

# Neural Networks

## Applications



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## Topics

- Applications
- Challenges
  - Design
  - Interpretability
- Final comments

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## ANN applications

- ANN is already present in our daily life
  - Recommend which messages to show in our social network app
  - Filter spam from your e-mail
  - Decide which results (and ads) to show in our web search
    - and in which order
  - Suggest films or books that someone will like
  - Decide if we can have a bank loan
  - Diagnose if you have a particular disease

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## Classical ANN applications

- Learn to read aloud
  - NETtalk (Sejnowski and Rosenberg, 1986)
- Learn to recognize spoken words
  - SPHINX (Lee, 1989)
- Learn to drive a car
  - ALVINN (Pomerleau, 1989)
- Learn to play backgammon
  - TD-GAMMON (Thesaurus, 1992)

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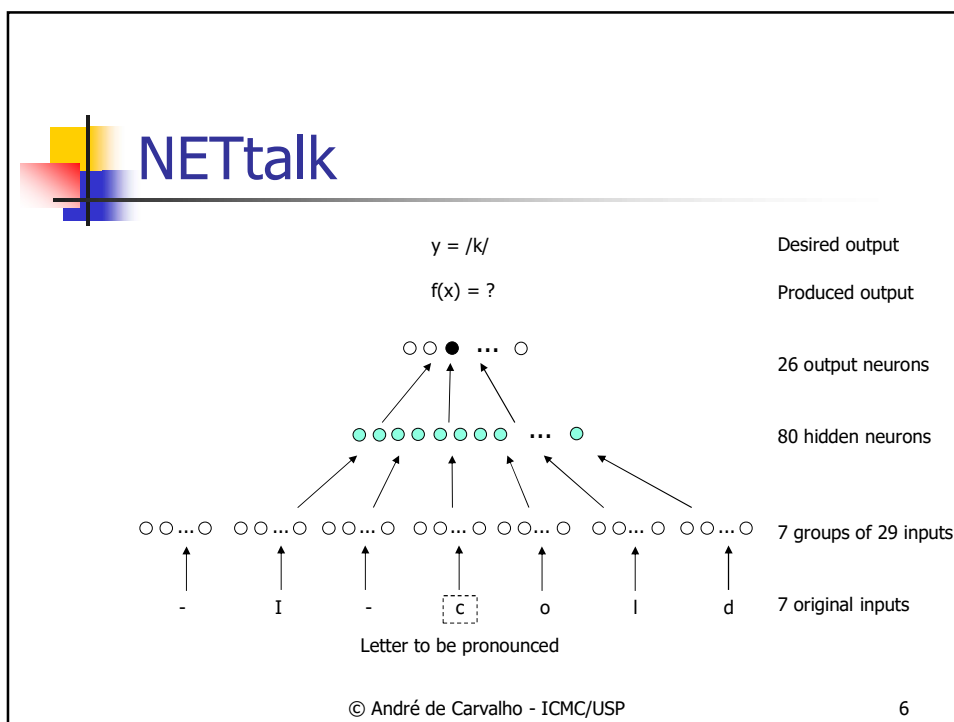
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## NETtalk


- *NETtalk: a parallel network that learns to read aloud (1986)*
  - Automatic system that learns to speak what is written in an English text
  - Maps texts in phonemes
    - Without using natural language processing or rules of phonetics
    - Uses a neural network with 1 hidden layer
      - Input layer: 7 groups with 29 inputs
      - Hidden layer: 80 neurons
      - Output layer: 26 neurons

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## NETtalk

- Input layer
  - Sliding window
  - 7 groups for the 7 text positions
    - 29 inputs = 26 letters + 3 punctuation/pause marks
- Hidden layer
  - Most neurons recognize more than one combination of letters

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## NETtalk

- Output layer
  - Code of the phoneme to pronounce the letter in the center of the input window
    - Besides silence, sound omission (syllabi), pause and end of reading
  - Output code depends only on the letters
    - Do not use information about neighbour phonemes


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# NETtalk

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# NETtalk

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- Dataset available in the UCI repository
  - 20008 English words
    - Together with their phonetic representations with stresses
  - 250 person-hours were dedicated to create and test the dataset

