

Semantic Description of Objects in Images Based on Prototype Theory

Omar Vidal Pino, Erickson R. Nascimento, Mario F. M. Campos
Universidade Federal de Minas Gerais (UFMG), Belo Horizonte, Brazil

Abstract

This research aims to build a model for the semantic description of objects based on visual features extracted from images. Inspired by the human approach used to represent categories, we proposed a novel *Computational Prototype Model* (CPM) that encodes and stores the object's image category's central semantic meaning: the semantic prototype. Also, we proposed a *Prototype-based Description Model* that uses our CPM model to describe an object's image highlighting its most distinctive features within the category. Our description method uses the proposed Prototypical Similarity Layer (PS-Layer) to

retrieve the category's semantics simulating the prototype-based concept of categorization of the Prototype Theory. Using different datasets, our experiments show that: *i*) the proposed CPM model successfully simulates the internal semantic structure of the categories; *ii*) the proposed semantic distance metric can be understood as the object typicality score within a category; *iii*) our semantic classification method based on prototypes can improve the performance and interpretation of CNN classification models; *iv*) our semantic descriptor encoding significantly outperforms other state-of-the-art image global encodings in clustering and classification tasks.

Introduction

Motivation

Although CNNs provide outstanding performance in image semantic processing tasks:

- CNN description models do not codify their image's representation based on the Cognitive Science foundations to represent the meaning.
- Is still unknown how to make an accurate interpretation of CNN classification models outputs.

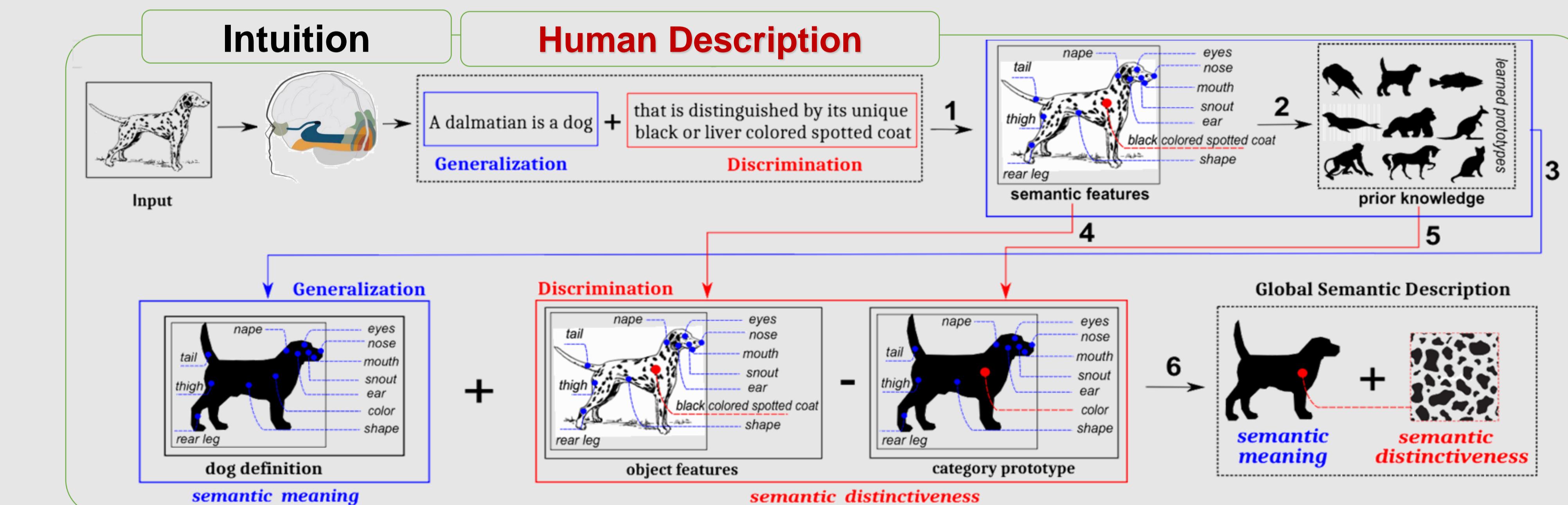
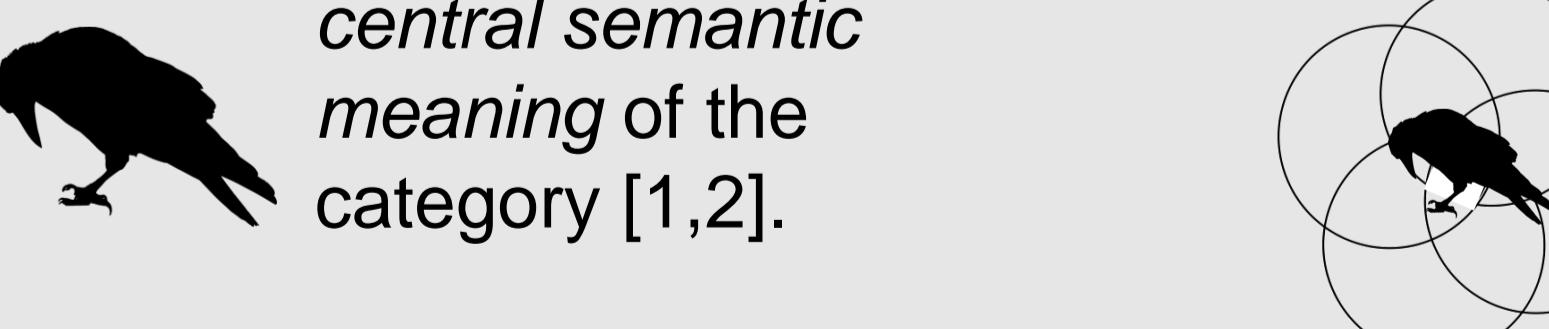
As an attempt to overcome these, we bring into view the Prototype Theory as a framework to represent the semantic meaning of the visual information contained in an image.

Background

Prototype Theory Foundations

- The prototype stands for the *central semantic meaning* of the category [1,2].
- Object categorization is made based on how similar a new exemplar is with the learned prototypes [3].

- Category internal semantic structure* [1,2]: category members are positioned around of the category prototype based on its degrees of typicality.



