

# Evolving Neural Networks

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# Agenda

1. Evolving Neural Networks
2. A bit of History
3. Machine Learning (ML)
4. AutoML
5. AutoML: Neural Architecture Search (NAS)
6. Evolving Neural Architecture Search (NAS)
7. Recap
8. Tips for data practitioners in industry

# 1. Evolving Neural Networks

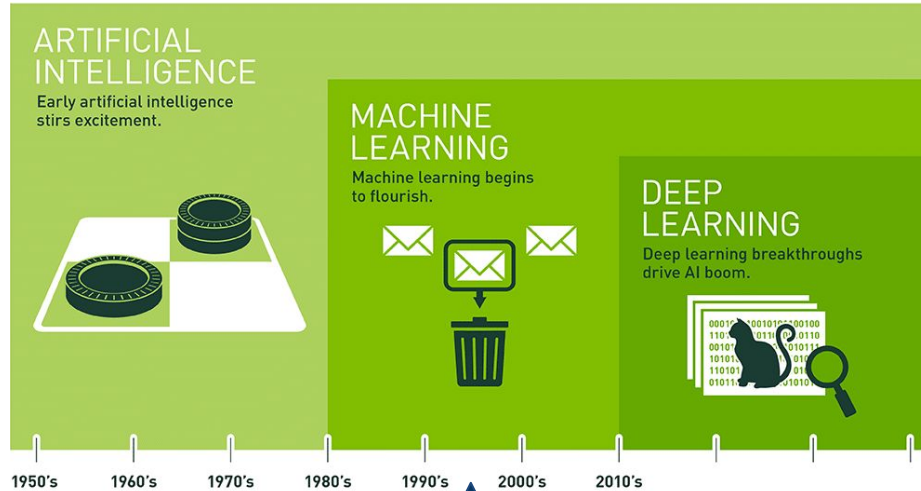
# Evolving Neural Networks

- Artificial Intelligence?
- Machine Learning?
- Dystopian society?

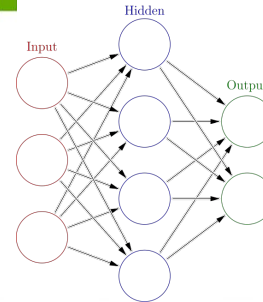


## 2. A bit of History

# A bit of History



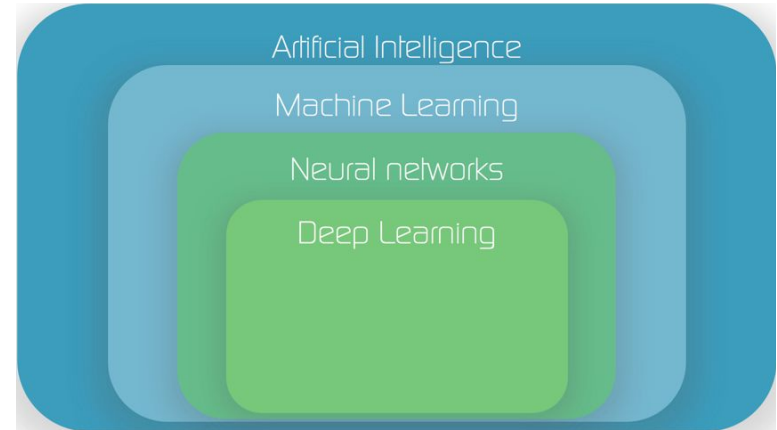
1. **A.I.**  
*Engineering of making Intelligent Machines and Programs*
2. **M.L.**  
*Ability to learn without being explicitly programmed*
3. **D.L.**  
*Learning based on Deep Neural Networks*



