Robust Conformal Prediction With a Single Binary Certificate

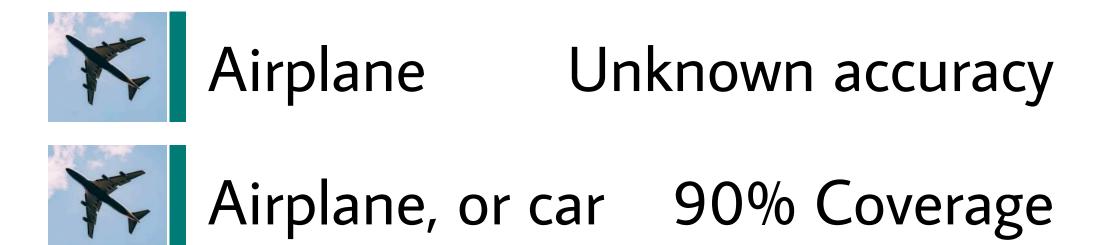
Soroush H. Zargarbashi, Aleksandar Bojchevski





What is Conformal Prediction?

Instead of a single prediction (unknown accuracy), predict a set of labels, guaranteed to include the true answer.



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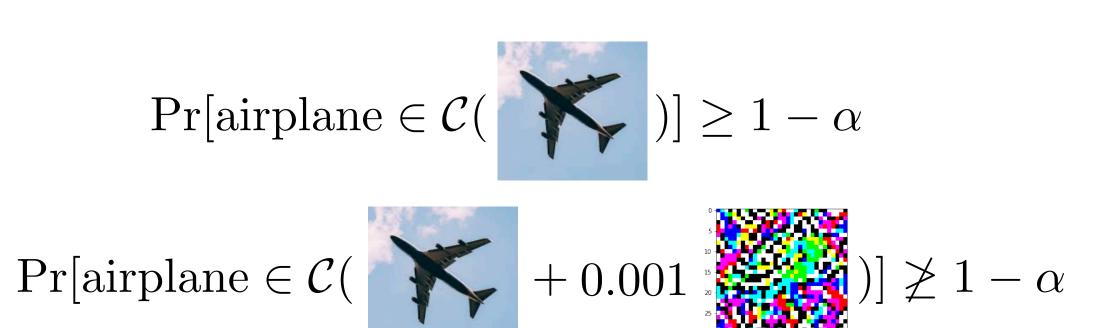
Airplane Unknown accuracy



Airplane, or car 90% Coverage

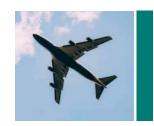
Conformal Guarantee can break!

Small noise can decrease the coverage drastically!



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$$\Pr[\text{airplane} \in \mathcal{C}(\text{0} + 0.001 \text{ } \text{0} \text{0})] \not\geq 1 - \alpha$$

Robust Conformal Prediction

Same guarantee, extended to the worst case noise.

$$\Pr[y_{n+1} \in \mathcal{C}(\boldsymbol{x}_{\text{noisy}}), \boldsymbol{x}_{\text{noisy}} \in \mathcal{B}(\boldsymbol{x}_{\text{clean}})] \ge 90\%$$

Our Results in One Glance

Prediction sets guaranteed to include the true label even with the worst case noise!