- Main characteristics (*prototypicality effects*) [1-5]:

non-equality
(salience effect, core/periphery)

non-discreteness
(demarcation problems, flexibly)

extensional

Difference of typicality and membership salience

intensional

Clustering into family resemblances

Fuzziness at the edges, membership uncertainty

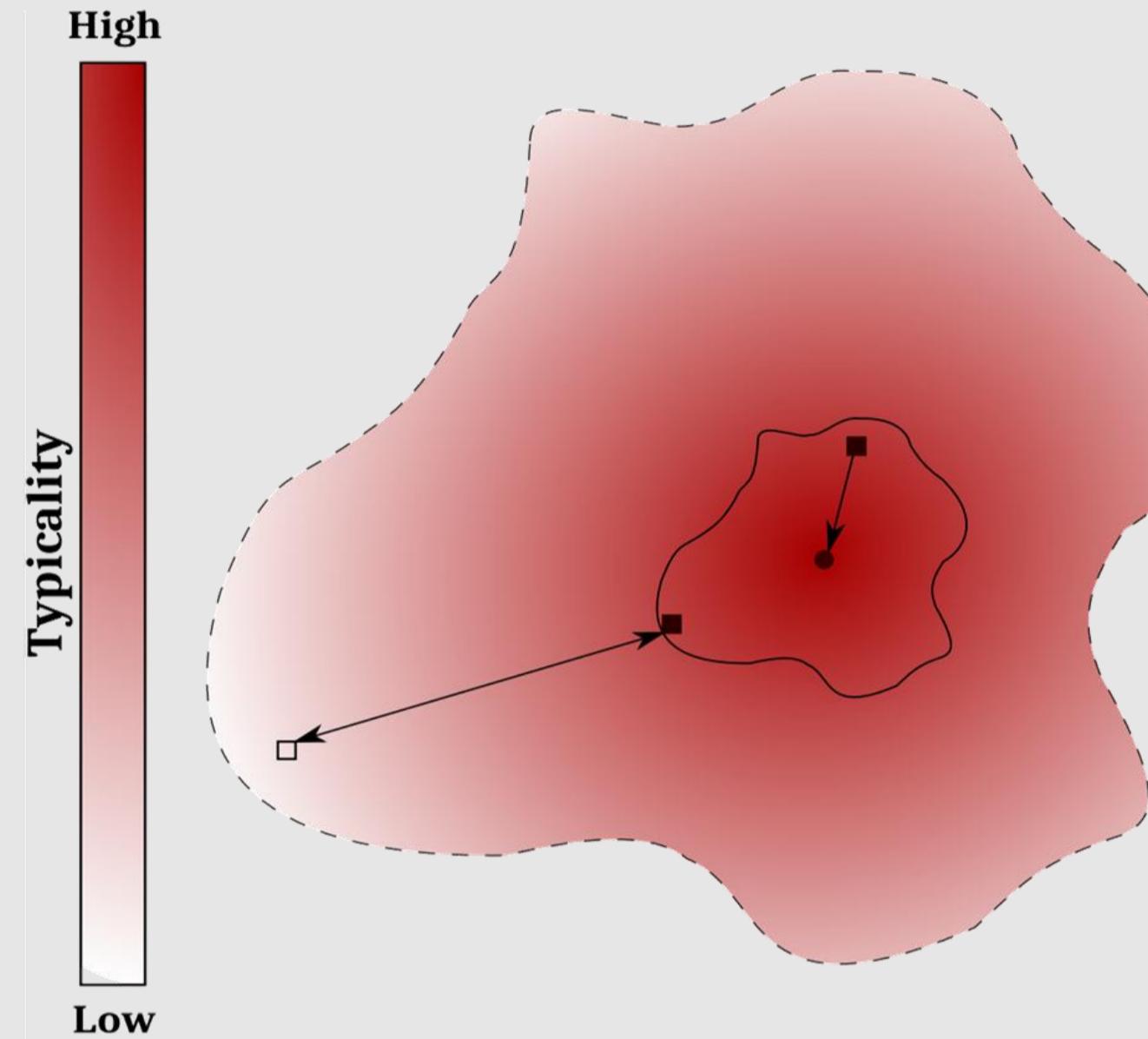
Absence of necessary-and-sufficient definitions

Semantic Description of Objects in Images Based on Prototype Theory

Omar Vidal Pino, Erickson R. Nascimento, Mario F. M. Campos
Universidade Federal de Minas Gerais (UFMG), Belo Horizonte, Brazil

Methodology

Computational Prototype Model (CPM)