### 3. Machine Learning (ML)

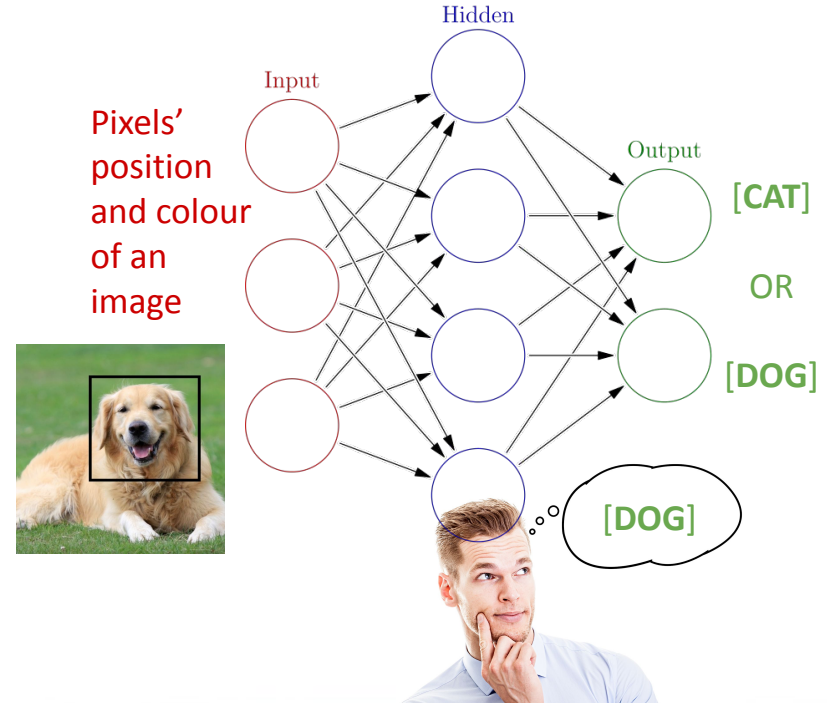
# Machine Learning (ML): Neural Networks

- *“A neural network is a series of algorithms that aims to recognize underlying relationships in a set of data through a process that mimics the way the human brain operates”*



# Machine Learning (ML): Neural Networks

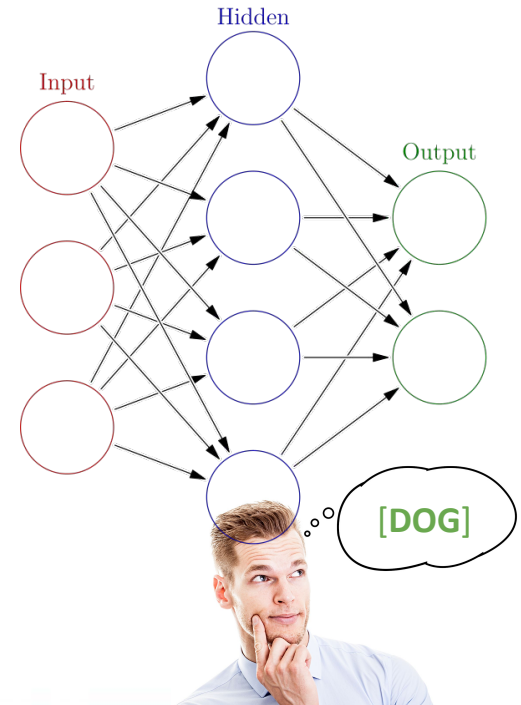
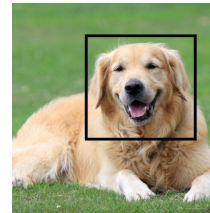
- *"A neural network is a series of algorithms that aims to recognize underlying relationships in a set of data through a process that mimics the way the human brain operates"*
- Ability to learn without being explicitly programmed
- Providing millions of data samples



# How will ML and NN evolve?

What is the future of ML and NN?  
How can they evolve?

