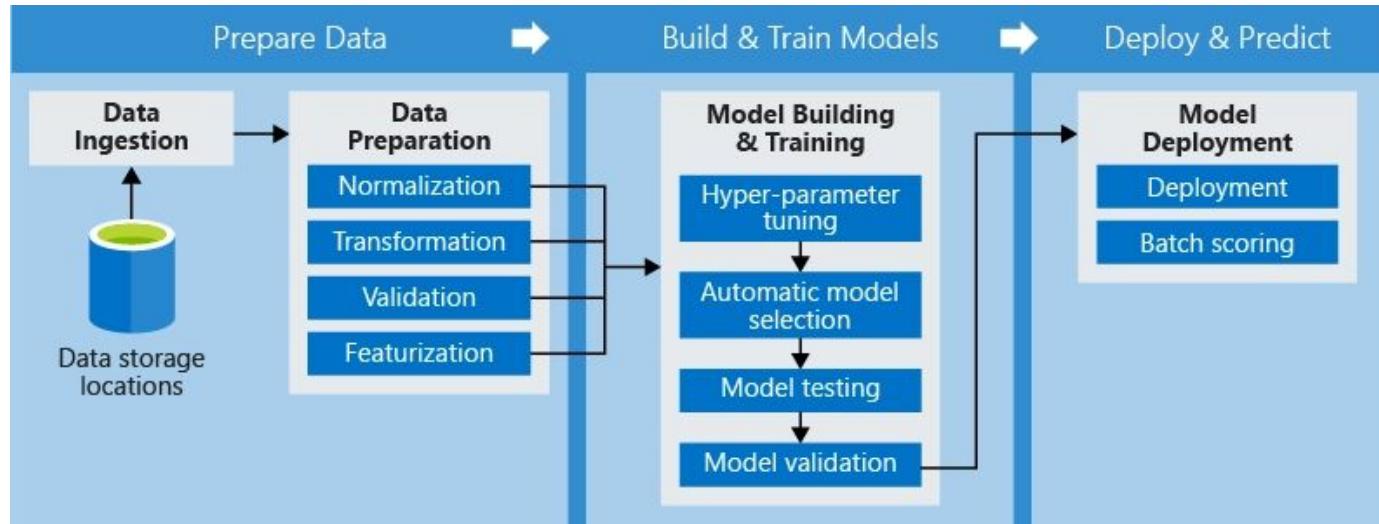
Segmentación de objetos

- Different type of tasks in image processing:
 - Object detection
 - Instance Segmentation
 - Semantic Segmentation
- Segmentation
 - Define in a pixel wise manner the class of instances or regions
 - Detail information useful for several applications