{Airplane, car}

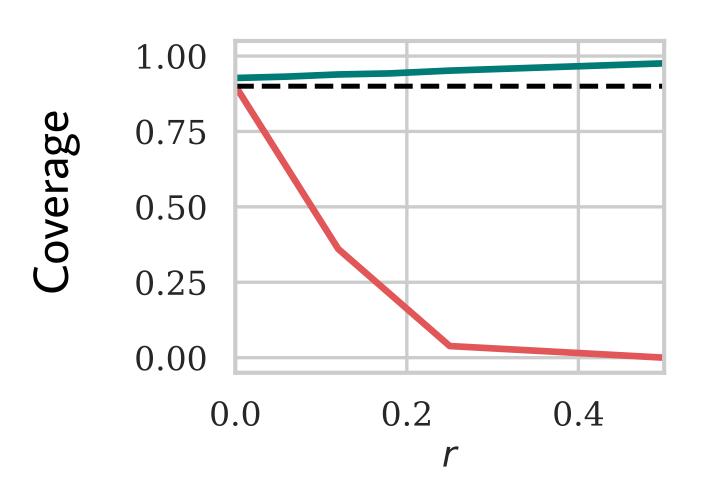
90% Coverage



{Plane, car, cat} Still 90% Coverage

CIFAR-10 under adversarial perturbation





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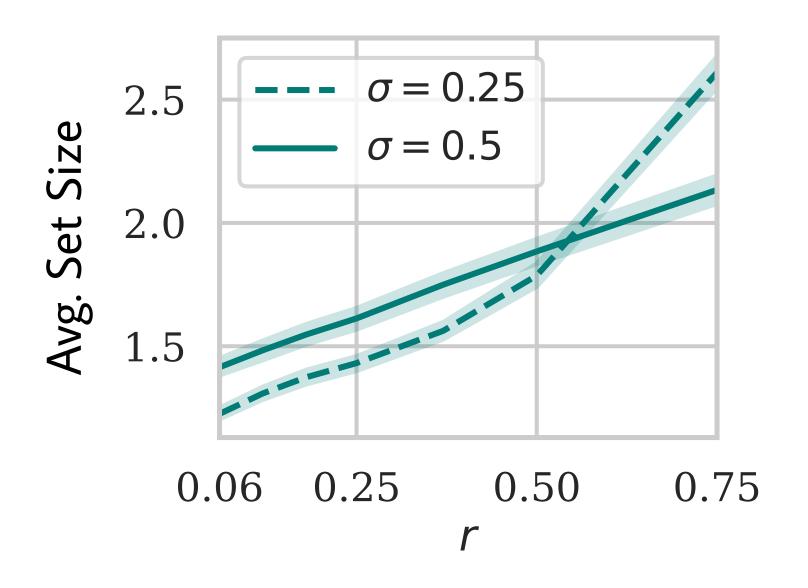
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Airplane, car, or cat Still 90% Coverage

Works with many existing smoothing certificates.

Using L1 De-randomized Certificate



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Airplane, or car

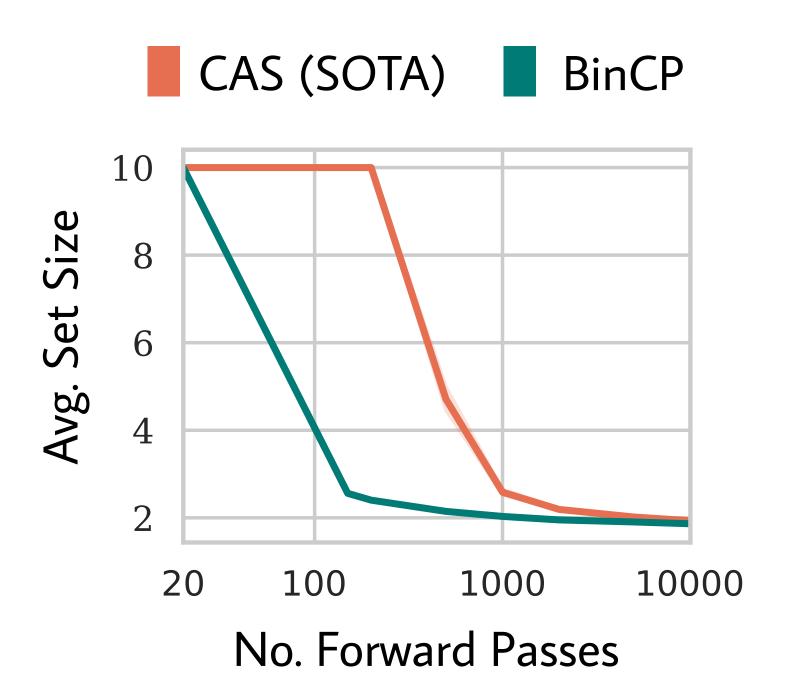
90% Coverage



Airplane, car, or cat Still 90% Coverage

Works with many existing smoothing certificates.

Returns smaller sets with fewer forward passes.



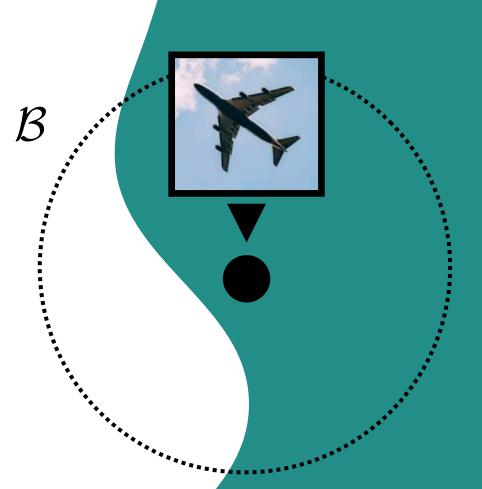
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Airplanes

(Decision Boundary)

