

# Hyper- parameter tuning

# Agenda

## 01 What is hyperparameter tuning?

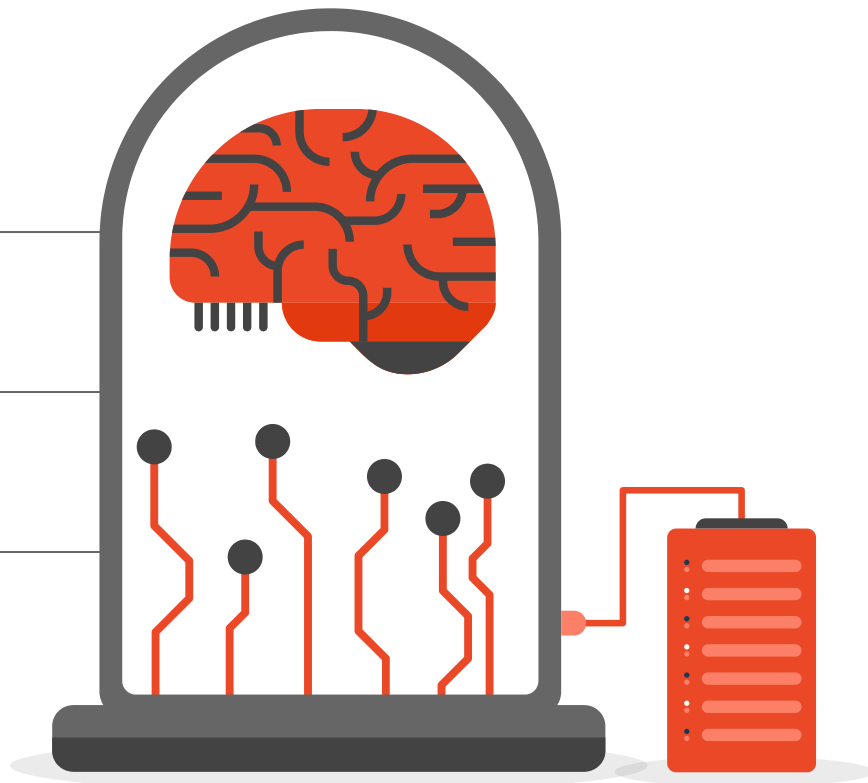
Why do we need it?

## 02 Common Hyperparameters

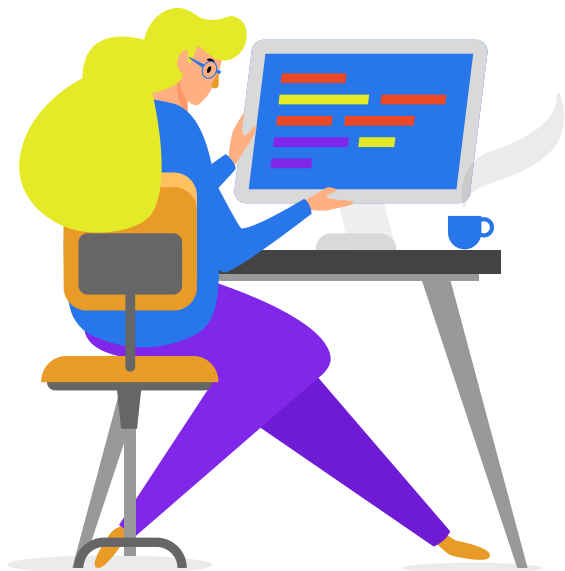
What hyperparameters are generally tuned?

## 03 Hyperparameter optimization algorithms

What algorithms are used to help us  
In tuning hyperparameters?



# What is hyperparameter tuning?



## Hyperparameter Tuning

- Optimizing hyperparameters to minimize loss
- Generalized solution
- Reasonable timeframe
- No “one correct way”