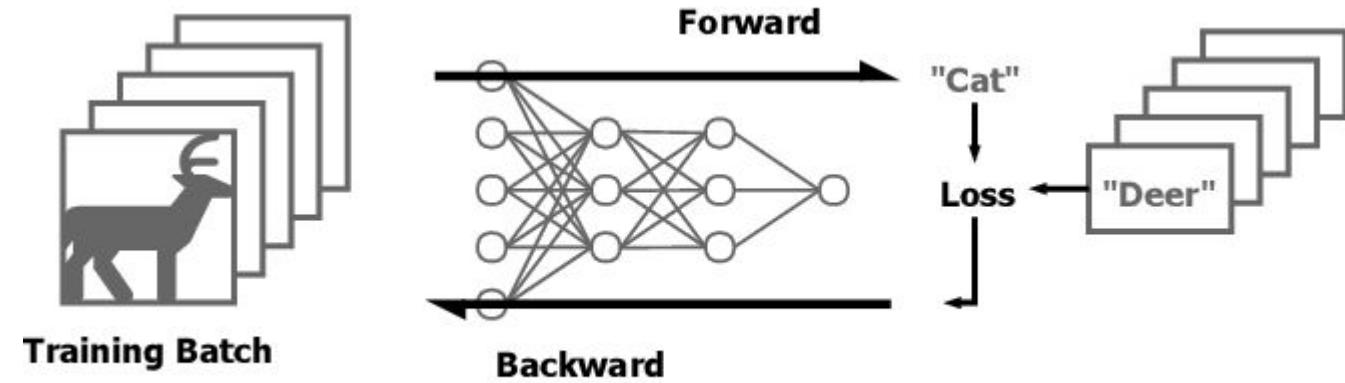
Pre-training

- Pipeline of DNN models
 - Data collection
 - Training
 - Deployment



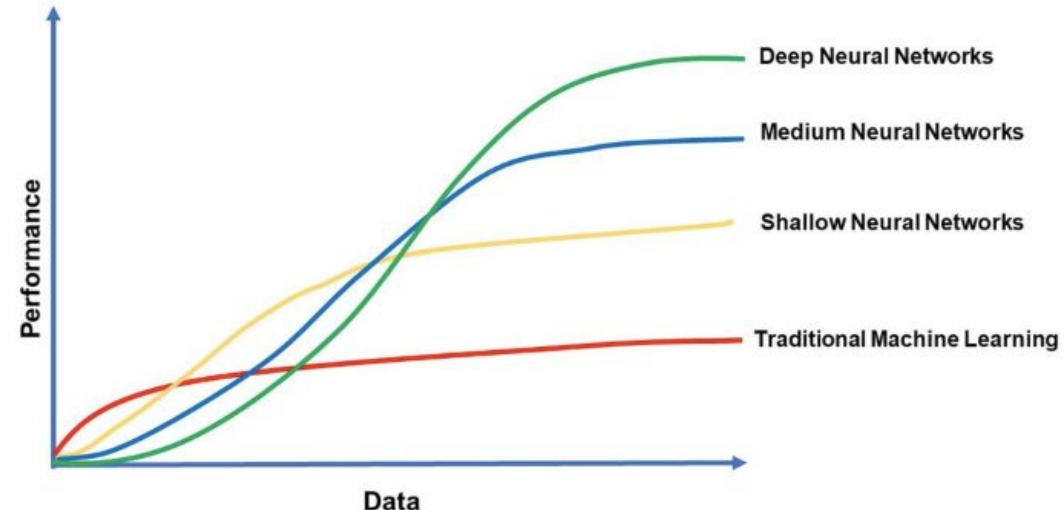
UH Pre-training

- Pipeline of DNN models
 - Data collection
 - Training
 - Deployment
- Training
 - Iterative process of improvement



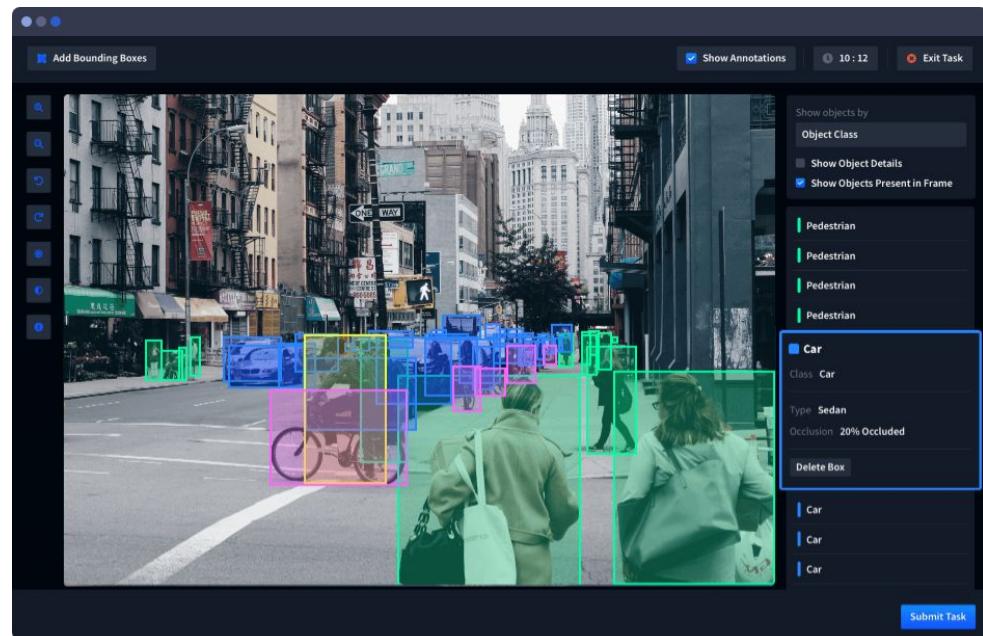
UH Pre-training

- Pipeline of DNN models
 - Data collection
 - Training
 - Deployment
- Training
 - Iterative process of improvement
 - Models perform better with more data



