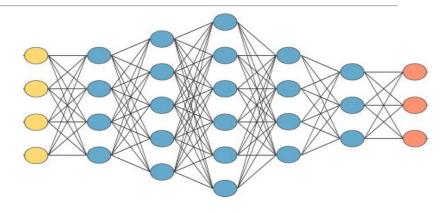
Efficient Hyperparameter Optimization of Deep Learning Algorithms Using Deterministic RBF Surrogates

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Deep neural networks are great but...

- They have many hyperparameters
- They are very sensitive to hyperparameter values
- Very hard to guess good hyperparameter values
- Solution: use hyperparameter optimization algorithms



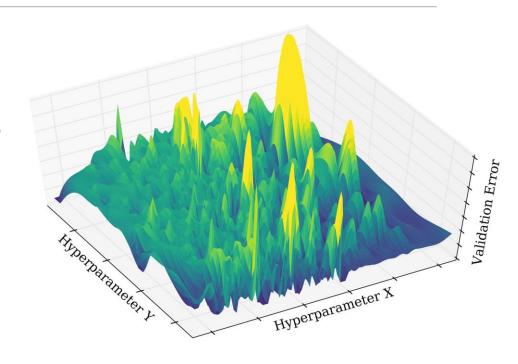
- number of layers and neurons
- learning rate and momentum
- dropout rate
- weight initialization
- and many others...

Hyperparameter optimization is **not** easy

$$E_{val} = F(\mathbf{x})$$

Validation Error Hyperparameters

- One hyperparameter evaluation requires DNN training to convergence that can take several hours
- The hyperparameter space has large number of local minima
- Difficult non-convex optimization in high dimensions

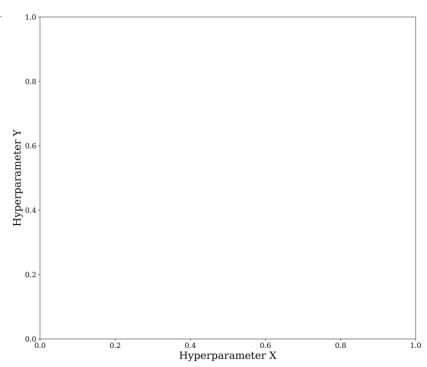


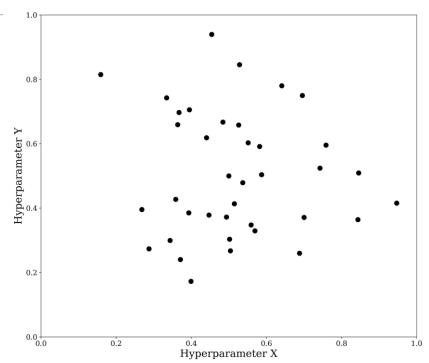
Our approach: Use surrogate model...

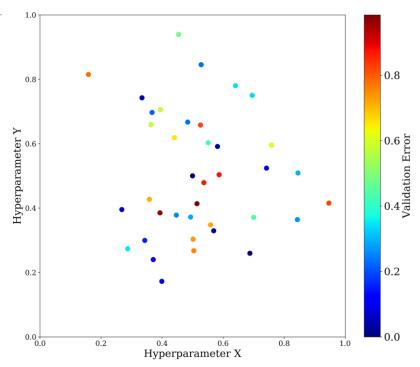
 Approximate the expensive hyperparameter evaluation with a surrogate model:

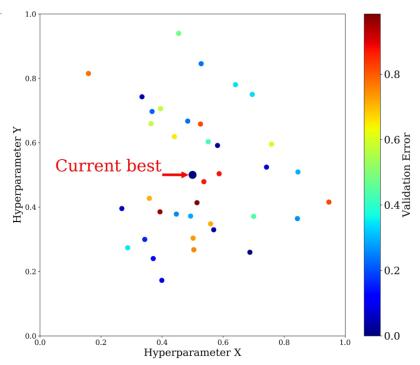
$$F(\mathbf{x}) \approx S_n(\mathbf{x}) = \sum_{i=1}^n \lambda(\|\mathbf{x} - \mathbf{x_i}\|)^3 + \mathbf{b}^\top \mathbf{x} + a$$

