Credal Wrapper of Model Averaging for Uncertainty Estimation in Classification

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Background and Motivation

Uncertainty resources in neural networks

- <u>Aleatoric uncertainty (AU)</u>: data uncertainty, inherent randomness of the data generation process
- Epistemic uncertainty (EU): model uncertainty, a lack of knowledge

Effectively estimating uncertainties in predictions and the proper distinction between its aleatoric and epistemic forms are both crucial to:

- achieve robust performance of neural networks
- <u>benefit various downstream decision-based tasks</u>, such out-of-distribution (OOD) detection

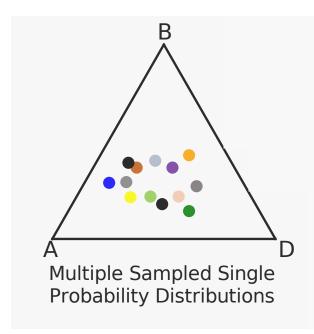
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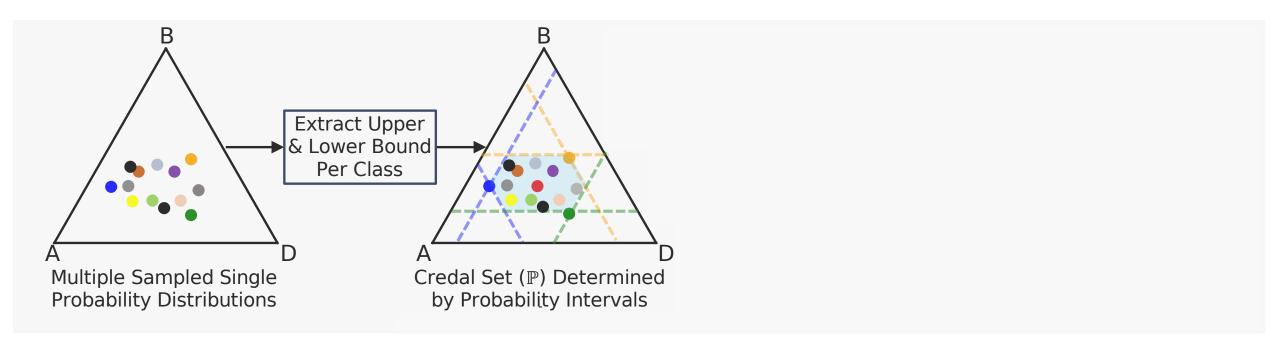
Widely applied approaches enabling total uncertainty (TU, AU and EU) estimation:

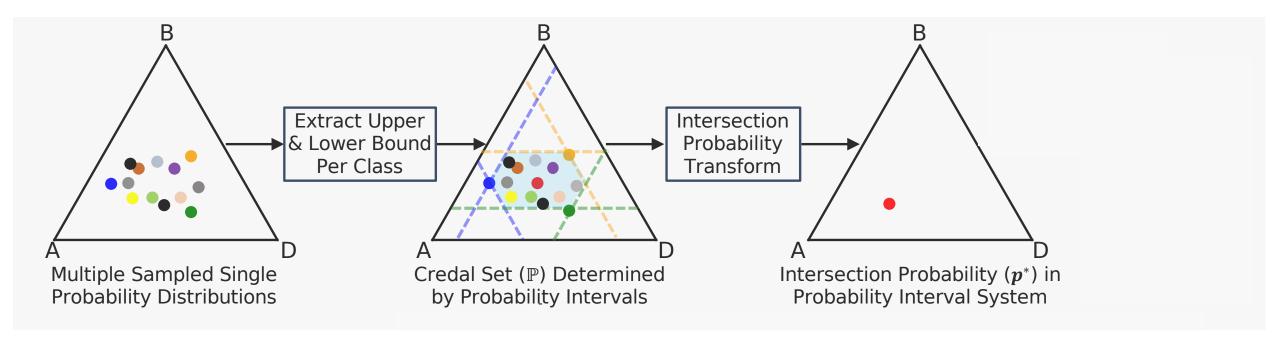
- Bayesian neural networks (BNNs) & Deep ensembles (DEs)
- multiple (SoftMax) probabilities for uncertainty estimation and averaging these probabilities for prediction, via multiple forward passes
- However, <u>a limited number</u> of forward passes at prediction time

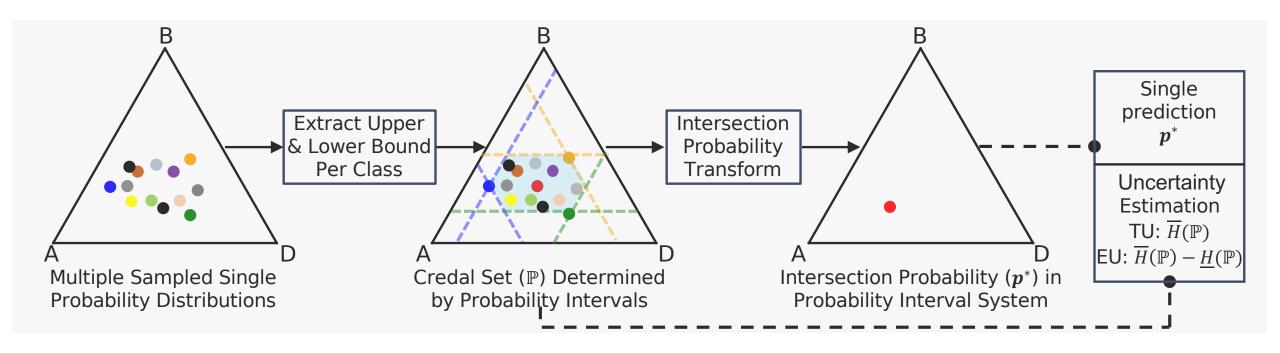
Main research objective:

• To <u>enhance</u> the <u>uncertainty quantification</u> and <u>prediction performance</u> of these approaches given this <u>constrained number</u> of predictive probabilities.









Performance of Credal Wrapper

- Experiment validations are conducted on: several OOD detection benchmarks, encompassing various dataset pairs (CIFAR10/100 vs SVHN/Tiny-ImageNet, CIFAR10 vs CIFAR10-C, CIFAR100 vs CIFAR100-C and ImageNet vs ImageNet-O) and using different network architectures (such as VGG16, ResNet-18/50, EfficientNet B2, and ViT Base)
- Compared to the BNN and DE baselines:
 - The proposed credal wrapper method exhibits <u>superior performance in</u> <u>uncertainty estimation</u>, leading to <u>improved OOD detection quality</u>
 - The usage of intersection probability achieves a <u>lower expected calibration</u> error on corrupted data

Thank You for Your Attention!