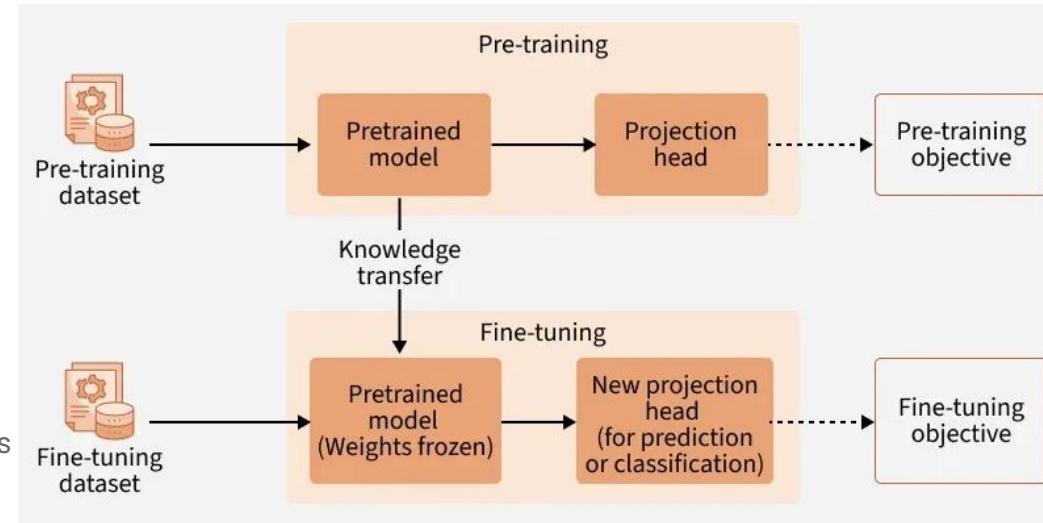
Pre-training

- Pipeline of DNN models
 - Data collection
 - Training
 - Deployment
- Training
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 - Models perform better with more data
 - Labeling is complicated
 - Several minutes per image (3-100)



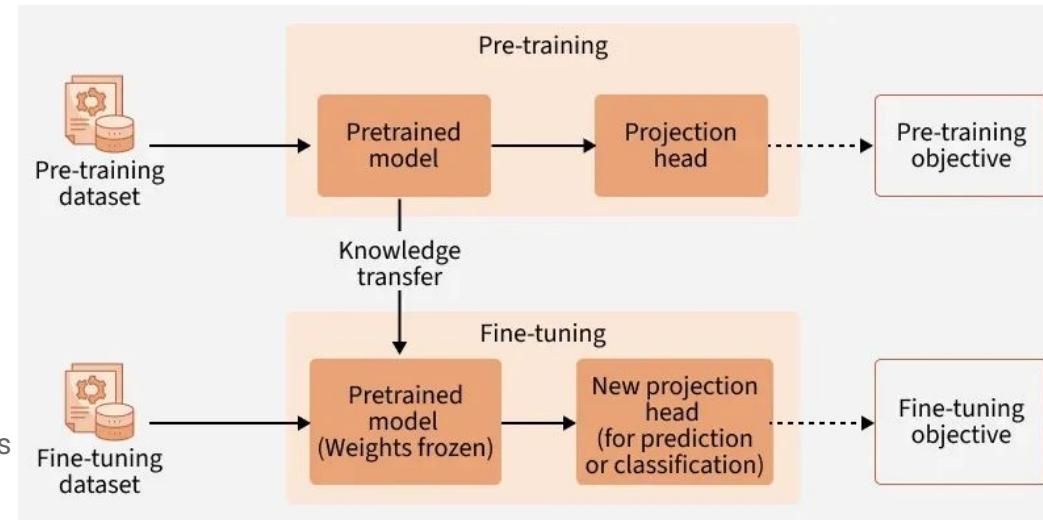
UH Pre-training

- Pipeline of DNN models
 - Data collection
 - Training
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- Training
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- Pre-training
 - Uso de modelos con pesos pre-entrenados para una tarea similar.
 - Entrenamiento secundario con una base de datos para una tarea específica