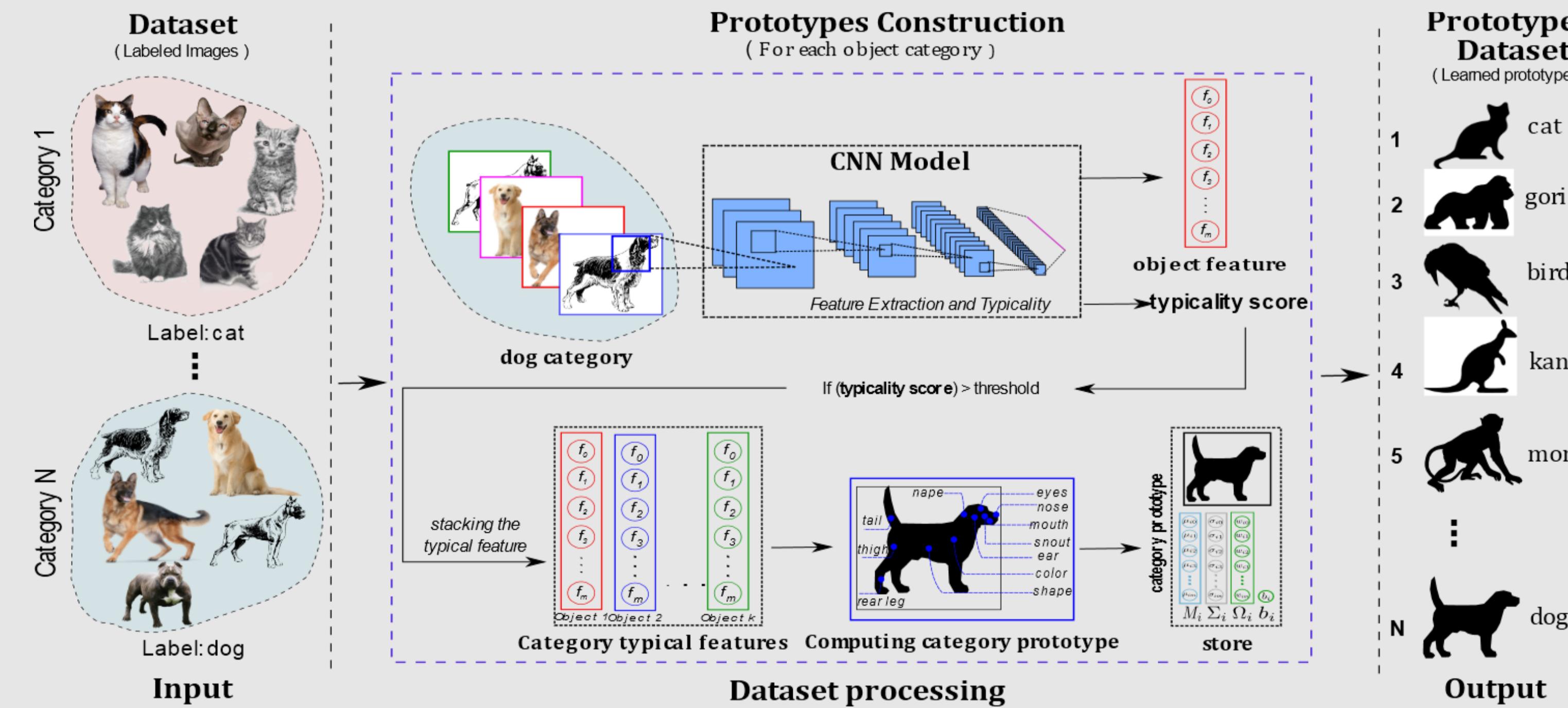
Legend

- Typical members
- Non-typical members
- Semantic Prototype
- Abstract Prototype
- Category edges (*undefined*)
- Semantic Prototype edges (*defined*)
- ↔ Prototypical distance
- ↔ Distance between objects

Category internal structure.

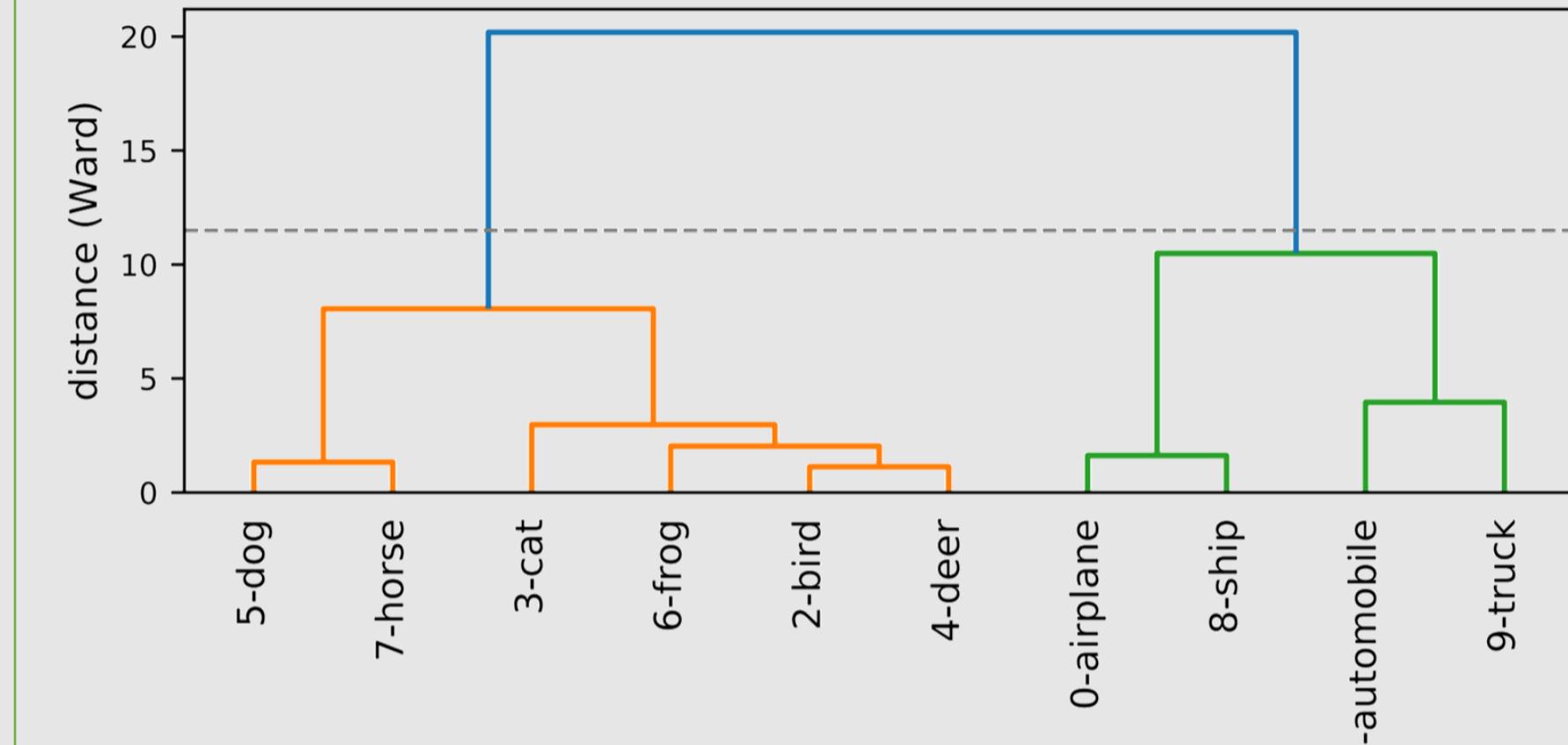
The figure shows our expected semantic representation of the category internal structure. We show the key definitions, the advantages and the constraints of our CPM Model.

