



Hyperparameter Optimization

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Hyperparameter Optimization

- **There are too many hyperparameters to be set by expert**
 - Learning rate, Momentum rate, Dropout, Normalization, Number of layers, number of nodes,
- **Is there any efficient way to set them?**
 - No

Hyperparameter Optimization

- **We can regard it as function optimization**
 - I will use 3-layer perceptron
 - I want to find out the optimal hyperparameters
 - Learning rate: η
 - Momentum rate: r
 - Dropout probability: p
 - Number of nodes in the first hidden layer: m
 - Number of nodes in the second hidden layer: n
 - Then, the accuracy of my NN is a function of η, r, p, m, n

$$Accuracy = f(\eta, r, p, m, n)$$

Hyperparameter Optimization

- **We can regard it as function optimization**

- I need to solve

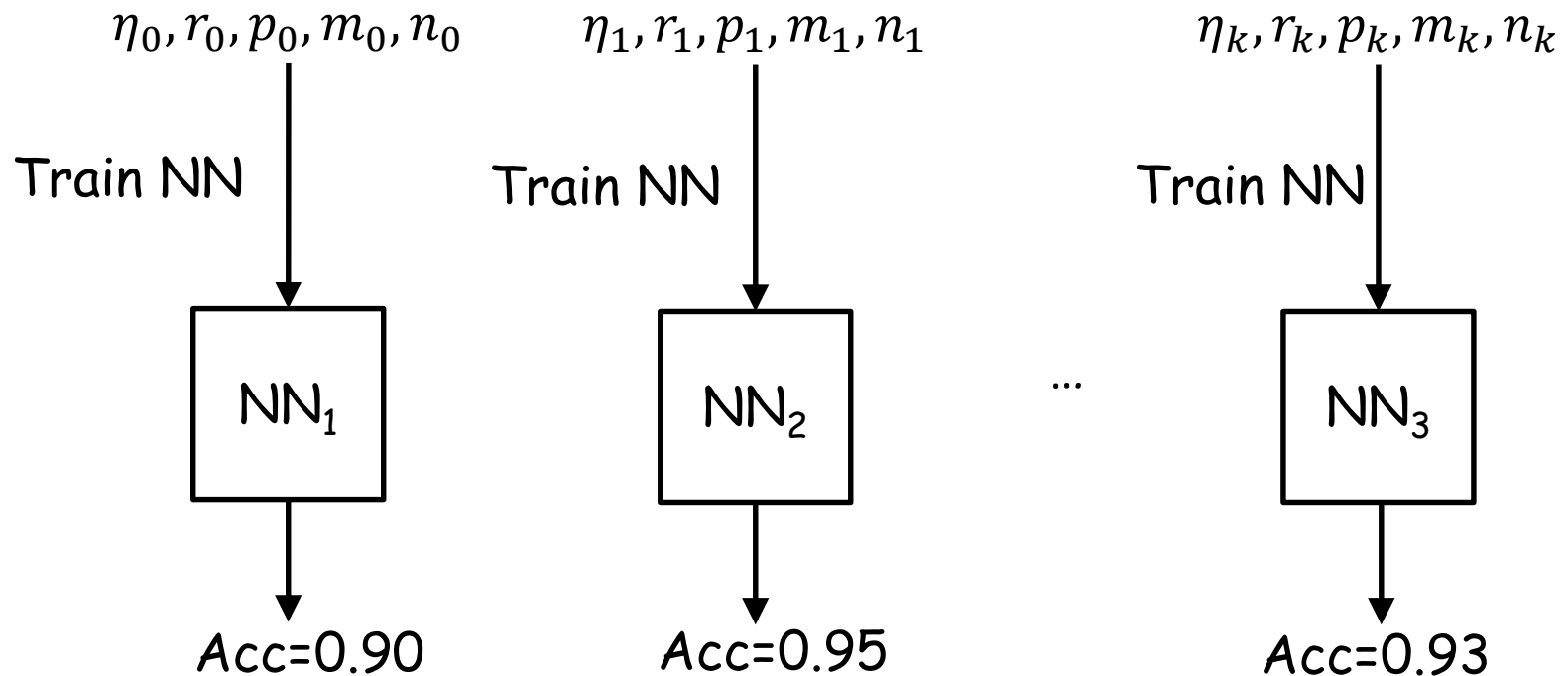
$$\operatorname{argmax}_{\eta, r, p, m, n} f(\eta, r, p, m, n)$$

- Problem

- I do not know what $f(\eta, r, p, m, n)$ is
 - But I can query. That is, for a some setting we can evaluate f
 - However, the evaluation is very expensive because we need to train a neural network !!