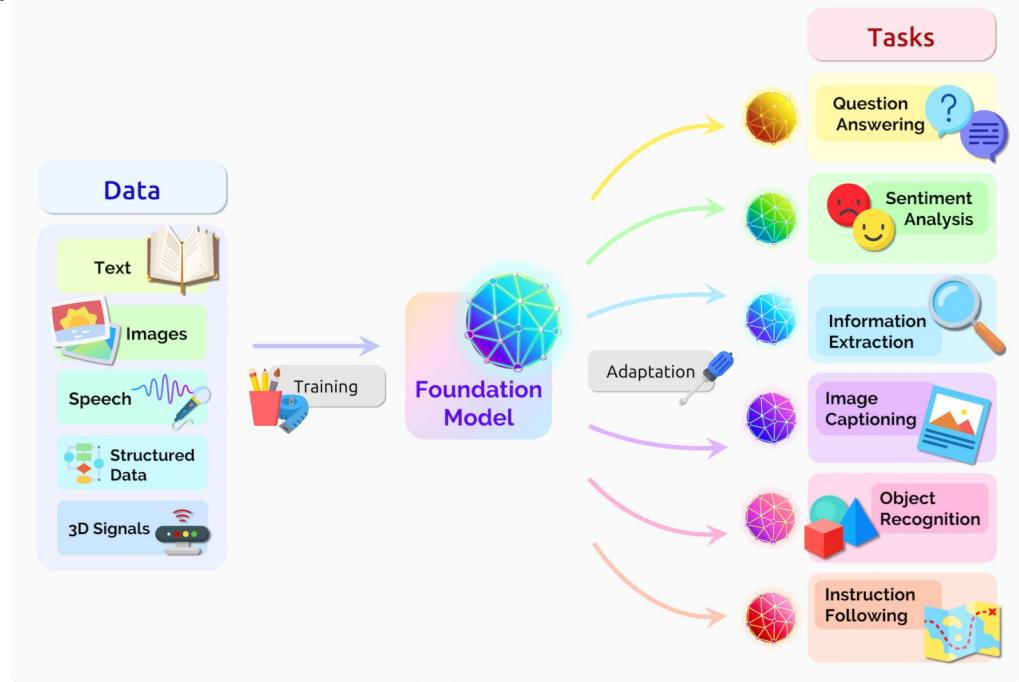
UOH Pre-training

- Pipeline of DNN models
 - Data collection
 - Training
 - Deployment
- Training
 - Iterative process of improvement
 - Models perform better with more data
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 - Several minutes per image (3-100)
- Pre-training
 - Uso de modelos con pesos pre-entrenados para una tarea similar.
 - Entrenamiento secundario con una base de datos para una tarea específica
 - Diferencia de tamaño relevante entre bases de datos
 - ImageNet 1.000.000
 - Fine tuned dataset: 1.000



Foundation Models

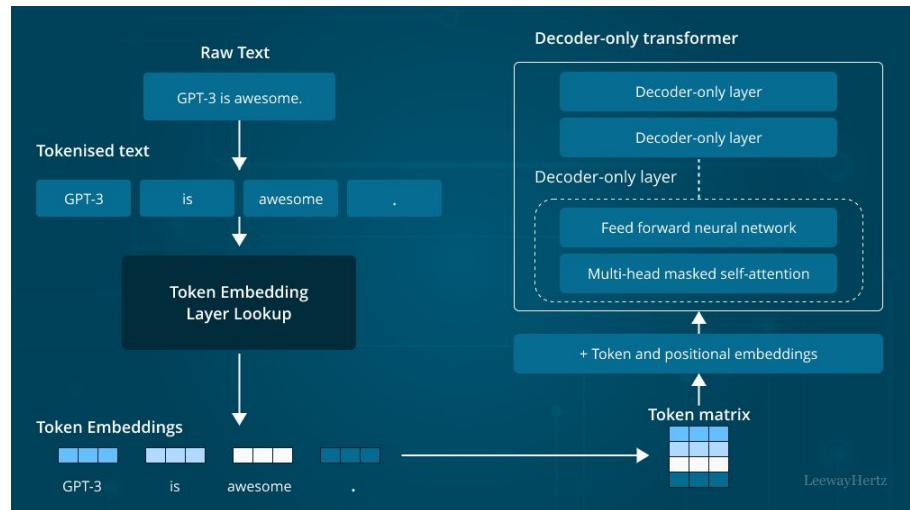
- Models pre-trained in massive amounts of data
 - Not necessarily labeled data