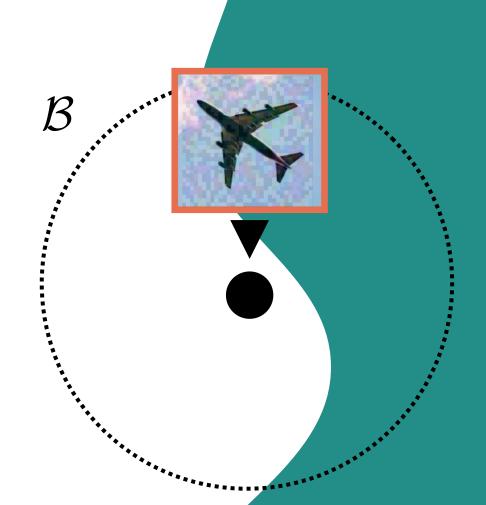


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Worst case noise: Small perturbation, same semantics, different prediction!



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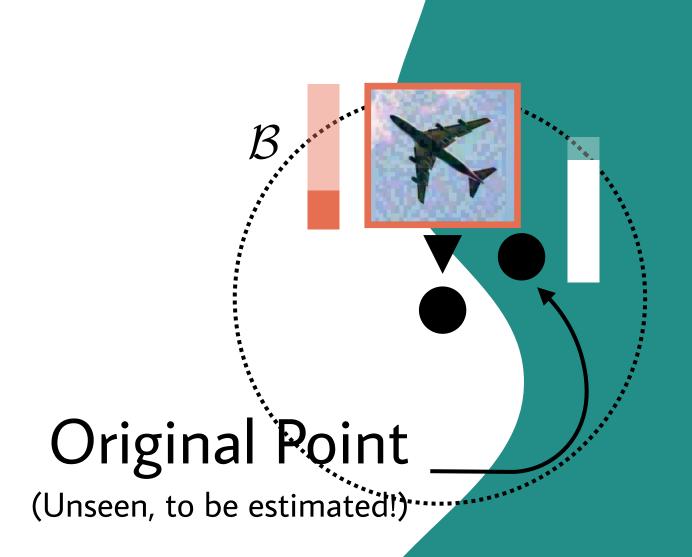
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How CP Works?

A score function captures agreement between input, and labels: e.g. softmax.

We accept any label with score above some computed threshold τ .

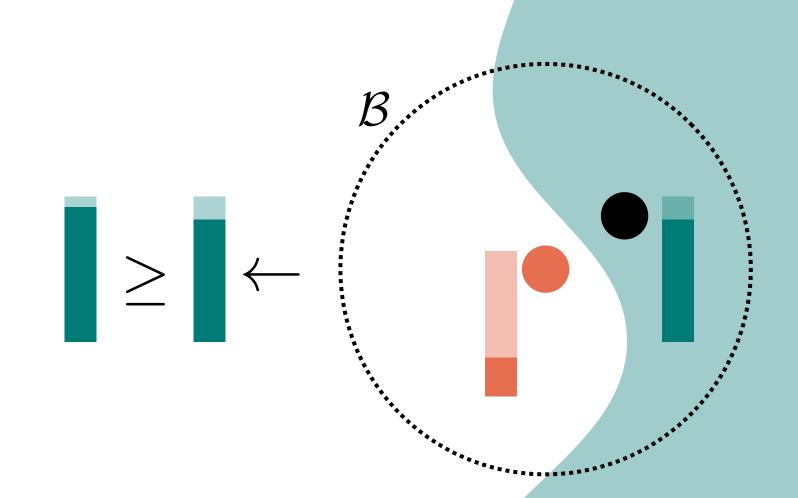


How Robust CP Works?

Upper bound the maximum score in the perturbation ball! How?

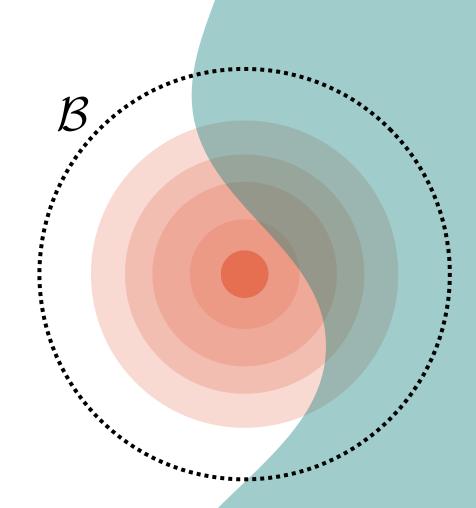
Neural network verifiers? Limited!

What else? Smoothing!



