

Deep learning and transit crew-schedule optimization

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2018 Fare Collection/Revenue Management
& TransTech Conferences

GIRO at a glance

Founded in
1979

Based in
Montréal

400+
skilled
employees



GIRO

Integrated software solutions for planning & managing transport-related operations



Public transport
(*HASTUS™* &
HASTUS-Rail™)

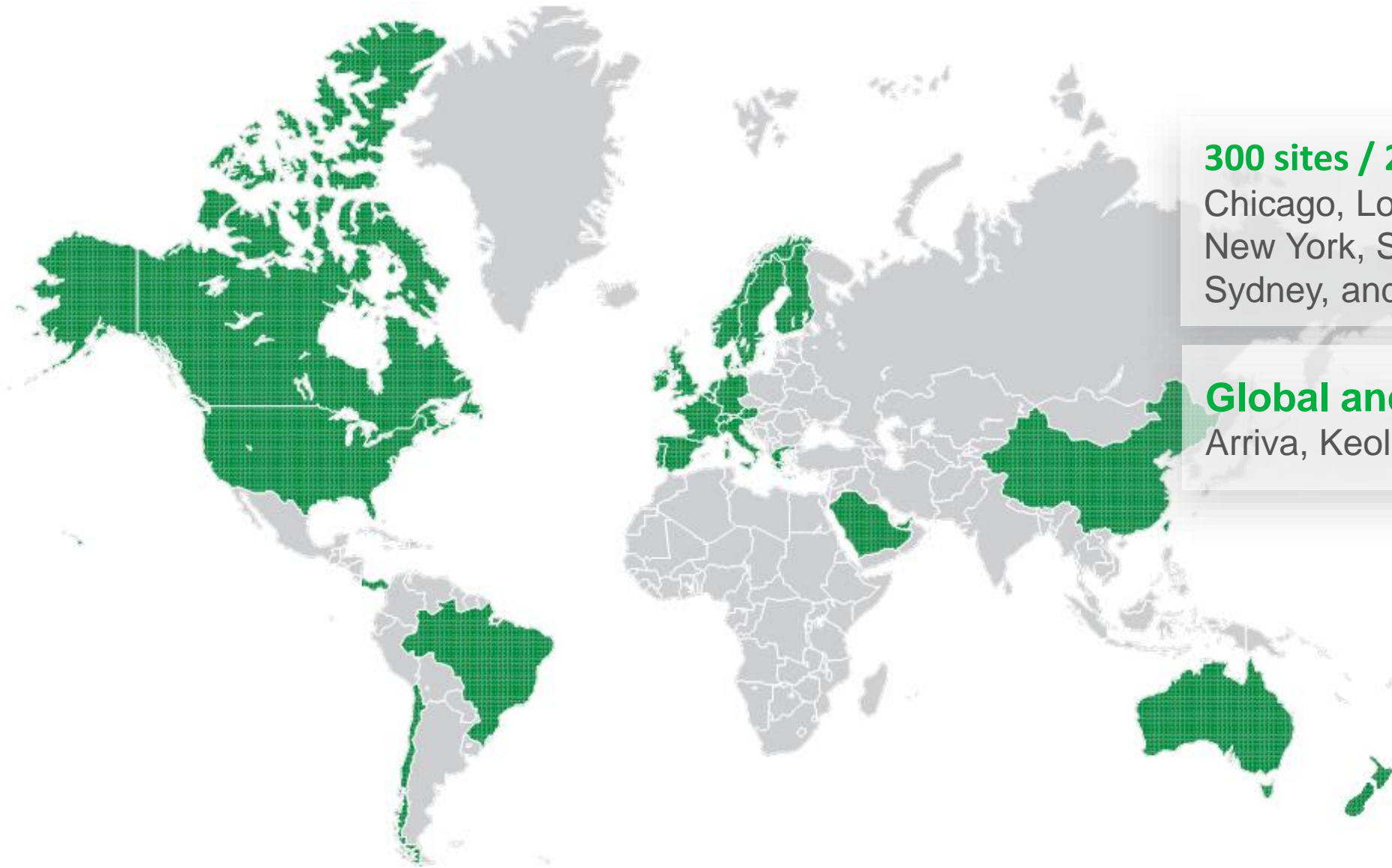


On-demand transport
(*HASTUS-OnDemand™*)

