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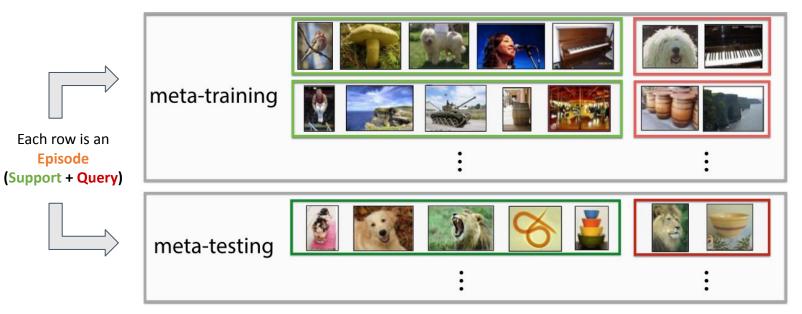
CS 6301.004 - Deep Learning For NLP

Group-11 | Spring 2023

- Few-Shot Learning is a sub-area of machine learning. It's about classifying new data when you have only a few training samples with supervised information (<u>neptune.ai</u>).
- Formulated as an N-way-K-shot problem (Episodes)
 - N := number of classes
 - K := number of samples per class
 - In a fixed setup, this remains same for all classes
 - In a variable setup, this varies across classes

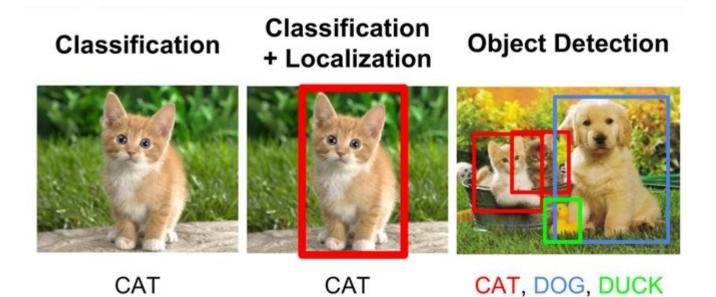
Support Set

Query Set



Here, it's a 5-way-1-shot setup (fixed episode variant)

Image: https://bair.berkeley.edu/blog/2017/07/18/learning-to-learn

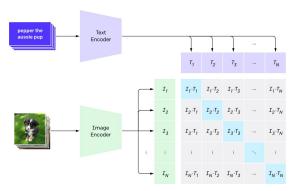


We will be dealing with **classification** only. i.e. Given an image containing a single object, classify it.

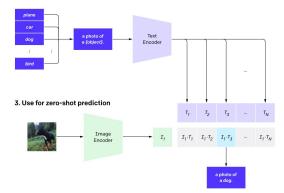
Image: https://www.kaggle.com/getting-started/169984

Related Works

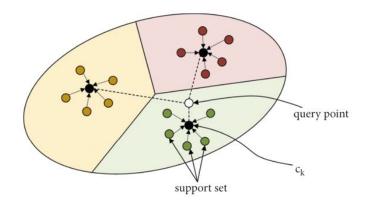
1. Contrastive pre-training

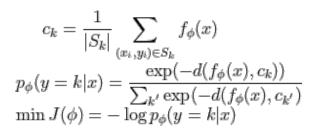


2. Create dataset classifier from label text



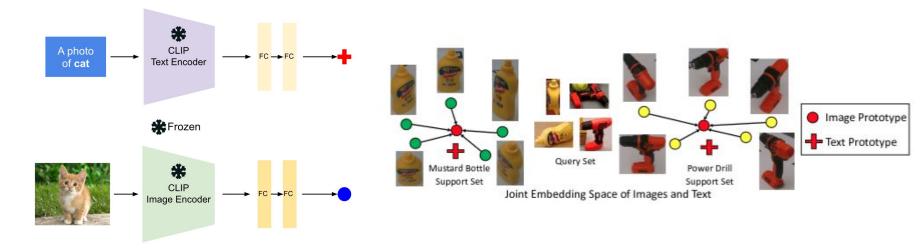
Radford, Alec, et al. "Learning transferable visual models from natural language supervision." *ICML* 2021.





Snell, Jake, Kevin Swersky, and Richard Zemel. "Prototypical networks for few-shot learning." NeurIPS 2017.

Proposed Method Proto-CLIP



$$P(y = k | \mathbf{x}^{q}, \mathcal{S}) = \alpha P(y = k | \mathbf{x}^{q}, \mathcal{S}_{x}) + (1 - \alpha) P(y = k | \mathbf{x}^{q}, \mathcal{S}_{y})$$

Loss: Negative Log Likelihood

Our proposed Proto-CLIP model learns a *joint embedding space of images and text,* where *image prototypes* and *text prototypes* are learned using *support sets* for few-shot classification.

Dataset: ImageNet | Metric: Accuracy

Questions?