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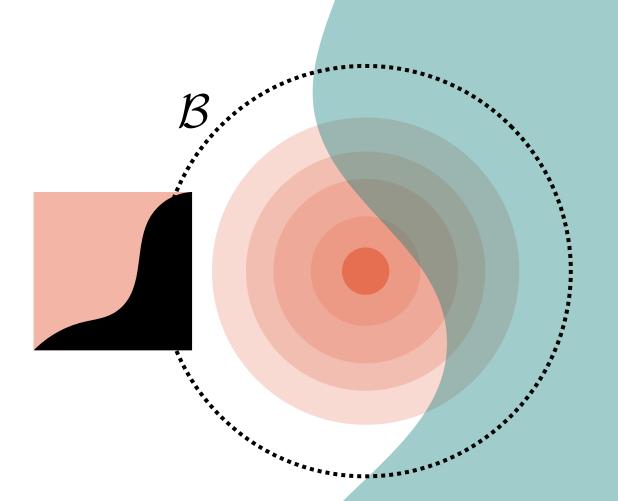
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$$S_i = s(\boldsymbol{x}_i + \boldsymbol{\epsilon})$$

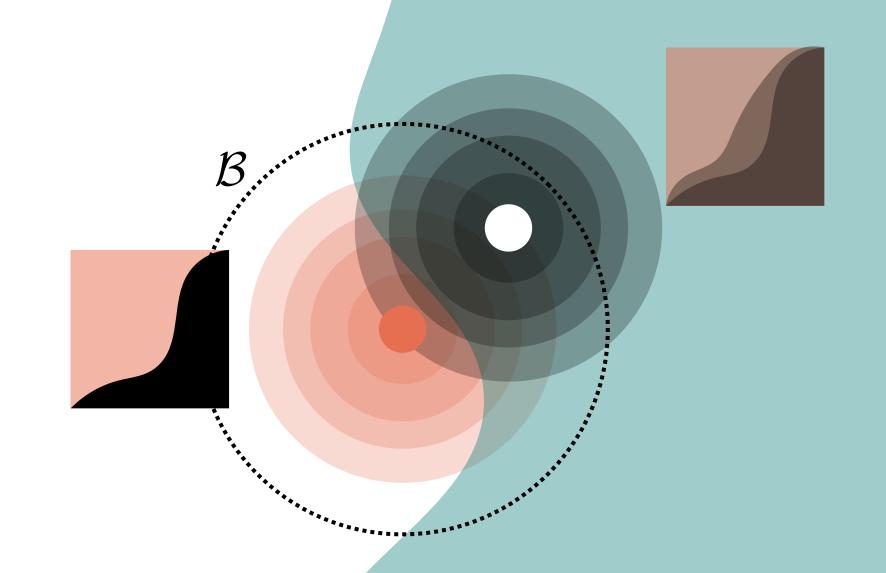