Off-line construction of the semantic prototypes dataset



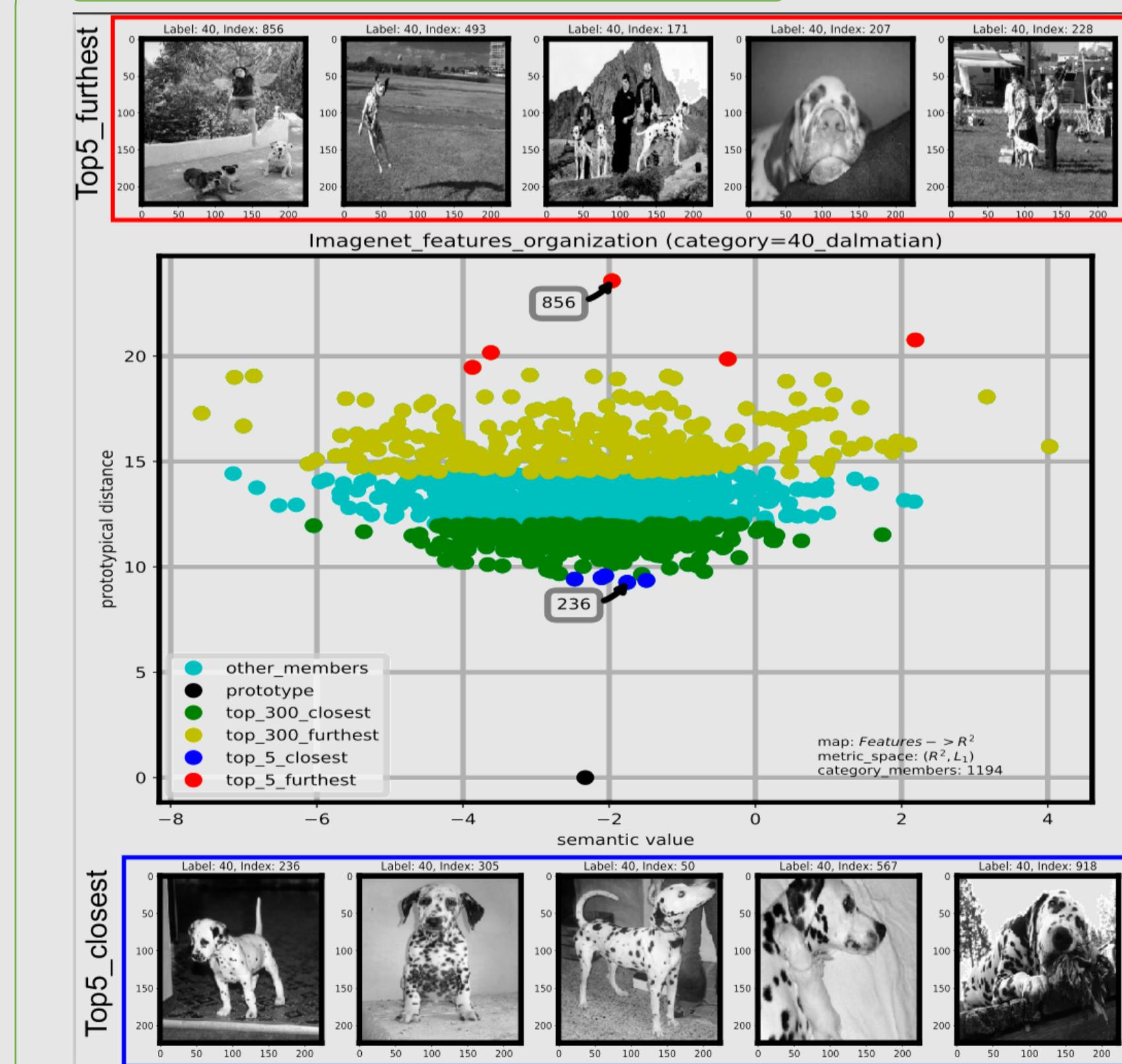
Results

Semantic prototype encoding



Hierarchical clustering of CIFAR10 semantic prototypes.

Prototypical organization

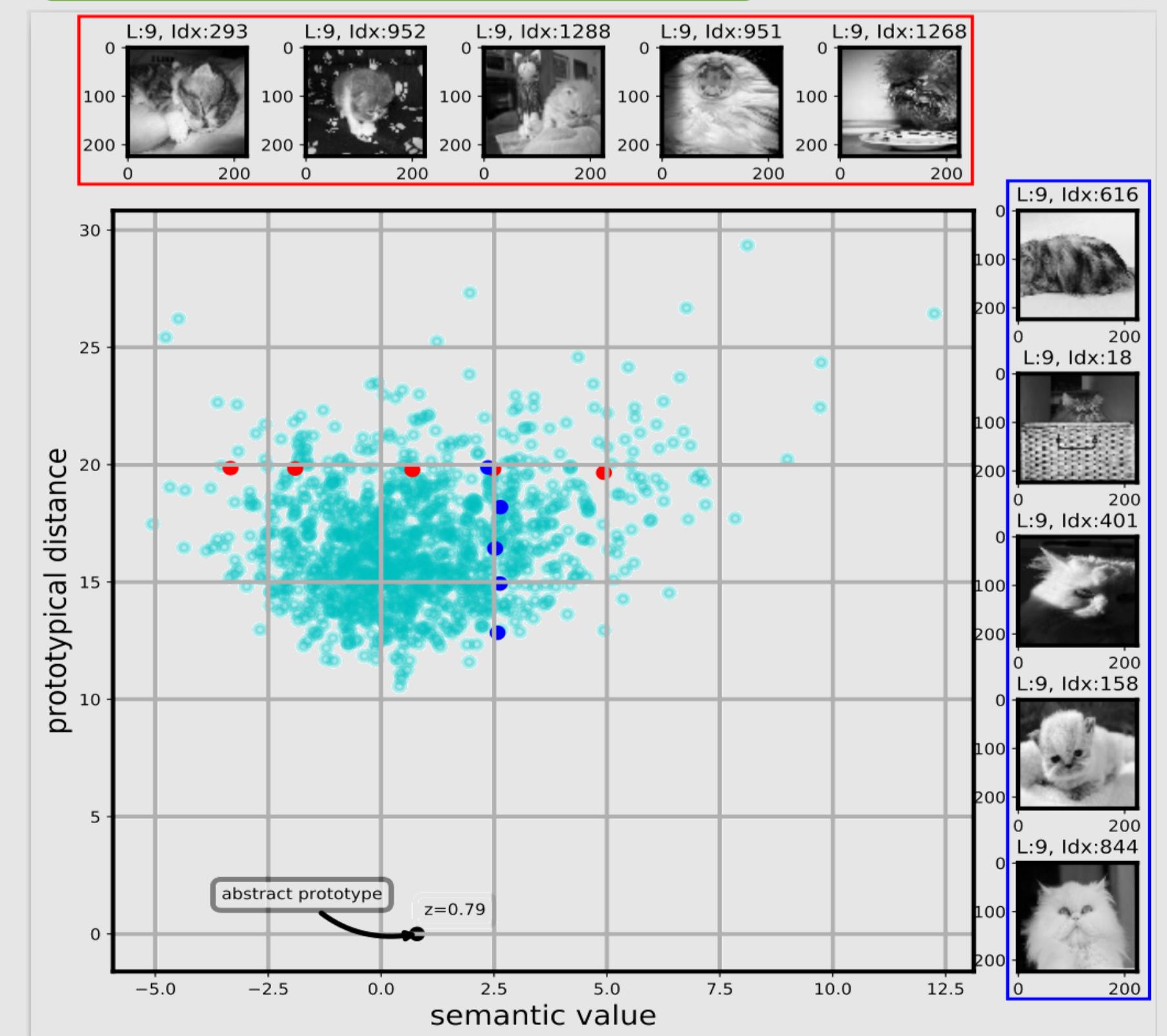


Dalmatian category of ImageNet dataset.

