

# MACHINE LEARNING

- Successful, but requires human labor and expertise
  - ▶ Pre-process data
  - ▶ Select/ engineer features
  - ▶ Select a model family
  - ▶ Optimize hyperparameters (algorithm parameters)
  - ▶ ...
- Deep learning lets us automatically learn features
  - ▶ Automates feature engineering step, with large amount of data
  - ▶ Even more sensitive to architectures, hyperparameters, ...

# AUTOMATIC MACHINE LEARNING I

- Can algorithms be trained to automatically build end-to-end machine learning systems?

Use machine learning to do better machine learning

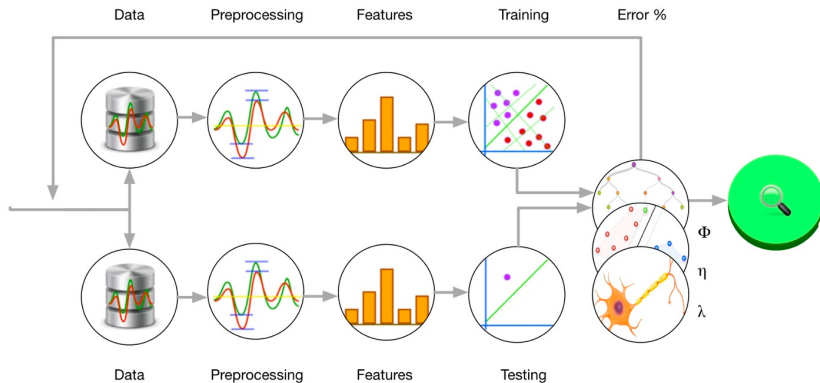
- Can we turn  
*Solution = data + manual exploration + computation*
- Into  
*Solution = data + computation (x100)*

# AUTOMATIC MACHINE LEARNING II

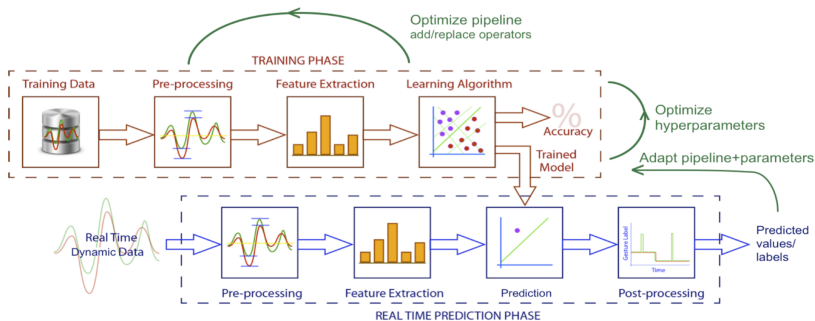
## Not about automating data scientists

- Efficient exploration of techniques
  - ▶ Automate the tedious aspects (inner loop)
  - ▶ Make every data scientist a super data scientist
- Democratisation
  - ▶ Allow individuals, small companies to use machine learning effectively (at lower cost)
  - ▶ Open source tools and platforms
- Data Science
  - ▶ Better understand algorithms, develop better ones
  - ▶ Self-learning algorithms

# MACHINE LEARNING PIPELINES



# AUTOMATING MACHINE LEARNING PIPELINES



# AUTOMATIC MACHINE LEARNING: TECHNIQUES

- **Bayesian Optimization:** Intelligently optimize pipelines/architectures by iteratively choosing better ones
- **Genetic algorithms:** Evolve pipelines/architectures to work better for a given application
- **Meta-learning:** learn from previous applications to predict useful pipelines/ architectures for new problems
- **Transfer Learning:** train models on one problem, then transfer (parts) of good solutions to solve new problems.
- **Reinforcement Learning:** Train many models, use performance as "reward" for certain approaches
- **Combinations of all of these**

# BLACK BOX OPTIMIZATION

