

Explainable Image Segmentation with Prototypes

class- and scale-independence

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The problem of semantic image segmentation

- Safety-critical application areas:
 - self-driving cars,
 - aviation,
 - medical imaging.
- Deep learning models:
 - achieve high accuracy,
 - lack explanations (“black-box” nature).
- Need for interpretable models:
 - reliability and transparency,
 - understanding the model’s behavior.

The problem of semantic image segmentation

CityScapes dataset [3]

- standard benchmark dataset for evaluating semantic segmentation
- contains images of urban environments with pixel-level annotations
- 30 semantic classes, including:
 - road
 - sidewalk
 - building etc.
- high-resolution images (1024×2048 pixels)



Self-explaining deep learning models

- built-in interpretability
 - through explanatory components
 - concept, prototype
- explanations understandable by humans
- combination of:
 - **intuitive reasoning** of classical ML models
 - **accuracy** of deep learning models



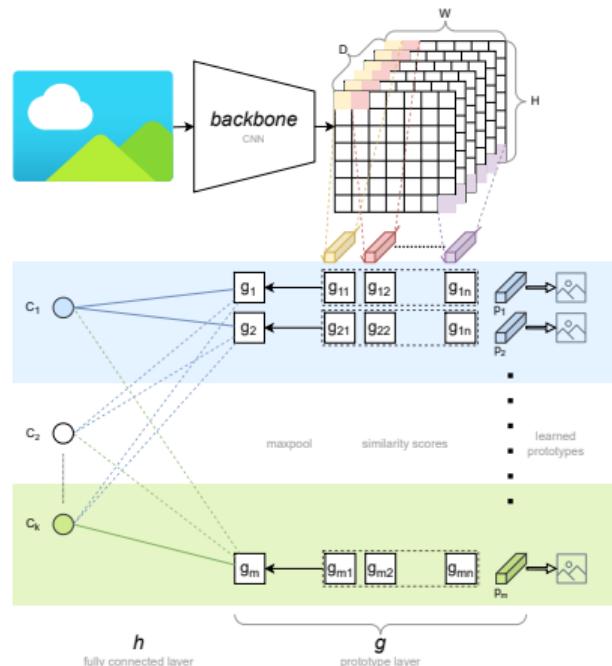
ProtoPNet [1]

strengths:

- patch-level prototypes
- “This looks like...” style reasoning

limitations:

- prototypes have learnable weights
 - prototype – patch representation
- class-specific prototypes

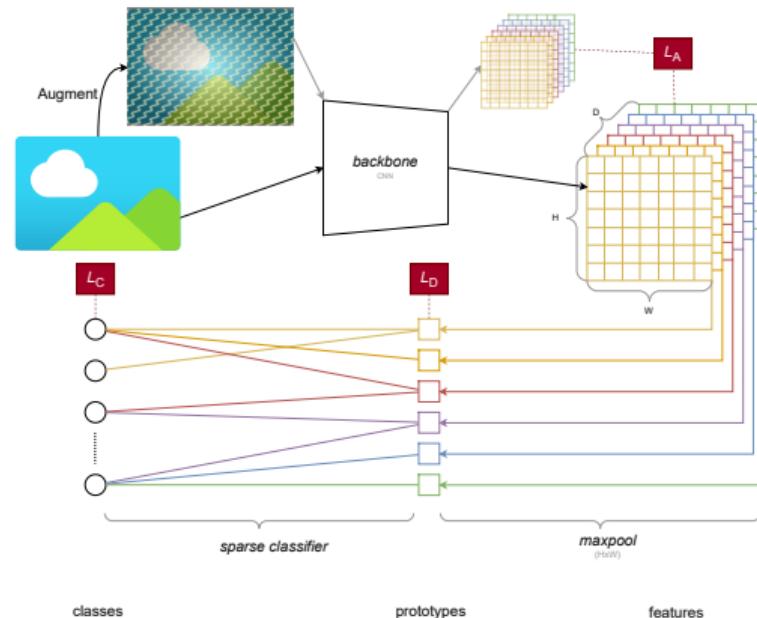




PIP-Net [5]