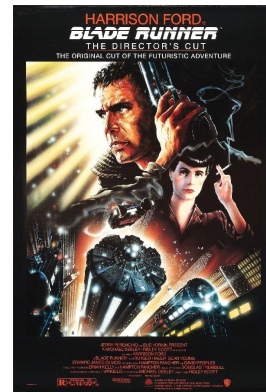
*“The biggest lesson that can be read from 70 years of AI research is that **general methods that leverage computation** are ultimately the most effective, and by a large margin” - Rich Sutton*



## 4. AutoML

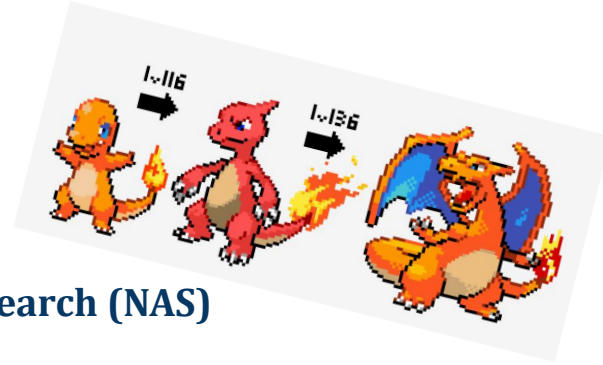
# AutoML vs ML

- **Automated ML** is the process of automating the tasks of applying machine learning to real-world problems.
- AutoML allows non-experts to make use of machine learning models and techniques.
- Advantages: simpler and faster solutions, models that often outperform hand-designed models.



# Evolving {insert fancy term}

- AI 
  - ML (and Neural Networks)
    - **AutoML: Neural Architecture Search (NAS)**
      - AutoML evolution/future?



## 5. AutoML: Neural Architecture Search (NAS)

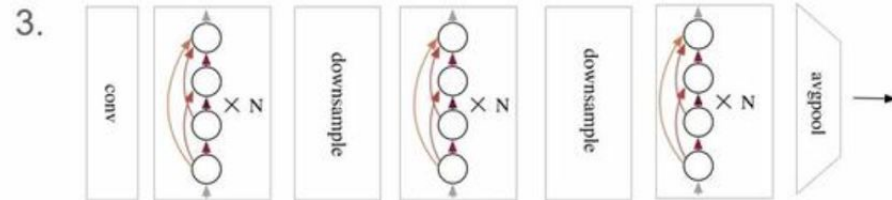
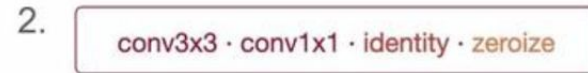
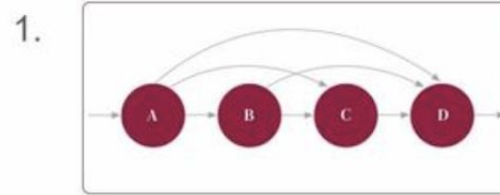
# AutoML: Neural Architecture Search (NAS)

- **NAS** is a technique for automating the design of neural networks.
  - Designing architectures by hand is hard!
  - NAS lets us get good architecture design automatically
  - However, NAS algorithms can be extremely expensive to run