# What is hyperparameter tuning?



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**Dataset**

**Model type**

**Comp. Power**

# Common hyperparameters

## Learning rate

The rate of adjustment each step

01

## Momentum

“Resistance” to rapid changes in values

02

## Number of hidden layers

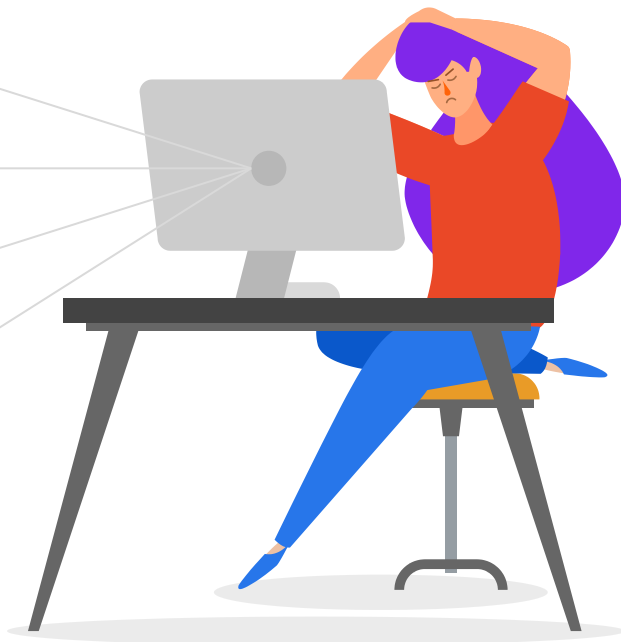
More isn't always better

03

## Number of neurons in each layer

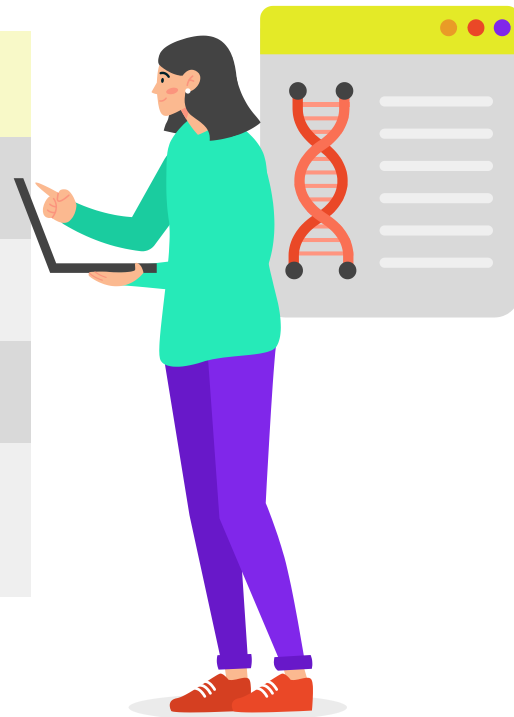
Memorization vs Generalization

04

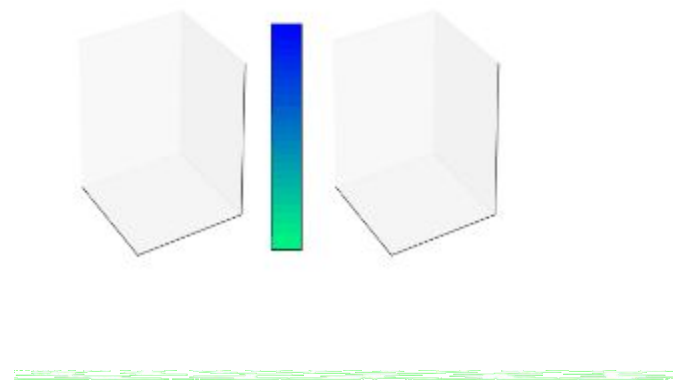
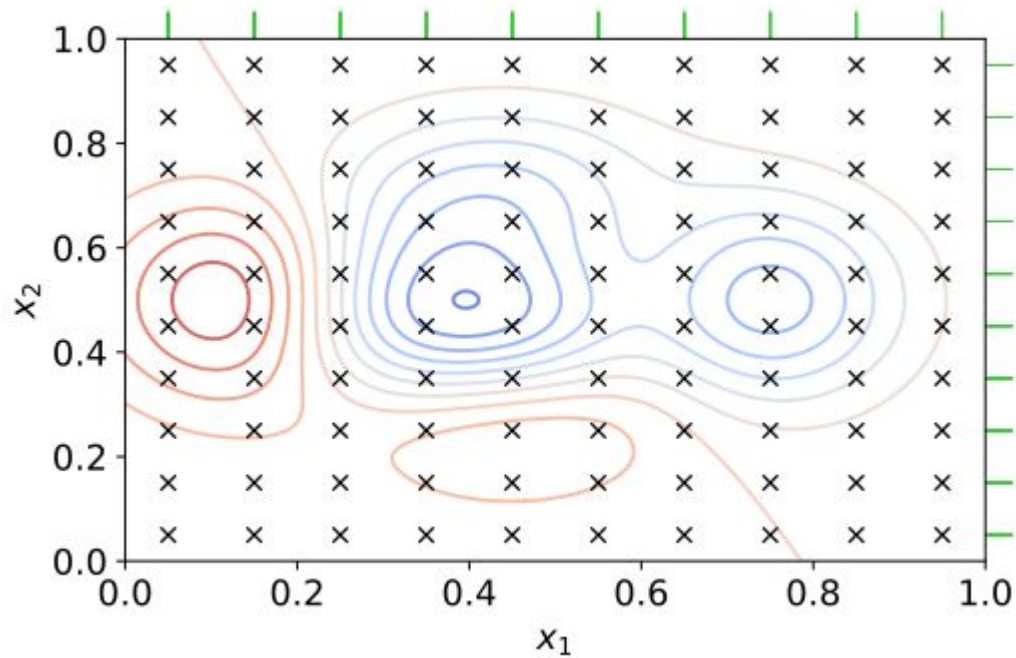