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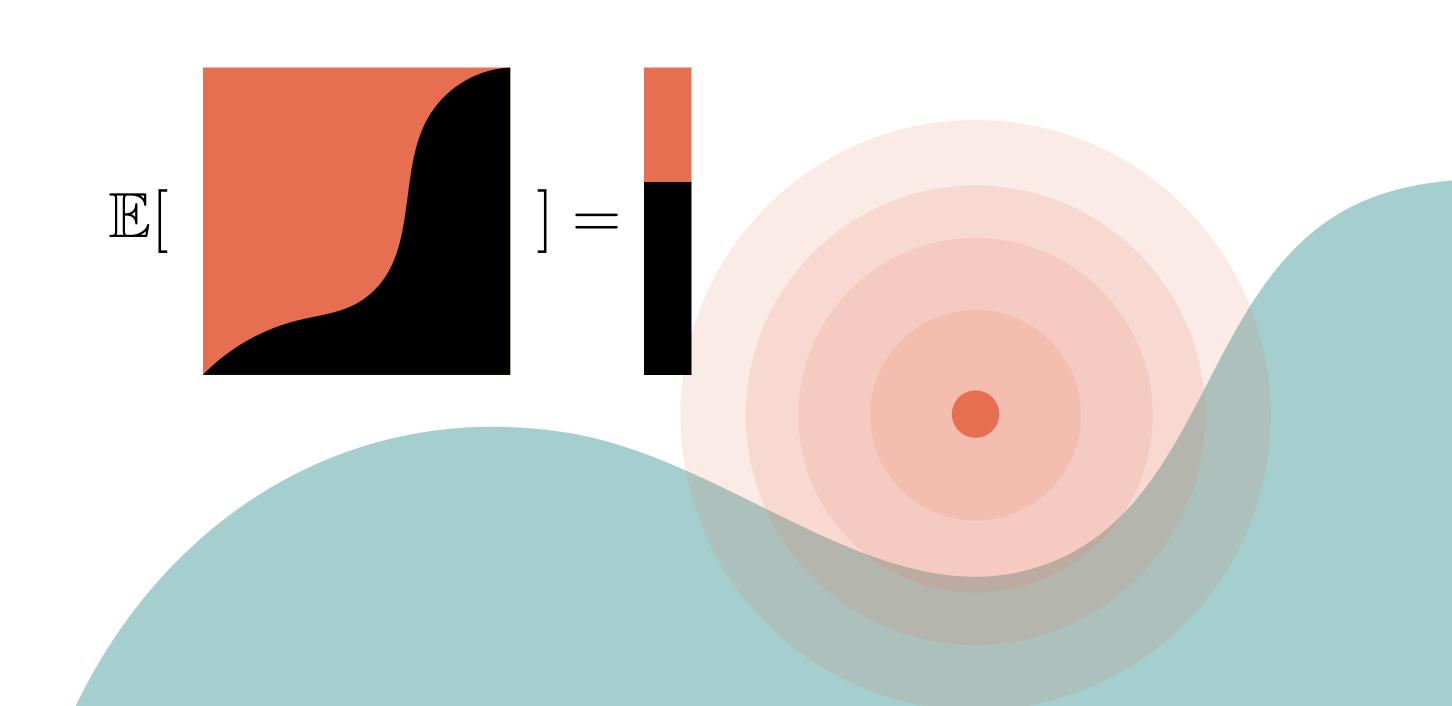
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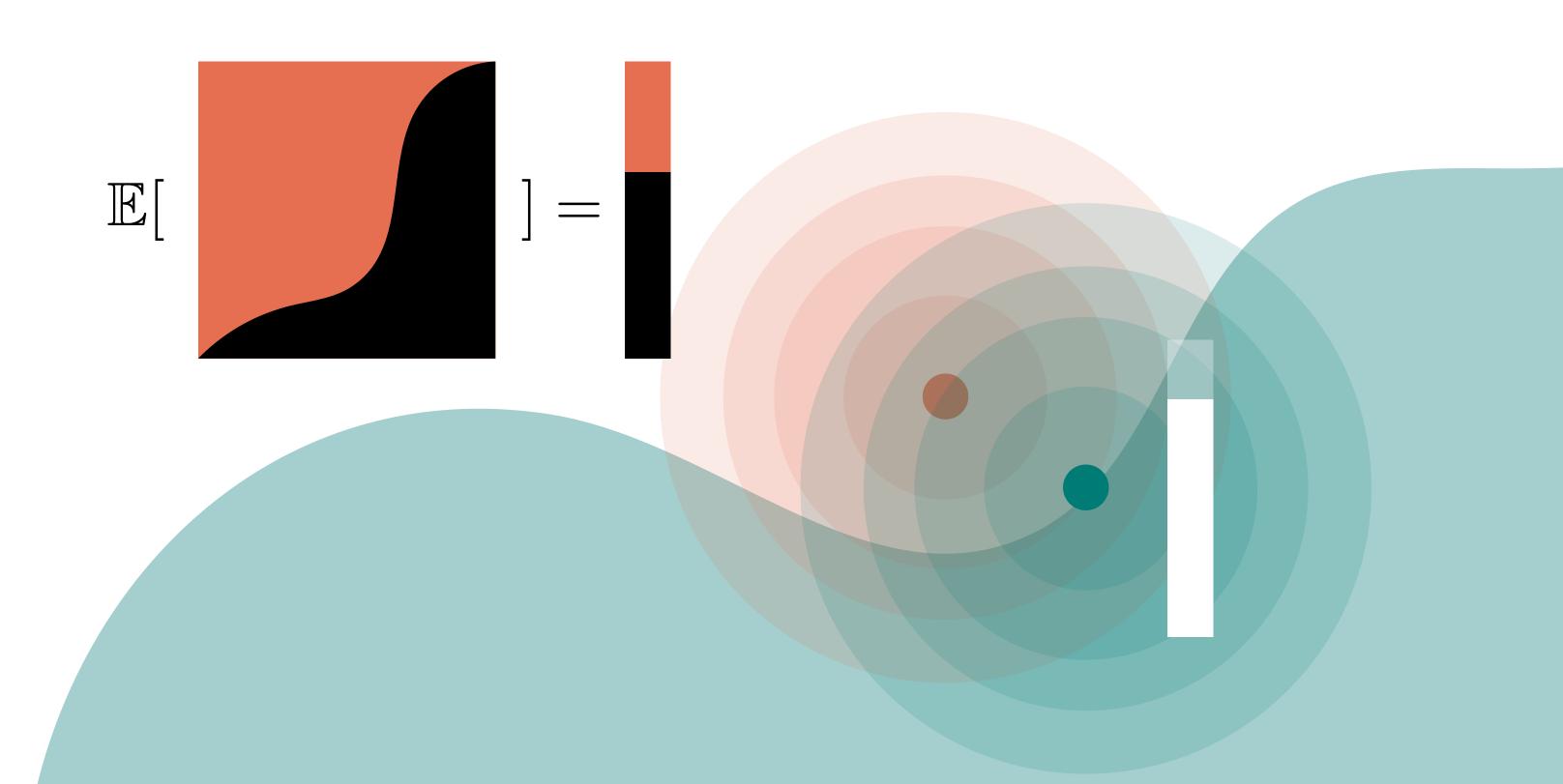
This distribution changes slowly around the input $oldsymbol{x}_i$.