Central - Peripheral meaning



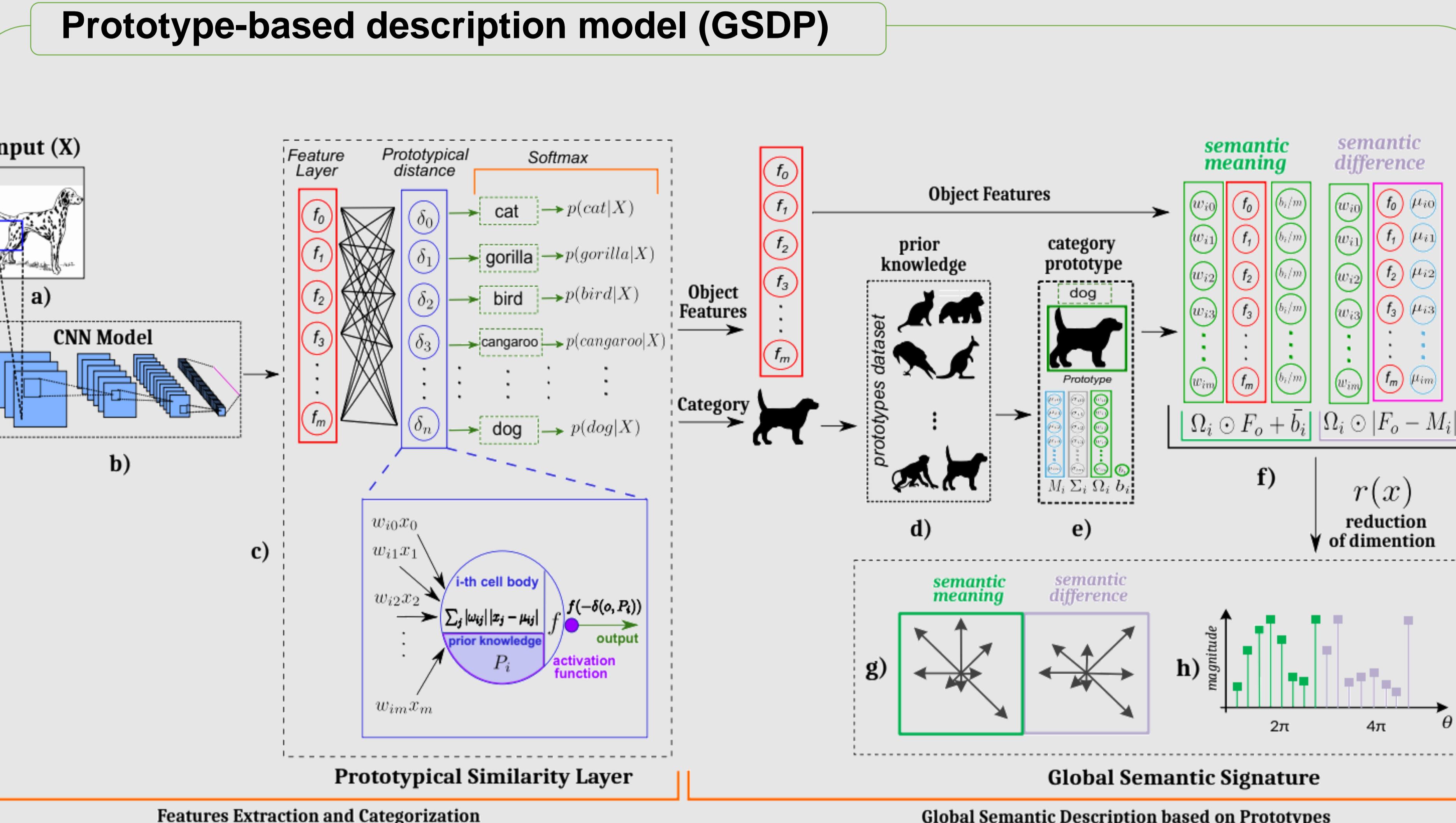
Image Typicality Score [6]



Semantic Description of Objects in Images Based on Prototype Theory

Omar Vidal Pino, Erickson R. Nascimento, Mario F. M. Campos
Universidade Federal de Minas Gerais (UFMG), Belo Horizonte, Brazil