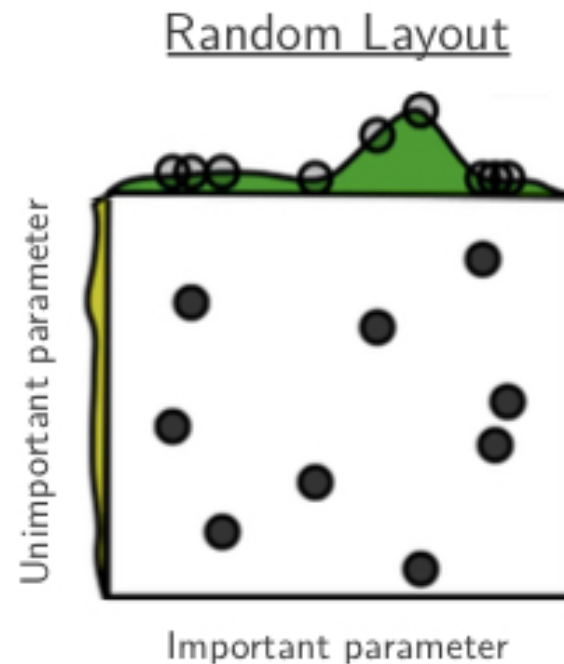
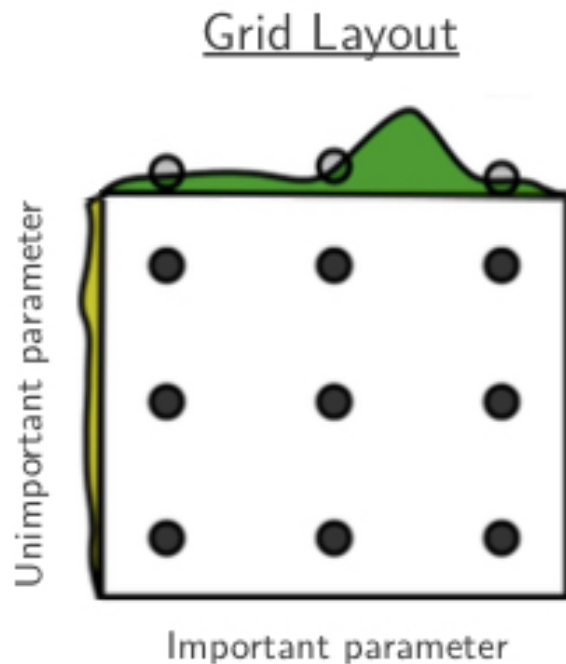
Hyperparameter Optimization

Choose settings



Some Simple Search Algorithm

- Random search, Grid search



Some Simple Search Algorithm

- **It is not efficient**

- It does not utilize the previous tries.
 - It would be better to less search the area with low potential and more search the area with high potential
- => Can we choose a better next point based on previous search results?

- **It is too costly**

- To evaluate how a set of hyperparameters is good, we need to train a neural network
 - we need to train a large neural network every time we try a new set of parameters
- => Can we gradually train NNs as search goes on?

Bayesian Optimization

■ Definition

$$\arg \max_x f(x)$$

- You don't know anything about $f(x)$
- You can query but it is very expensive
- Any good idea??

Bayesian Optimization

■ Any Good Idea??

- No information on $f(x)$...
- First choose a random point, x_1 , and evaluate $f(x_1)$
- Guess the shape of $f(x)$ based on $(x_1, f(x_1))$
- Based on the guess, choose the next point, x_2 , and evaluate $f(x_2)$
- Guess shape of $f(x)$ based on $\{(x_1, f(x_1)), (x_2, f(x_2))\}$
- Repeat those steps

Bayesian Optimization

- Overall Description