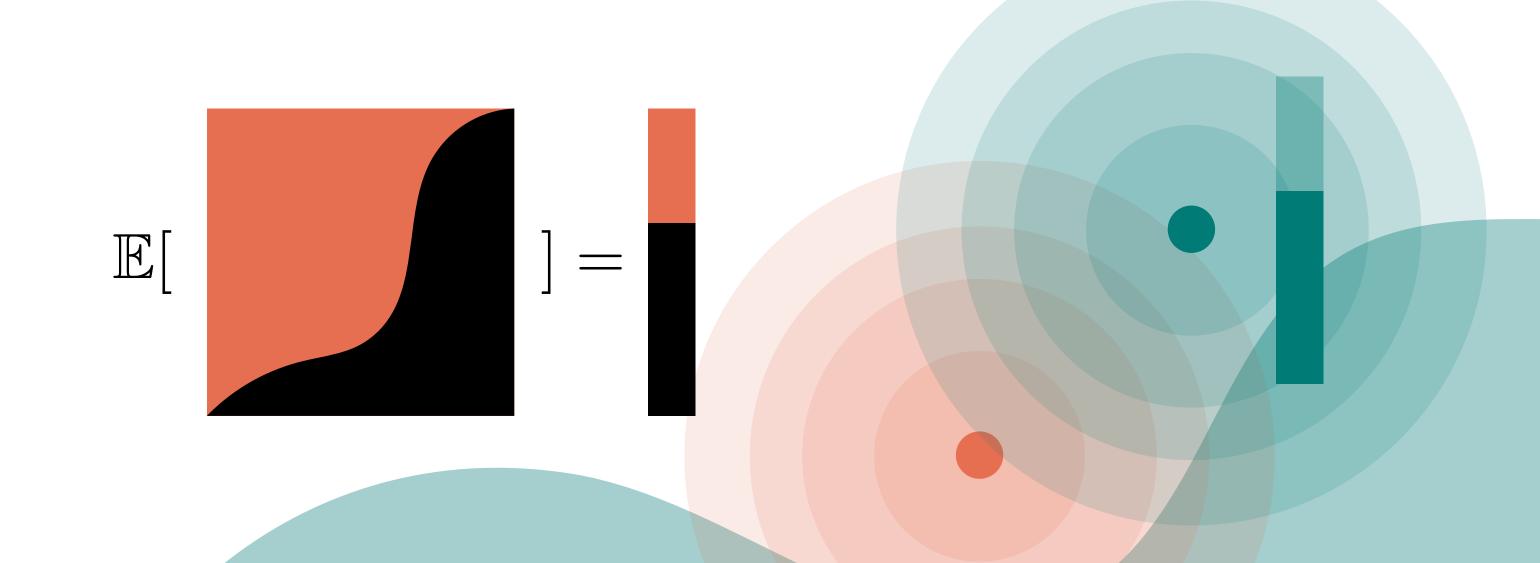
How to summarize a distribution $S_i = s(\boldsymbol{x}_i + \boldsymbol{\epsilon})$ Into a single number? Baseline: Take the mean! $\mathbb{E}[S_i]$