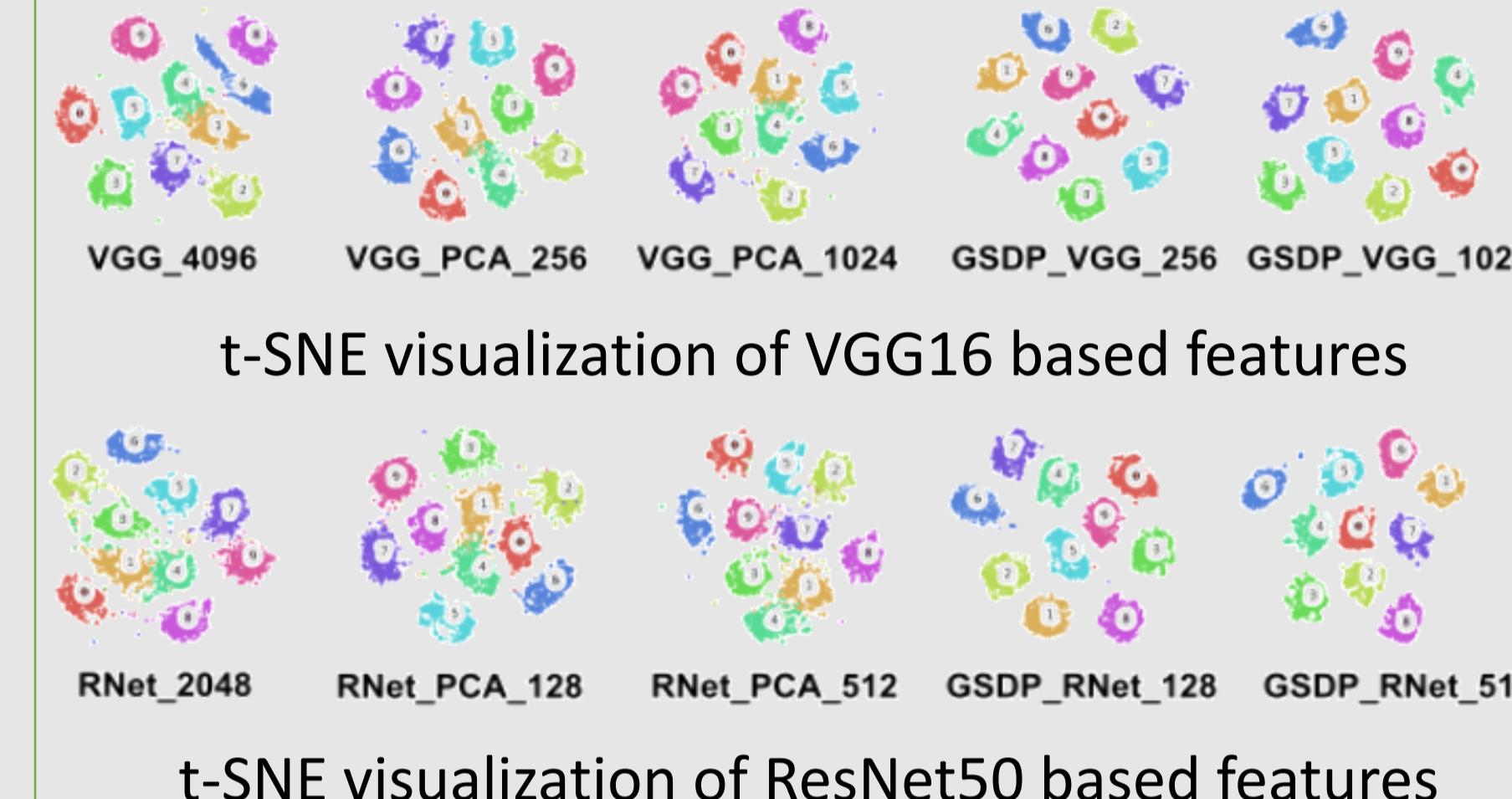
Methodology



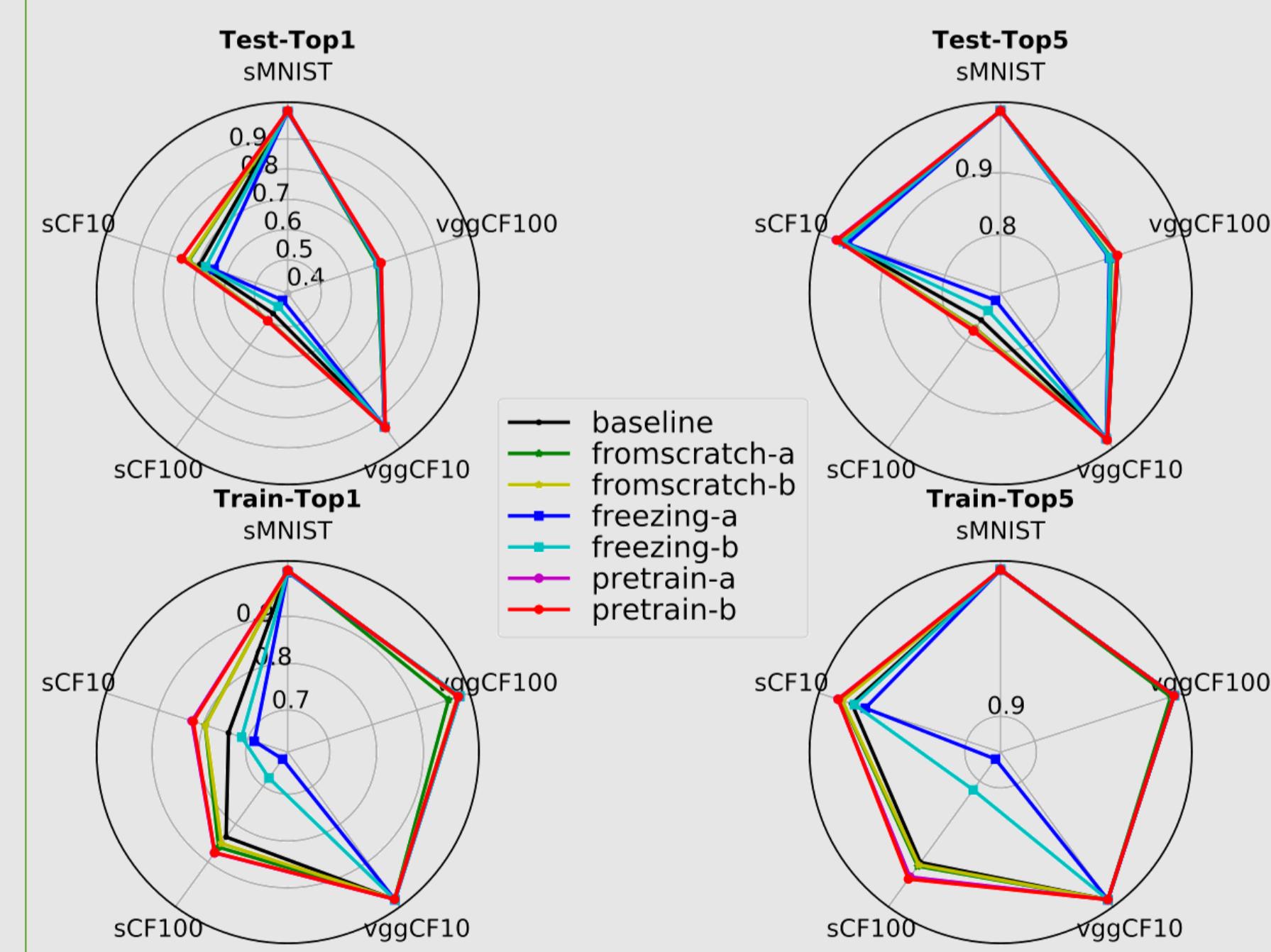
Set of steps to transform the visual information received as input into a Global Semantic Descriptor signature. The methodology workflow can be divided into two main stages: **1) Feature extraction and Categorization** and **2) transformation of CNN-object features into our Global Semantic Signature**. a) input image; b)-c) features extraction and classification using a pre-trained CNN-classification model. Our Prototypical Similarity Layer (PS-Layer) is used to convert a common CNN-model into a prototype-based CNN-classification model; d) prototype dataset; e) category prototype selection; f) global semantic description of object using category prototype; g) graphic representation of our descriptor signature resulting from the dimensionality reduction function ($r(x)$); and h) Global Semantic Signature.