# NAS spaces

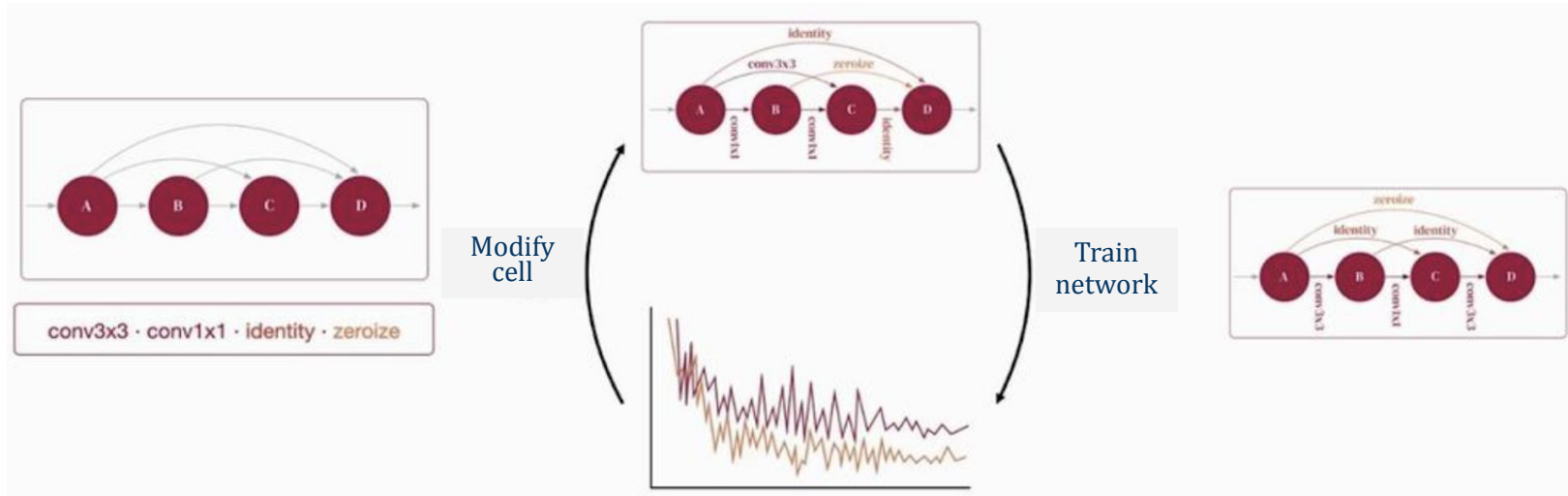
NAS spaces are composed of:

1. A cell structure
2. A list of possible operations
3. A network skeleton



# NAS requires extensive training

1. Define cell space
2. Generate architectures and train them
3. Output the best found



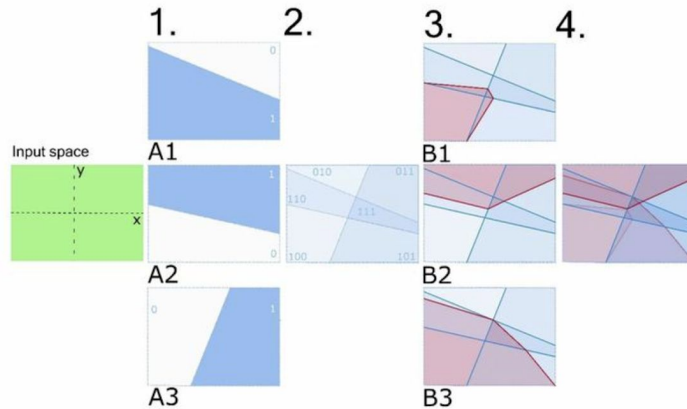
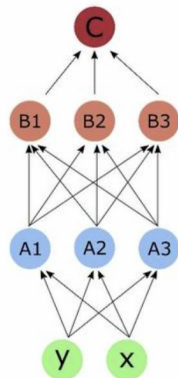
This is very expensive