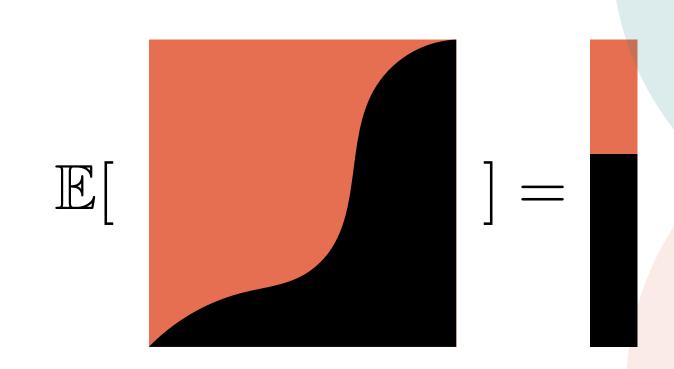
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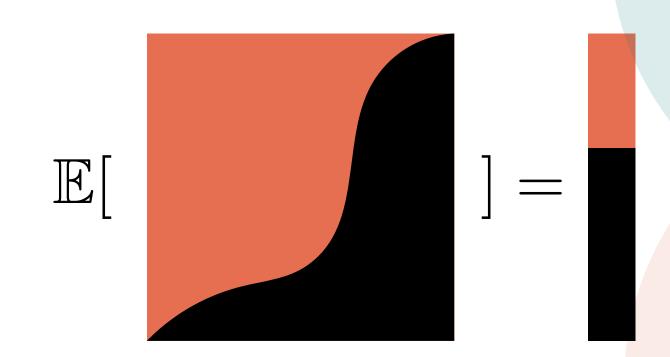


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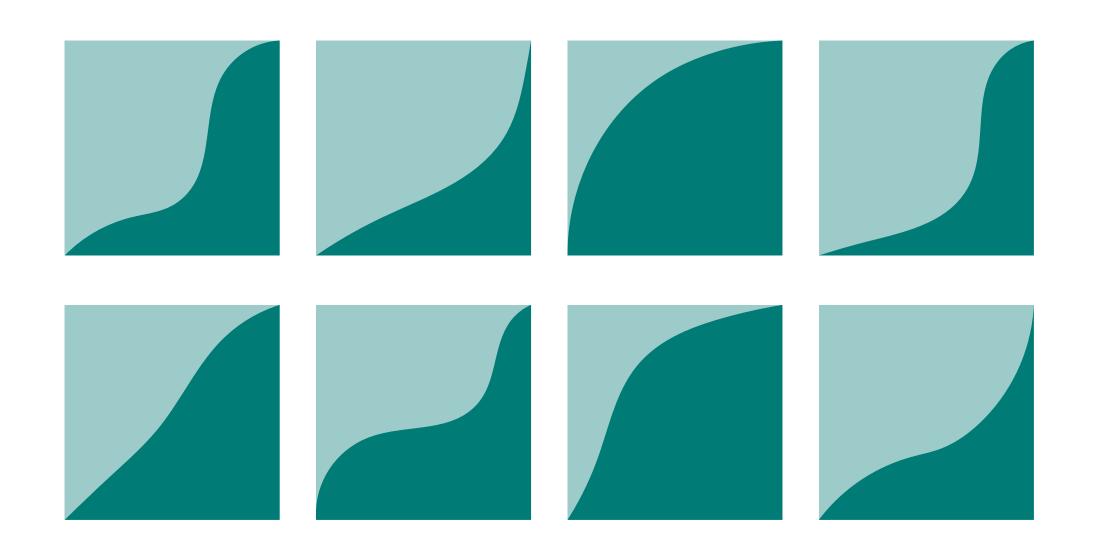


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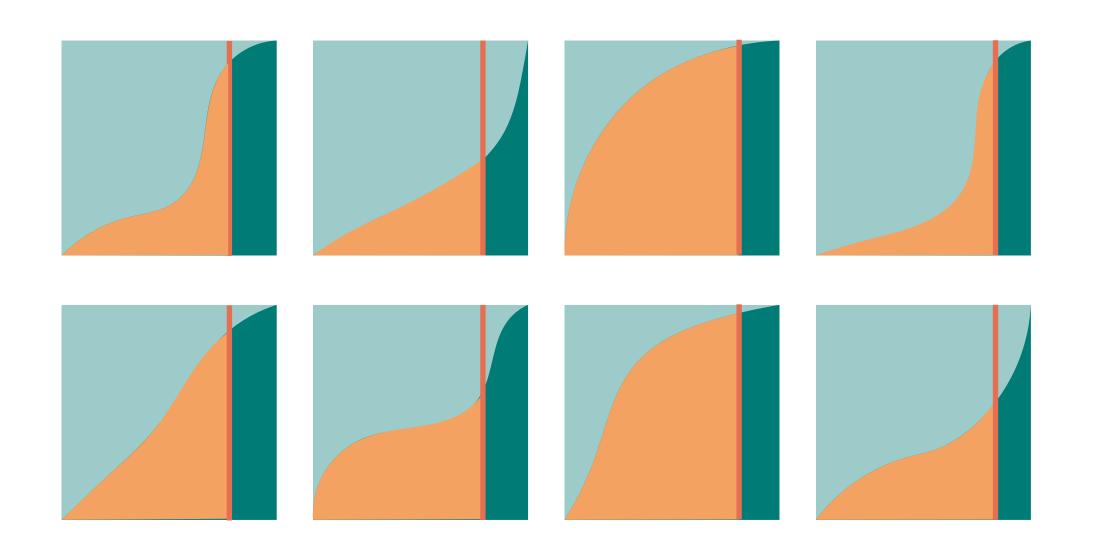
Many Monte-Carlo samples are needed!