Results

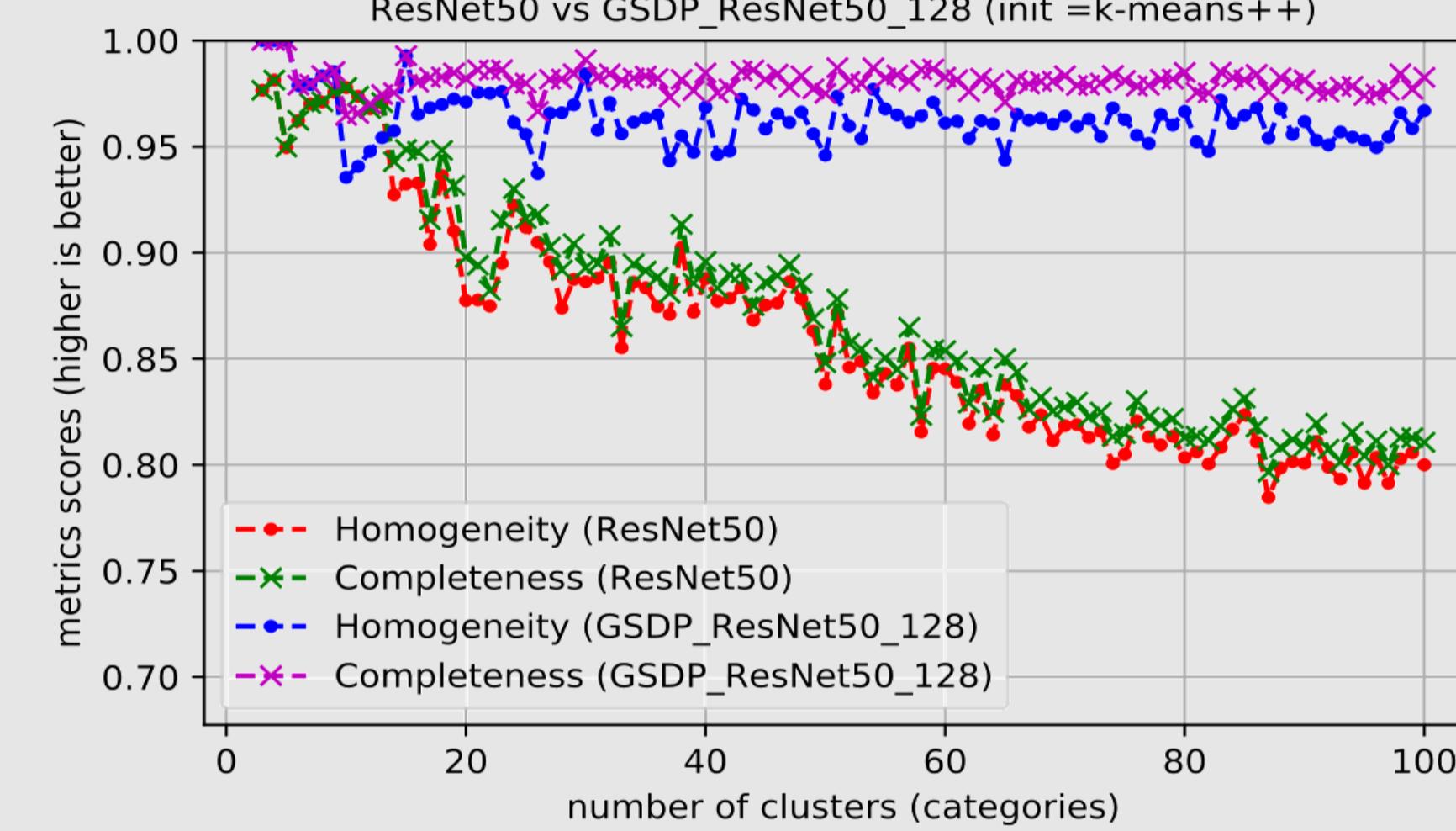
GSDP - Signature Information Analysis



Prototype-based Semantic Classification (PS-Layer Performance Summary)



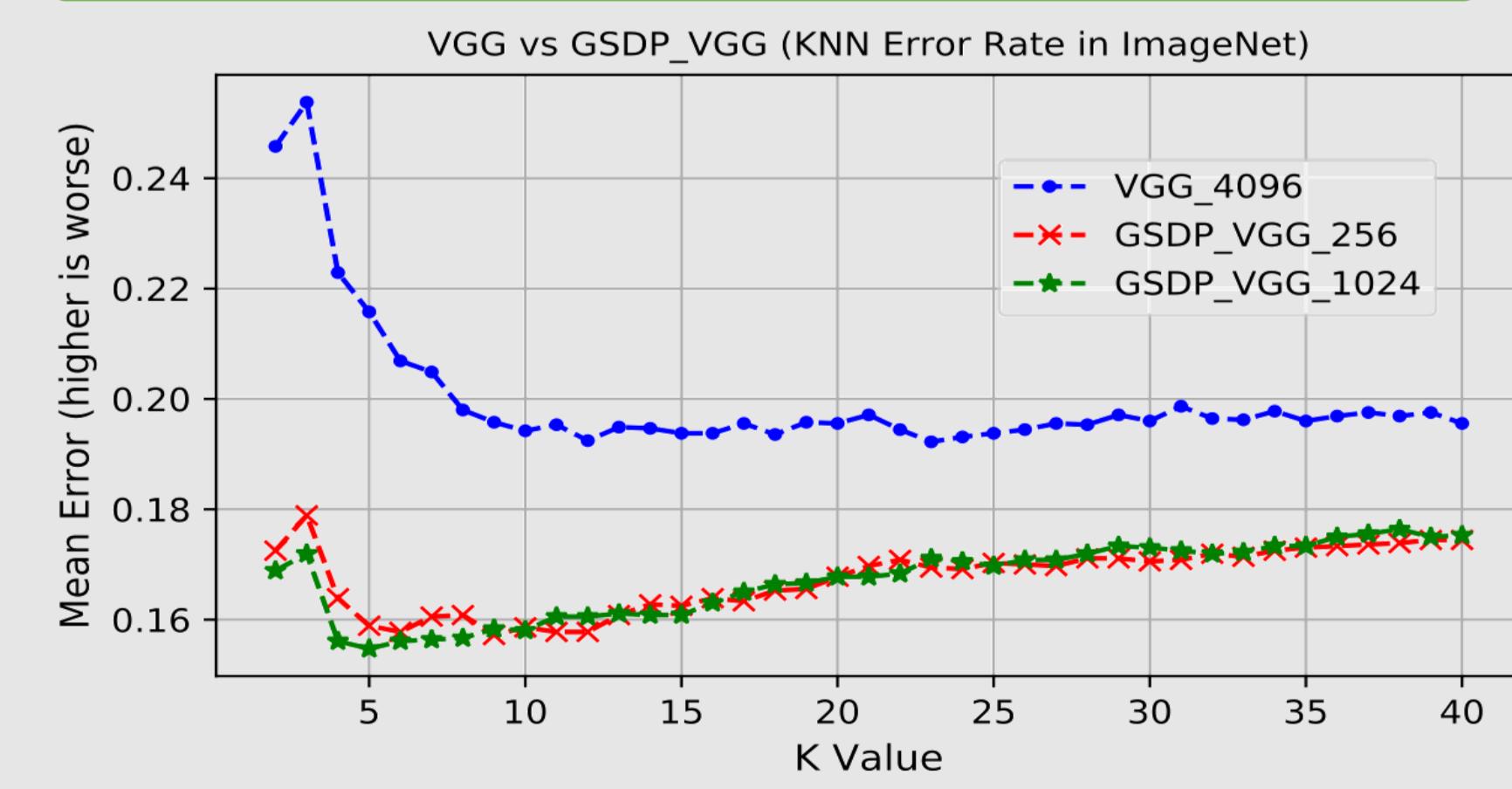
Cluster Metrics Score History [7] (ImageNet Dataset)