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## NETtalk

- Dataset has four predictive attribute for each word:
  - Letter representation
  - Phonemic representation
  - Stress and syllabic structure
  - An integer indicating foreign and irregular words

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## NETtalk

- Predictive performance
  - 95% for training data and 78% for test data
  - Compared with Dectalk
    - Rule-based expert systems developed by linguists
    - Dectalk Presented a better predictive performance
      - Developed after a decade of studies by linguists

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## Reading a poem

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The crow crooked on more beautiful and free,  
He journeyed off into the quarter sea.  
his radiant ribs girdled empty and very –  
least beautiful as dignified to see.

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## ANN poet

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- Jack Hopkins, Researcher from University of Cambridge, UK
- ANN was trained with thousands of lines of poetry
  - More than 7 million words of 20th-century English poetry
    - Most of the poems came from online books

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## ANN poet

- After trained, created several poems
- Researcher asked more than 70 people whether the poem was written by
  - Human
  - ANN
- Poem with more human votes was written by an ANN

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## Sophia




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# ALVINN



Dean Pomerleau  
CMU

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# ALVINN

- *Autonomous Land Vehicle In a Neural Network (1989)*
  - Autonomous car based on neural networks
    - CMU PhD Thesis
  - Input received by a camera on top of the vehicle
  - Travelled 4500 km at 110 Km/h, coast to coast, in American public roads
    - Except for 80 Km

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# ALVINN



<https://www.youtube.com/watch?v=r4JrcVEkink>

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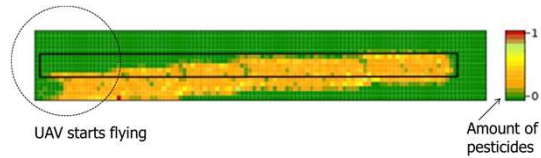
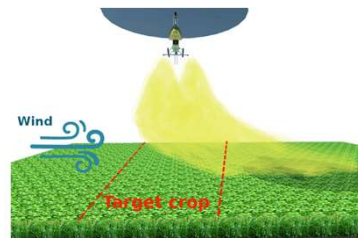
# Autonomous vehicles



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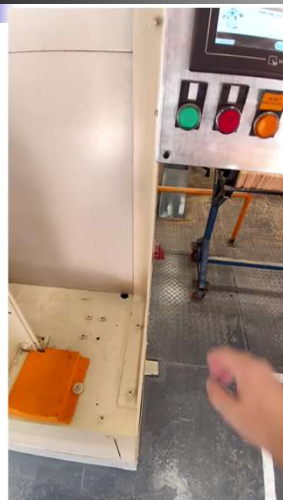
## Pesticide spread in a crop field



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## Wood classification



- Prototype of system to classify wooden board
- 100 boards
  - 50% good and 50% bad
    - 14 types of fault
  - 150 boards/minute or 27,7 meters/minute
  - Predictive accuracy: 97%
    - Human operator predictive accuracy: 93%

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## ANN architecture size

- Large networks:
  - Learning can be faster
  - Minimum local is easier to avoid
  - Solution space is less restricted
  - Fault tolerance is higher
- Small networks:
  - Generalization may be better
  - Require less computational resources
  - Knowledge extraction is easier

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## ANN Architecture design

- ANN architecture usually has a strong impact on predictive performance
  - Neither too large, nor too small
- Alternatives
  - Train and error
  - Constructive
  - Pruning
  - Evolutionary

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## Explainable AI (XAI)

- ML algorithms induce models with different levels of interpretation
  - Black box
    - Difficult to interpret
  - White box
    - Easy to interpret
    - Explain how decisions are made
  - Gray box

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## Explainable AI

- Although ANNs have been successfully employed in several problems
  - They still suffer from significant limitations
- One of them is the lack of comprehensibility of what they learned
  - How they make decisions
  - Essential in applications that affect human life
    - GDPR and LGPD
  - There are methods for knowledge extraction from ANNs

ICCIMA 2001

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## Explainable AI

- The methods for extraction knowledge from ANNs have some restrictions:
  - Distributed learning representation
  - Some methods work only with features with discrete values
  - Difficulty of their application to large networks

ICCIMA 2001

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## Data Age

### Tech Giants Are Paying Huge Salaries for Scarce AI Talent

# Newsweek

[Tech & Science](#)
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#### Artificial Intelligence: Salaries Heading Skyward

While the average salary for a Software Engineer is around \$100,000 to \$150,000, to make the big bucks you want to be an AI or Machine Learning (Specialist/Scientist/Engineer)...