Foundation Models



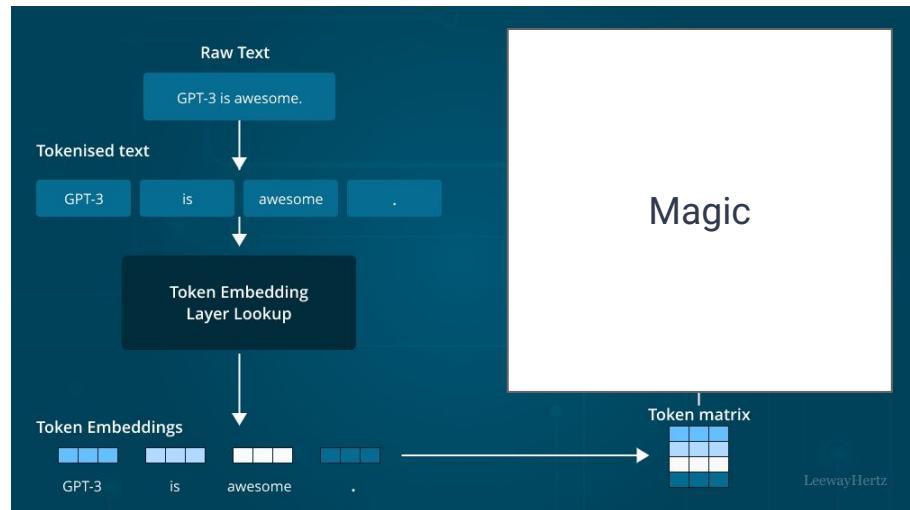
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Foundation Models



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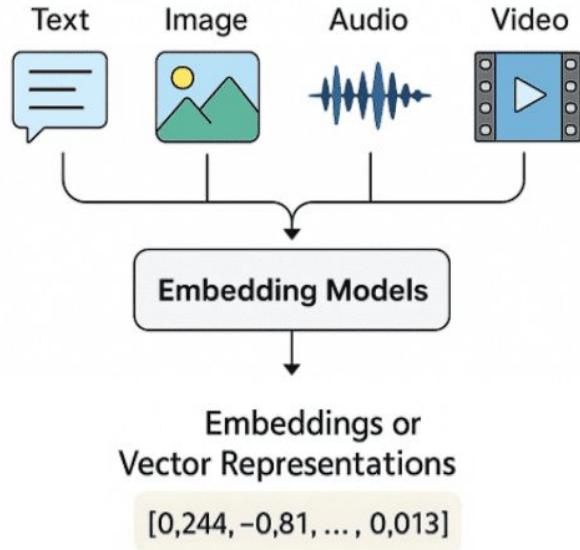