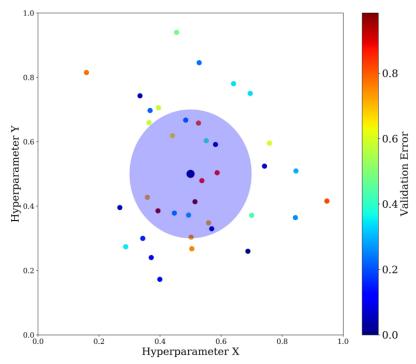
 Experiments show that Cubic Radial Basis Function surrogate with polynomial tail fits the highly non-convex and spiky hyperparameter space surprisingly well

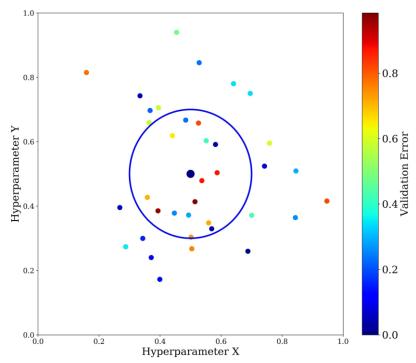






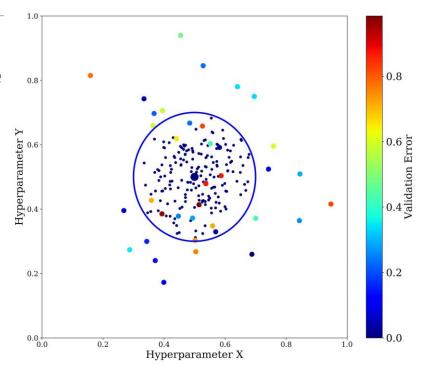






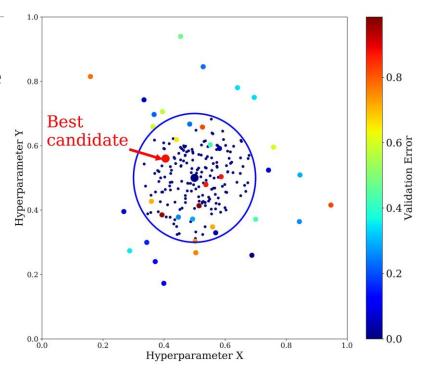
Explore the surrogate space for optimal hyperparameter values by evaluating candidate points around the current best found solution:

 Rank the candidate points according to weighted average of the surrogate value and distance to the current best solution

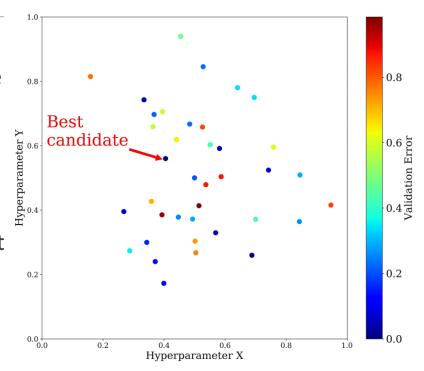


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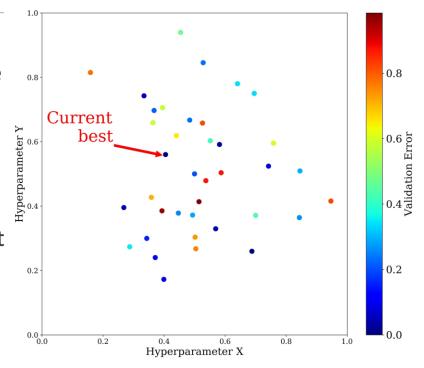
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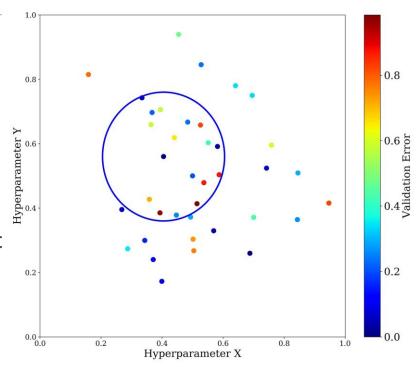
- Rank the candidate points according to weighted average of the surrogate value and distance to the current best solution
- 2. Perform expensive evaluation of the highest ranked hyperparameter values