**NEWSWEEK MAGAZINE**

### Need a Job? Why Artificial Intelligence Will Help Human Workers, Not Just Robots

## Surprise! Machine learning jobs are high-paying and in-demand

By [Ryan Daws](#)   
 Editor of AI News. A gadget lover, music purveyor, and ex-host of a consumer technology show.  
 Posted on March 15, 2019


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#### Businesses Expect AI To Have A Positive Impact On Growth, Productivity, Innovation, Job Creation: The Economist - Microsoft Study

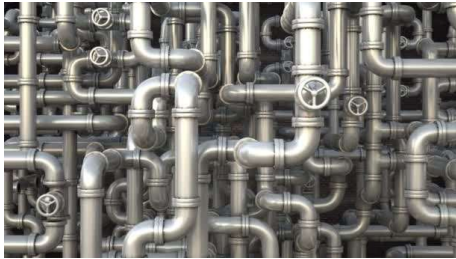
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# End-to-End Machine Learning


They are not independent

<p>Modify ML algorithm</p> <p>Oversample/ Undersampling</p> <p>Re-code Features</p> <p>Feature selection</p>		<p>Tune hyper-parameters</p> <p>Deal with missing values</p> <p>Check overfitting</p> <p>Discover bugs</p>
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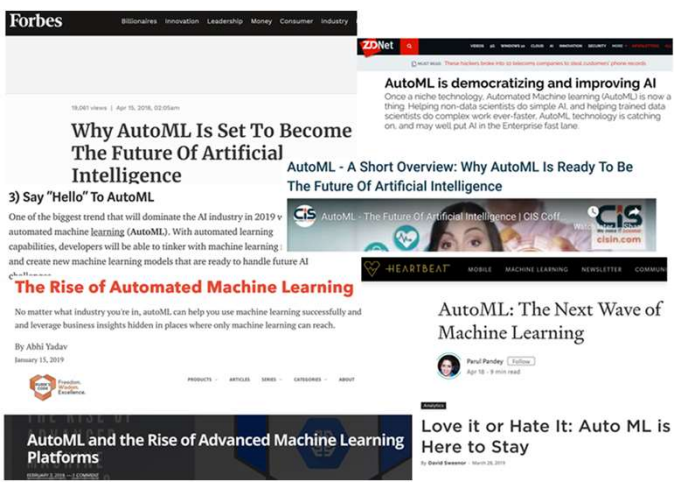
Adapted from Rick Caruana, Research opportunities in AutoML Microsoft Research

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# AutoML



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## AutoML tools

Auto-WEKA: Wizard, Experiment Builder, Experiment Runner, Trained Model Runner, WEKA

Auto-Sklearn

DataRobot

Amazon Rekognition

TPOT

H<sub>2</sub>O.ai

Cloud AutoML Vision

AutoML.org Freiburg

clarifai

CreateML Apple

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## AutoML Books

Automated Machine Learning: Methods, Systems, Challenges (Springer)

Hands-On Automated Machine Learning (O'Reilly)

Practical Automated Machine Learning on Azure (O'Reilly)

AutoML: Standard Requirements, Practical Tools for Self-Assessment (The Art of Service)


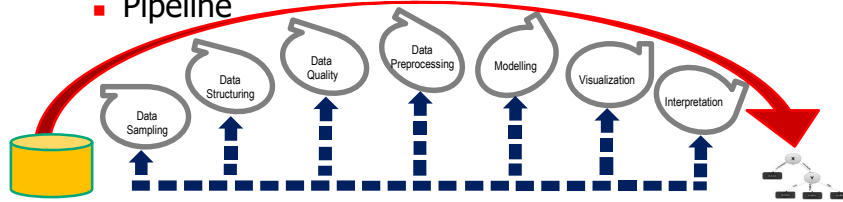
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## Pajé AutoML