## 6. Evolving Neural Architecture Search (NAS)

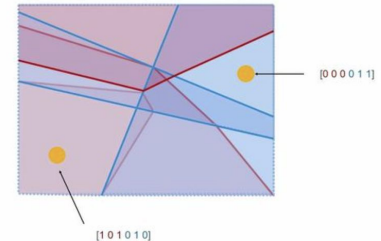
# Evolving NAS

**Neural Architecture Search without Training**, Mellor et Al. (ICML 2021)

**Investigating linear regions of networks at initialisation**



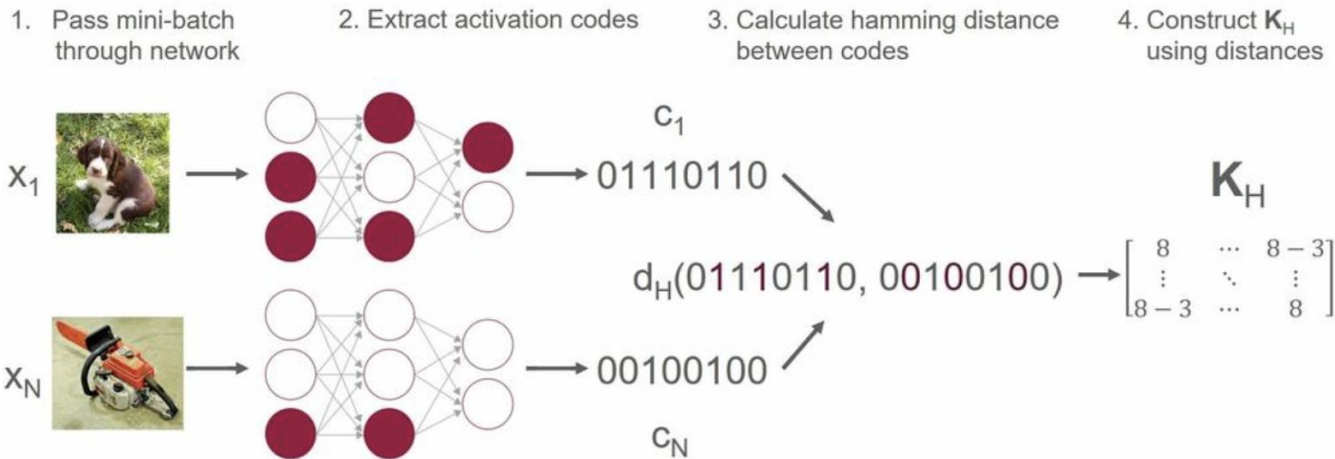
**Extracting binary codes**



# Evolving NAS

Neural Architecture Search without Training, Mellor et Al. (ICML 2021)

## Calculating $K_H$



# Evolving NAS

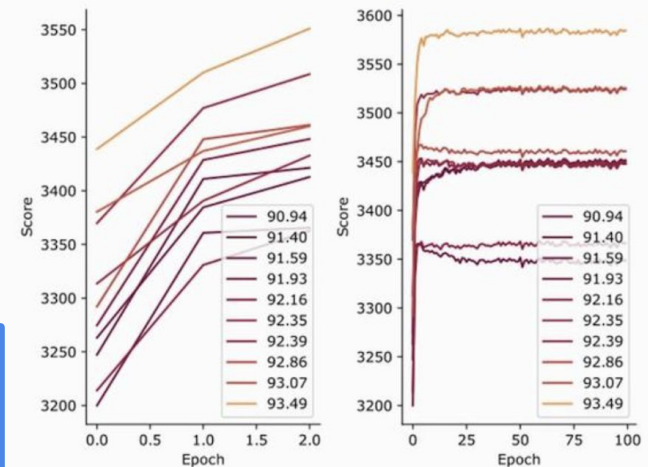
## Neural Architecture Search without Training, Mellor et Al. (ICML 2021)

Algorithm:

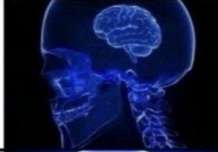
1. Randomly sample N architectures from search space
2. Score using  $\log|K_H|$ , pick top scorer
3. Takes seconds, no training required!

| Method         | Search (s) | CIFAR-10<br>Test | CIFAR-100<br>Test | ImageNet16-120<br>Test |
|----------------|------------|------------------|-------------------|------------------------|
| NASWOT(N=1000) | 248.23     | 93.10 $\pm$ 0.31 | 69.10 $\pm$ 1.61  | 45.08 $\pm$ 1.55       |
| REA            | 12000      | 93.22 $\pm$ 0.16 | 70.11 $\pm$ 0.61  | 45.54 $\pm$ 0.92       |

$\log|K_H|$  trajectories across training



# Evolving {insert fancy term}

|                                      |                                                                                     |
|--------------------------------------|-------------------------------------------------------------------------------------|
| AI                                   |  |
| ML                                   |  |
| AutoML                               |  |
| AutoML<br>without Training           |  |
| PPT about AutoML<br>without Training |  |

# AutoML vs ML: my two cents

- **Automated ML** is the process of automating the tasks of applying machine learning to real-world problems.