Cross-Dataset Cluster Metrics Score (Coco Dataset)

Descriptor	Size	FPS	Metrics Scores				
			H	C	V	ARI	AMI
Deep Features Performance on Coco [38](CrossDataset)							
ResNet50 [6]	2048	10.6	0.29	0.36	0.32	0.17	0.31
ResNet50_PCA_128	128	12.5	0.32	0.34	0.33	0.17	0.31
ResNet50_PCA_512	512	12.5	0.34	0.35	0.34	0.20	0.33
GSDP_RNet_128 (our)	128	9.6	0.43	0.69	0.53	0.16	0.52
GSDP_RNet_512 (our)	512	9	0.34	0.47	0.40	0.09	0.39

Classification Error Rate (ImageNet Dataset)



Semantic Description of Objects in Images Based on Prototype Theory

Omar Vidal Pino, Erickson R. Nascimento, Mario F. M. Campos
 Universidade Federal de Minas Gerais (UFMG), Belo Horizonte, Brazil

Conclusion

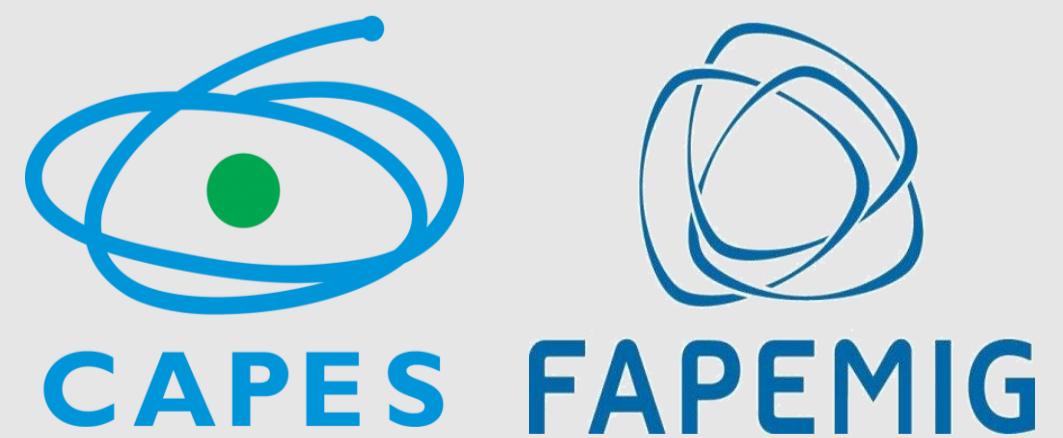
In this Ph.D. dissertation we introduced and evaluated three models based on the foundations of Prototype Theory to propose semantic representations of object's categories and object's images:

- i. A Computational Prototype Model (CPM),
- ii. A novel Prototype-based Description Model (GSDP) and,
- iii. A Prototype-based Classification Model (PS-Layer).

Experiments showed that our CPM model can capture the object's visual typicality and the central and peripheral meaning of objects' categories. Our novel GSDP representation introduces a new approach to the semantic description of object's images; and experiments in large image dataset shows that it is discriminative, small dimensioned and encodes the semantic information of category members. Our PS-Layer introduces the image typicality property in semantic category learning process and experiments conducted showed that it can outperform some CNN-Models architecture.

Our prototype-based description model proposes a starting point to introduce the theoretical foundation related to *the representation of semantic meaning and the learning of visual concepts* of the Prototype Theory in the CNN-Descriptors family.

Acknowledgment



References

Publications & Awards

- Part of this work was published in:
 - 2019 IEEE Winter Conference on Applications of Computer Vision (WACV 2019)[8].
- An extension journal paper is under review in the Transactions on Image Processing (TIP).
- The Ph.D. dissertation of this work was also selected to be presented at the WACV 2019 Doctoral Consortium.

References

1. E. Rosch. Cognitive representations of semantic categories. *Journal of Experimental Psychology*, 1975.
2. E. Rosch and C. B. Mervis. Family resemblances: Studies in the internal structure of categories. *Cognitive psychology*, 1975.
3. Rosch, Eleanor. Principles of categorization. *Cognition and categorization*, 1978.
4. D. Geeraerts. Diachronic prototype semantics: A contribution to historical lexicology, 1997.
5. D. Geeraerts. Theories of lexical semantics. Oxford University Press, 2010.
6. B. Lake, W. Zaremba, R. Fergus, and T. Gureckis, “Deep neural networks predict category typicality ratings for images,” in Proceedings of the 37th Annual Conference of the Cognitive Science Society. Cognitive Science Society, 2015.
7. J. Yang, D. Parikh, and D. Batra. Joint unsupervised learning of deep representations and image clusters. CVPR, 2016.
8. O. Pino, E. Nascimento, and M. Campos, Prototypicality effects in global semantic description of objects, in Proceedings of the IEEE Winter Conference on Applications of Computer Vision (WACV), Jan 2019, pp. 1233–1242.