# Hyperparameter optimization algorithms

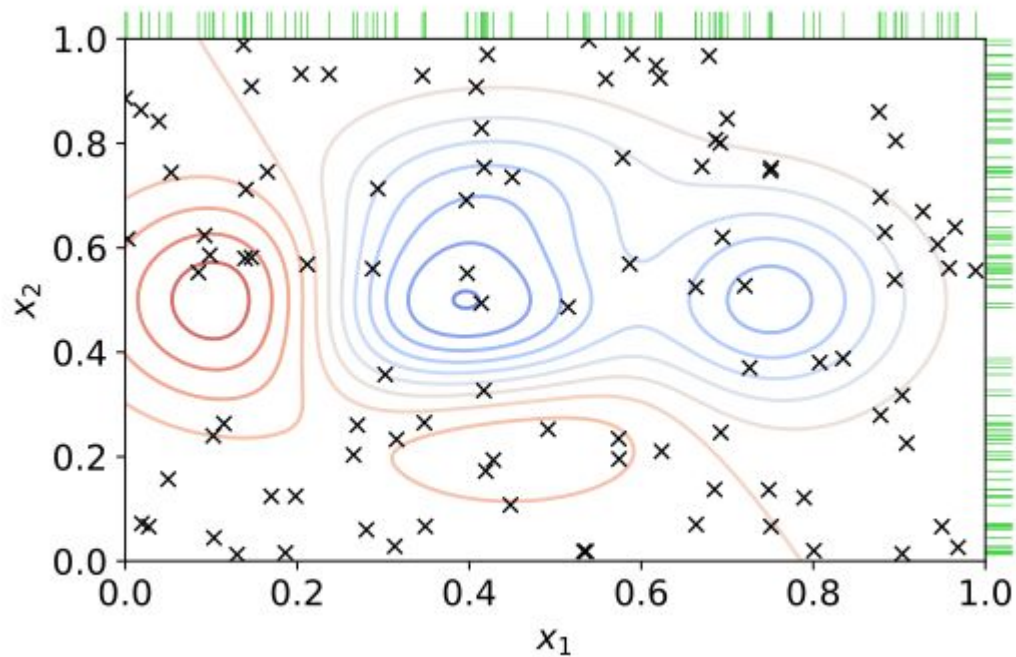
Algorithms	
Manual	Manually setting parameters
Grid Search	Computes all possible combination
Random Search	Randomly picks N combinations
Automated Hyperparameter Tuning	E.g. Bayesian Optimization and Genetic Algorithms