Foundation Models

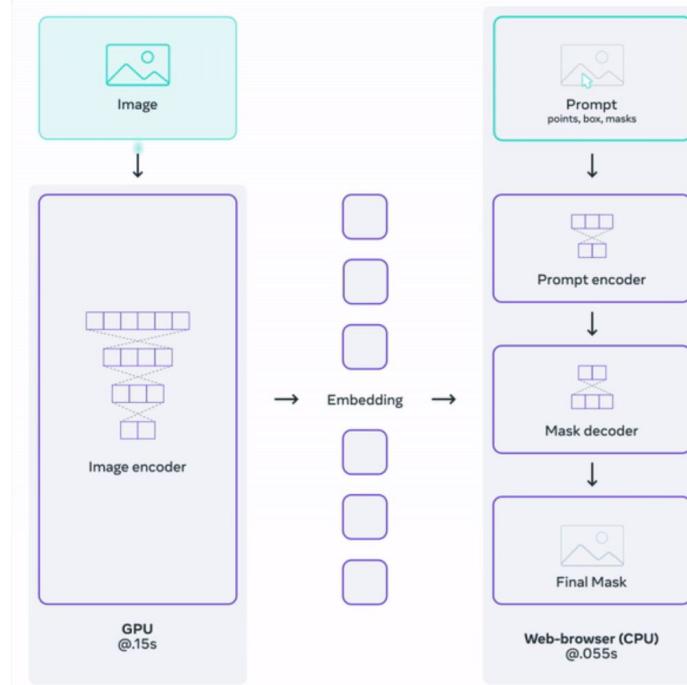


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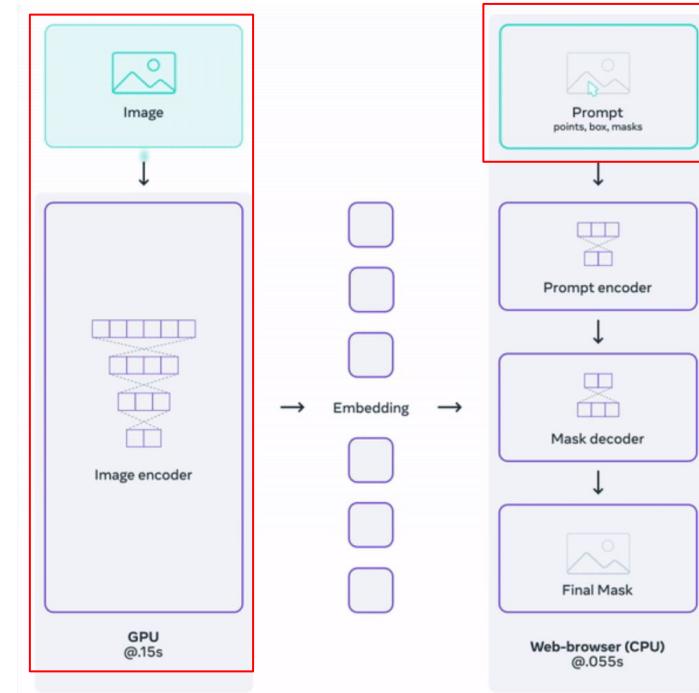
Embedding Models



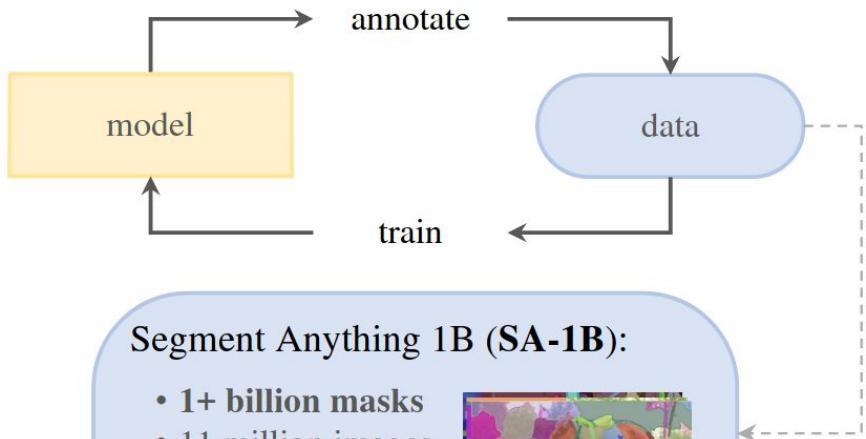
- Segment Anything model



- Segment Anything model
 - Combina la información de una imagen y algun prompt



- Segment Anything model
 - Combina la información de una imagen y algun prompt
 - Prompts validos
 - Puntos
 - Bounding boxes
 - Segmentos
 - Texto
 - Foundation model entrenado con miles de millones de datos
- Paper
 - <https://arxiv.org/pdf/2304.02643>



Segment Anything 1B (SA-1B):

- 1+ billion masks
 - 11 million images
 - privacy respecting
 - licensed images