We take the quantile of this distribution as the conformity score.

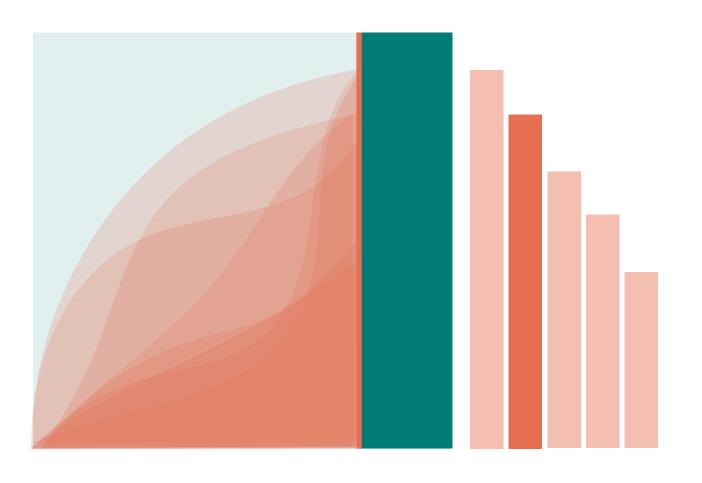


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We compute the conformal quantile of quantiles.



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Robust BinCP

Turns out we only need one binary certificate. Read our paper to see why!

Due to the binary variable we can use tighter confidence intervals.

Vanilla threshold Robust threshold

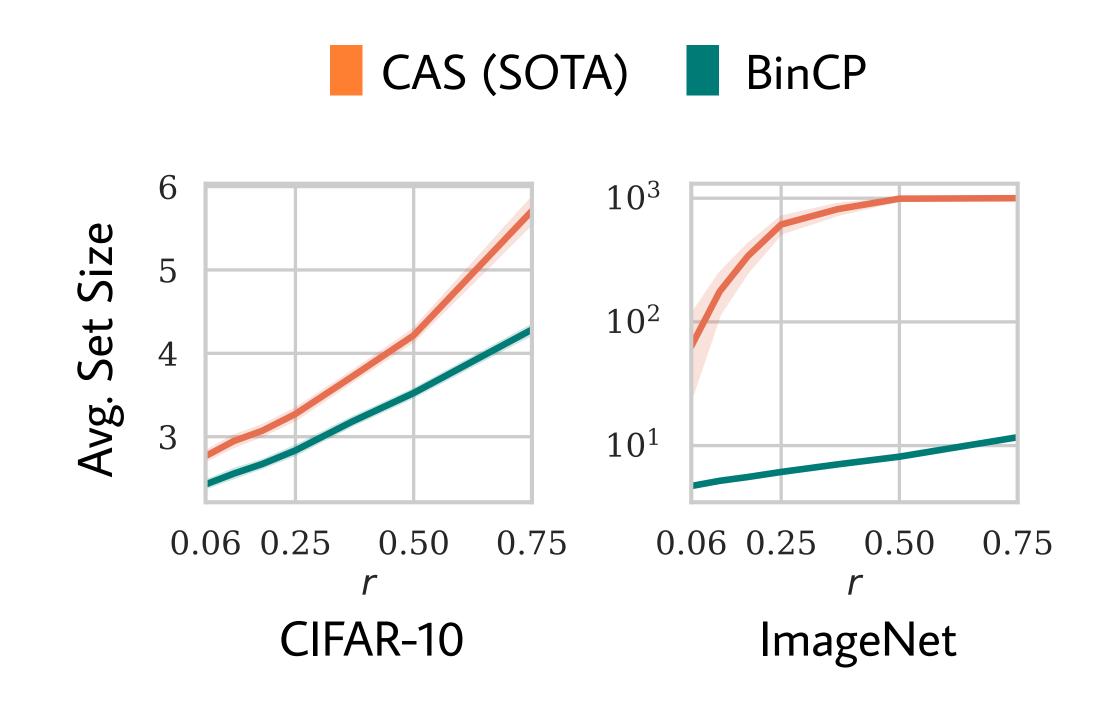
Results

Works on any binary certificate providing probability lower bounds.

Uses tighter confidence intervals

Smaller prediction set for fewer forward passes.

Even more improvements on datasets with many classes.



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Paper, Code =