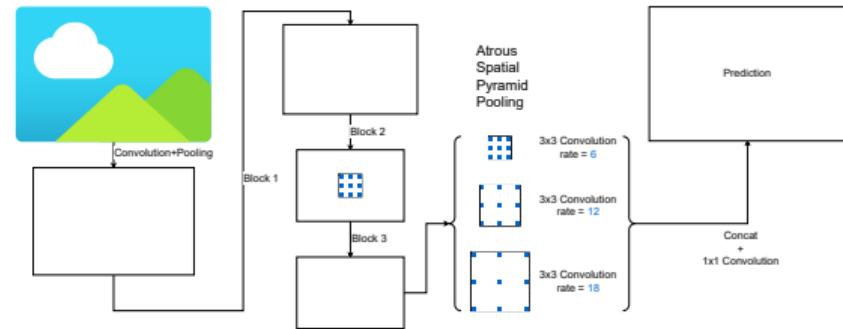
strengths:

- concept activations
 - no learned prototypes
 - unsupervised learning
- “sparse classifier”
- cross-class prototypes



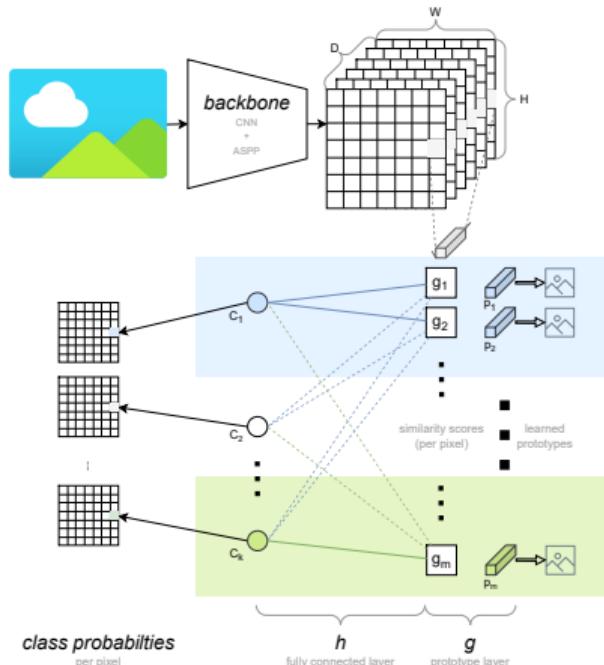
DeepLab v3 [2]

- based on the ResNet [4] architecture
- foundation for multiple prototype-based models
- Atrous Spatial Pyramid Pooling (ASPP) = multi-scale representation



ProtoSeg [7]

- similar to ProtoPNet
- based on the DeepLab v3 architecture
- prototype diversity ensured using Kullback–Leibler divergence

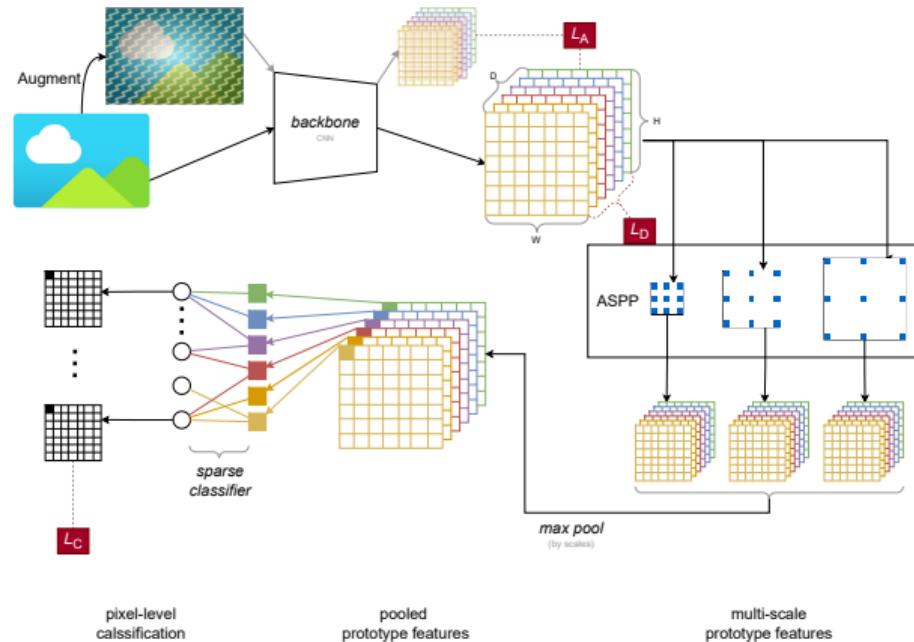


ScaleProtoSeg [6]