- AutoML allows non-experts to make use of machine learning models and techniques.

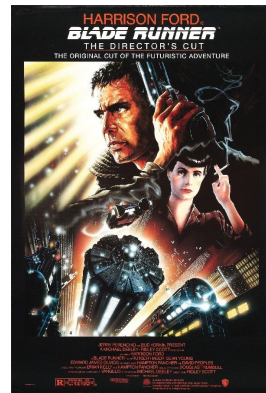


Who will do the data **pre-processing, feature engineering, extraction** and **selection**?

- Advantages: simpler and faster solutions, models that often outperform hand-designed models.



- In practice, it requires a great effort to implement.
- Need vast resources (**time** and **computational power**)
- Does not guarantee better performance than hand-designed models.

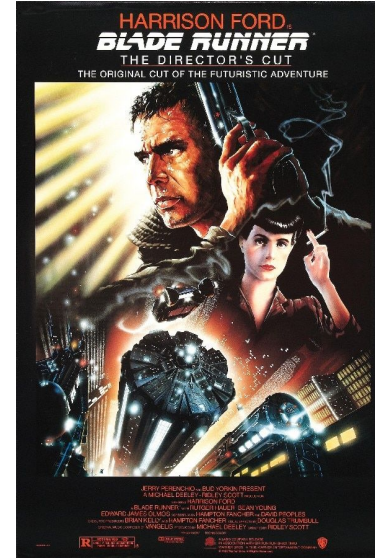


# AutoML vs ML: my two cents



81%

AUDIENCE SCORE  
50,000+ Ratings



91%

AUDIENCE SCORE  
250,000+ Ratings

## 7. Recap

# Recap

1. Introduced AI, ML and Neural Networks
2. Compared AutoML to classic ML
3. Took a realistic approach to AutoML: Neural Architecture Search (NAS)
4. Evolving NAS: Neural Architecture Search without Training (paper)
5. Challenged the ideal vision of AutoML

**If AutoML is not the only way to evolve ML,  
what can I do?**



**TECH DAYS 2021**

Evolutionary Architectures for the Next Decade



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## 8. Tips for data practitioners in industry



**TECH DAYS 2021**

Evolutionary Architectures for the Next Decade



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# Tips for data practitioners in industry

## Tip 1: Datasets merging, labels with pseudo labels

- No practitioner would train their model on only one dataset - the more, the better.
- The pre-trained model might be a viable solution; yet it also raises concerns on system load and the latency of the model inference pass.
- Explicit dataset merging is the best solution an industry practitioner would want to adopt.



# Tips for data practitioners in industry

## Tip 2: Do dataset adjustment

- Data Supplement
    - Random selection at the beginning
    - Active selection with the current model to label
    - Least money, do the best
  - Data Cleaning
    - Actively select data for review, re-label data
  - Actively Select Method
    - Importance (by losses)
    - Uncertainty (by entropy)
    - Similarity (by distance)
- **There are no fixed datasets in industry!**

# References

- [Evolving Deep Neural Networks](#) (article)
- [Automated Machine Learning](#) (wiki)
- [The Bitter Lesson](#) (article)
- [Neural Architecture Search without Training](#) (ICML 2021, paper)
- [Do we need AutoML... or AutoDM \(Automated Data Management\)?](#) (article)
- The data centric research is currently lacking (ICML 2021, presentation)

# Evolving Neural Networks

Thanks for listening