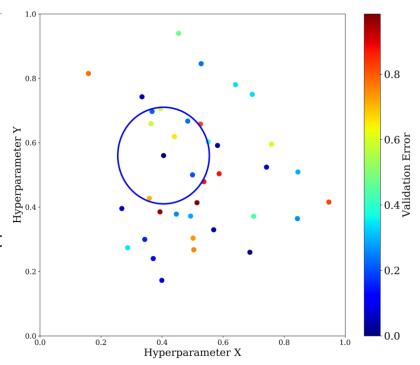
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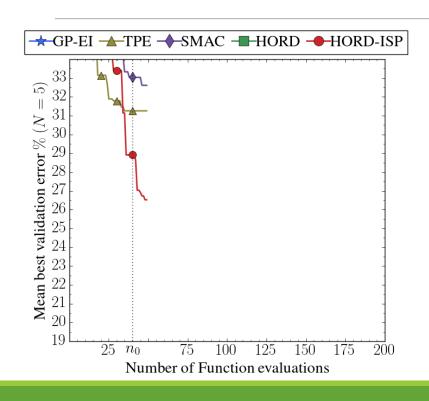
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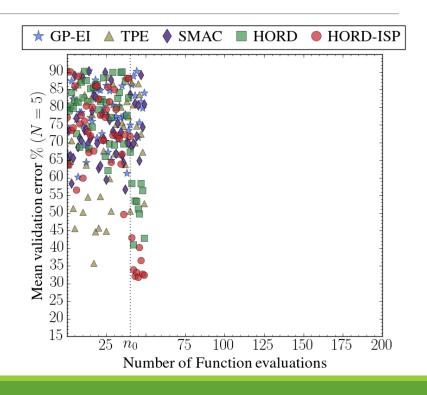


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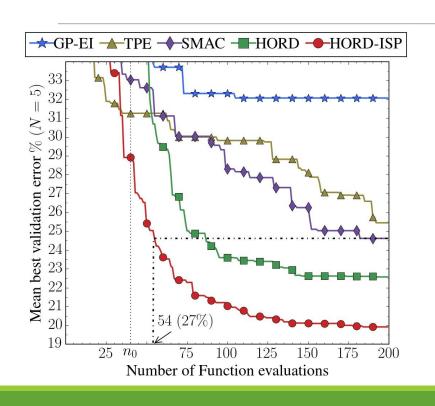


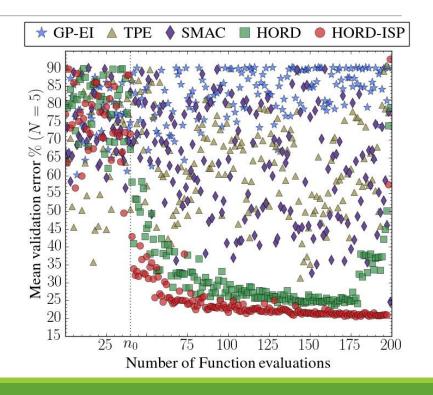
Results: Optimizing 19 CNN hyperparameters





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Thank you. Questions?

"Efficient Hyperparameter Optimization of Deep Learning Algorithms Using Deterministic RBF Surrogates", **AAAI-17**

<u>Ilija Ilievski</u>, Taimoor Akhtar, Jiashi Feng, and Christine Annette Shoemaker

Paper available at:

https://ilija139.github.io

Code:

https://github.com/ilija139/HORD