- prototypes at multiple scales
- leveraging the capabilities of ASPP

Proposed architecture

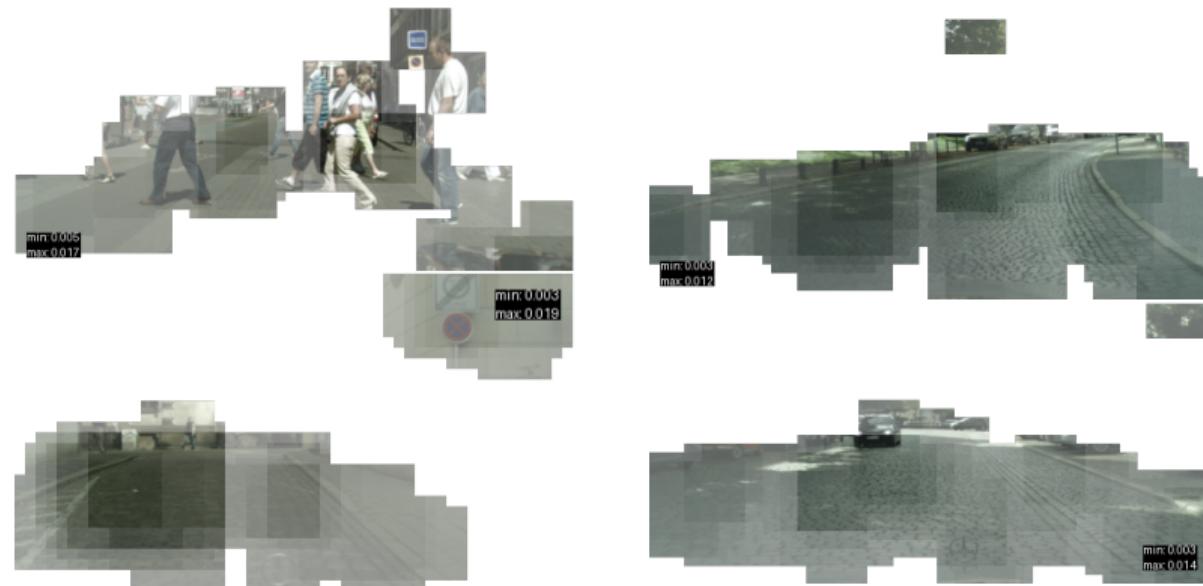
- spatially aligned latent representations
- scale-independent prototypes
 - ASPP: shared weights
 - activations examined at multiple scales
- class independence
 - “sparse classifier”



Several prototypes belonging to one concept

concept contribution
to semantic classes:

- road (12.29)
- sidewalk (3.68)
- wall (1.29)
- ground (0.71)



Bibliography I

-  Chaofan Chen, Oscar Li, Chaofan Tao, Alina Jade Barnett, Jonathan Su, and Cynthia Rudin.
This looks like that: Deep learning for interpretable image recognition, 2019.
-  Liang-Chieh Chen, George Papandreou, Florian Schroff, and Hartwig Adam.
Rethinking atrous convolution for semantic image segmentation, 2017.
-  Marius Cordts, Mohamed Omran, Sebastian Ramos, Timo Rehfeld, Markus Enzweiler, Rodrigo Benenson, Uwe Franke, Stefan Roth, and Bernt Schiele.
The cityscapes dataset for semantic urban scene understanding.
In *Proc. of the IEEE Conference on Computer Vision and Pattern Recognition (CVPR)*, 2016.
-  Kaiming He, Xiangyu Zhang, Shaoqing Ren, and Jian Sun.
Deep residual learning for image recognition, 2015.

Bibliography II

-  Meike Nauta, Jörg Schlötterer, Maurice van Keulen, and Christin Seifert.
Pip-net: Patch-based intuitive prototypes for interpretable image classification.
In *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition (CVPR)*, pages 2744–2753, June 2023.
-  Hugo Porta, Emanuele Dalsasso, Diego Marcos, and Devis Tuia.
Multi-scale grouped prototypes for interpretable semantic segmentation.
In *2025 IEEE/CVF Winter Conference on Applications of Computer Vision (WACV)*, pages 2869–2880. IEEE, 2025.
-  Mikołaj Sacha, Dawid Rymarczyk, Łukasz Struski, Jacek Tabor, and Bartosz Zieliński.
Protoseg: Interpretable semantic segmentation with prototypical parts, 2023.

Thank you for your attention!

Questions?

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