# Grid search

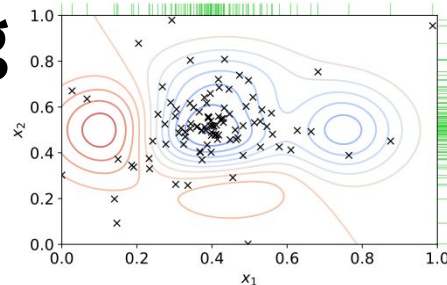


# Random search



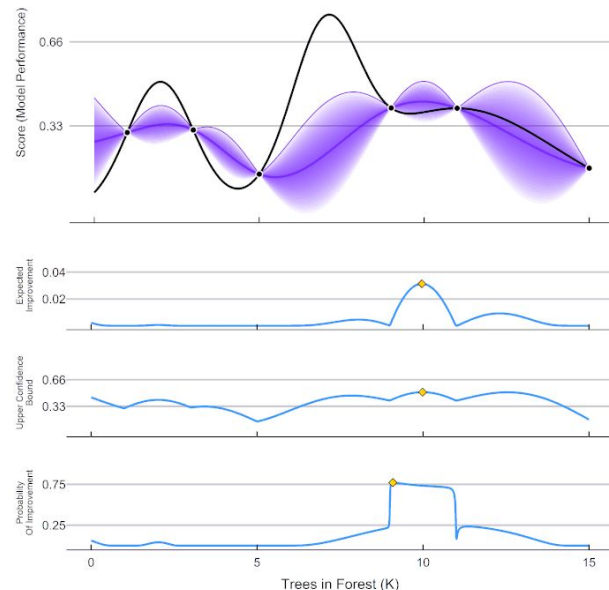


# Automated Hyperparameter Tuning (Bayesian Optimization)



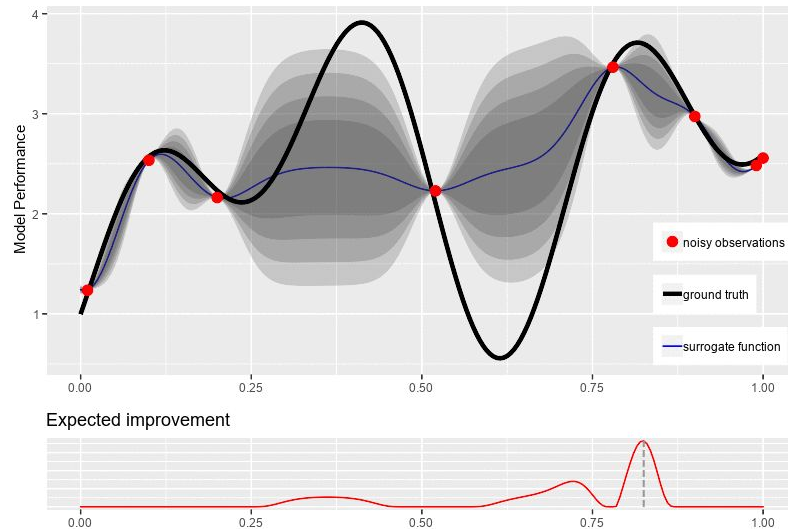
- Two key differences:
  - Uses a probabilistic surrogate function
  - Uses previous results
- Has two extra parts:
  - Surrogate function (SF)
  - Acquisition function (AF)
- Can use most probability functions
  - Surrogate: Gaussian probability
  - Acquisition:
    - Probability of improvement
    - GP upper confidence bound
    - Predictive entropy search
    - A portfolio of multiple acquisition strategies

ParBayesianOptimization in Action (Round 1)



# Automated Hyperparameter Tuning (Bayesian Optimization) cont.

- Consists of 5 steps:
  1. Initialize the surrogate function (SF)
  2. Use the Acquisition function (AF) to find the best performing set,  $X$
  3. Update the acquisition function based on SF
  4. Calculate  $OF(X)$
  5. Calculate the difference between  $OF(X)$  and  $SF(X)$  and update SF based on result



# Sources:

- T. Yu and H. Zhu, “Hyper-parameter optimization: A review of algorithms and applications,” *arXiv.org*, 12-Mar-2020. [Online]. Available: <https://arxiv.org/abs/2003.05689>. [Accessed: 12-Oct-2022].
- “What is hyperparameter tuning?,” *Anyscale*. [Online]. Available: <https://www.anyscale.com/blog/what-is-hyperparameter-tuning>. [Accessed: 12-Oct-2022].
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- T. Nunes, “Hyperparameter tuning,” *Medium*, 06-Aug-2020. [Online]. Available: <https://medium.com/analytics-vidhya/hyperparameter-tuning-8ca311b16057>. [Accessed: 12-Oct-2022].