- End-to-end AutoML (AutoDS)
- Main focus
  - Data pre-processing
  - Expandable
  - Explainable ML
  - Pipeline


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## Pajé Timeline (PhD Thesis)



<u>Ricardo Sovat</u> , Uma Abordagem Híbrida Baseada em Casos e Redes Neurais. Uma aplicação: escolha e configuração de modelos de redes neurais.	<u>Mariá C. V. Nascimento</u> , Meta-heurísticas para o problema de agrupamento de dados em grafo.	<u>Rodrigo C. Barros</u> , Automatic design of decision tree induction algorithms.	<u>Davi P. Santos</u> , Seleção e controle do viés de aprendizado ativo.		
<u>Estéfane G. Lacerda</u> , Model Selection of RBF Networks via Genetic Algorithms.	<u>Bruno F. de Souza</u> , Meta-aprendizagem aplicada à classificação de dados de expressão gênica.	<u>André L. D. Rossi</u> , Meta-aprendizado aplicado a fluxos contínuos de dados.	<u>Luis Paulo Garcia</u> , Noise detection in classification problems.	<u>Rafael G. Mantovani</u> , Use of meta-learning for hyperparameter tuning of classification problems.	
<u>Claudia Regina Milaré</u> , Extração de Conhecimento de Redes Neurais Artificiais.					



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**Support**

AMDA  
Machine Learning  
in Data Analysis

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Analytics  
Massive Data Analysis Laboratory

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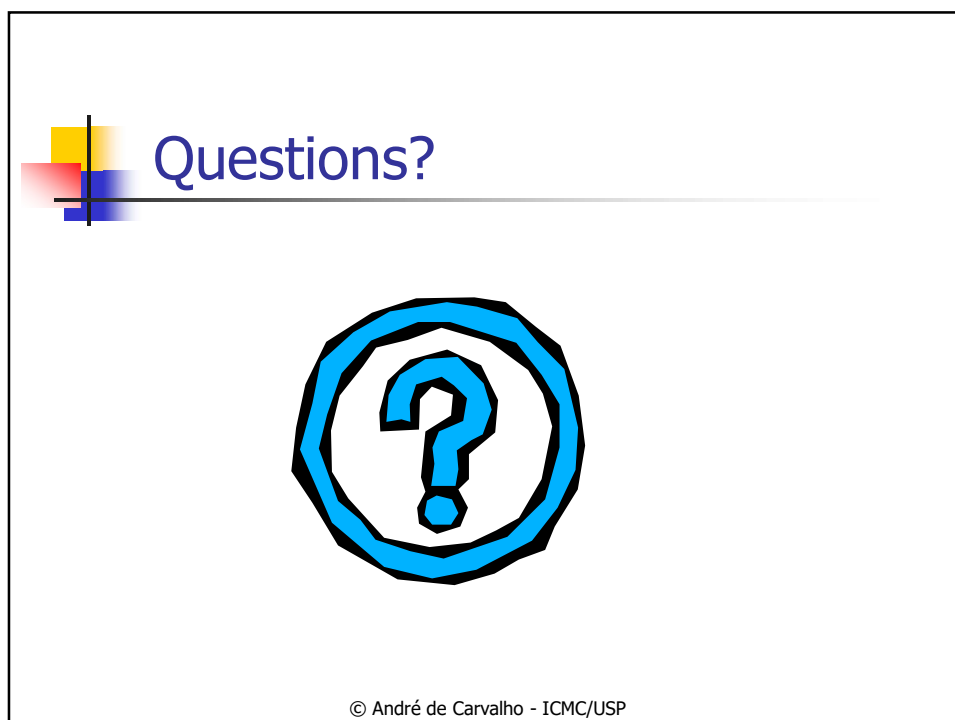
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**Questions?**

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