Operations research and optimization

- At the root of our company
- Still at the core of our products

Our global market presence



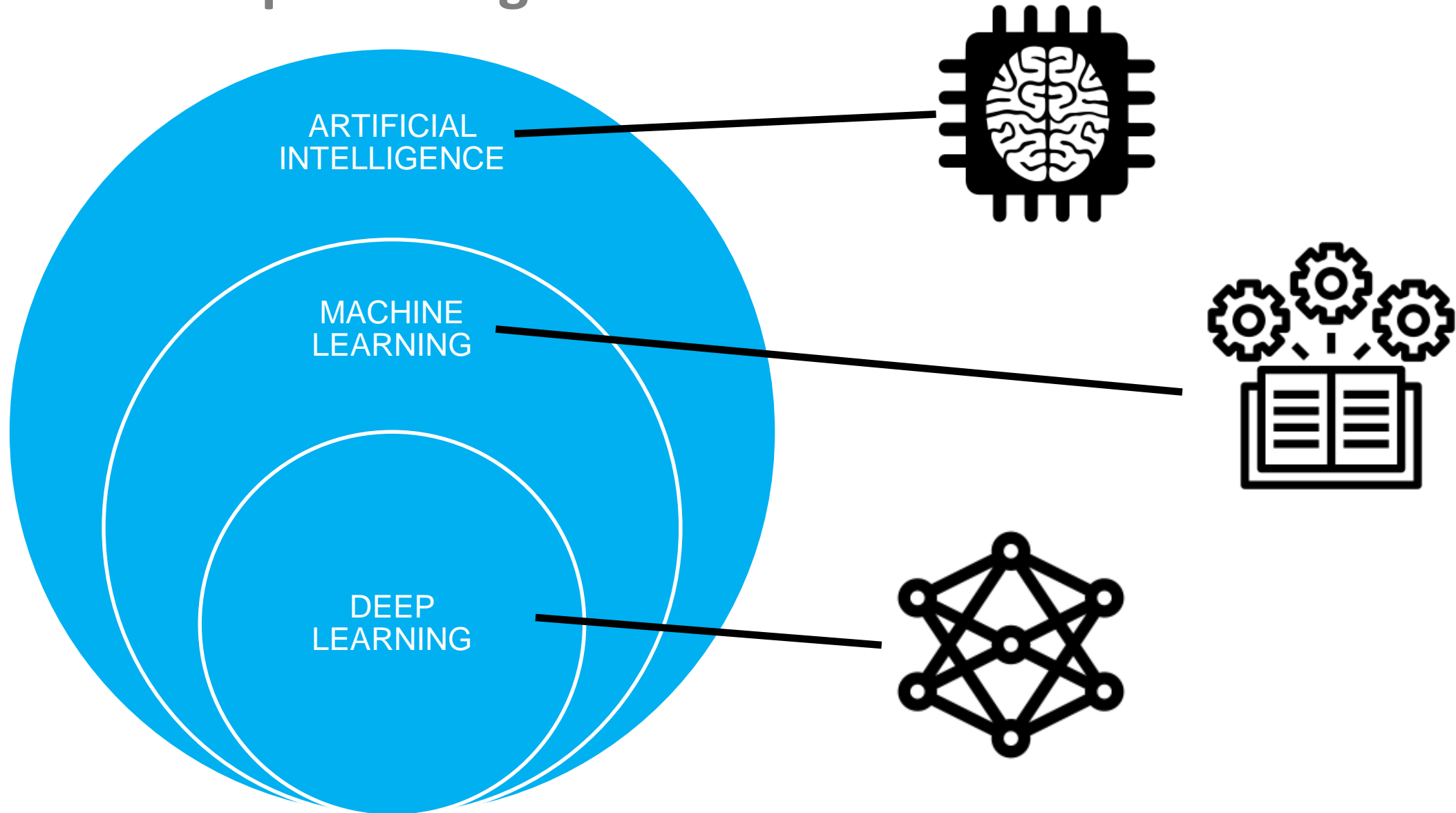
300 sites / 26 countries

Chicago, Los Angeles, Montréal,
New York, Stockholm, Singapore,
Sydney, and more

Global and local organizations

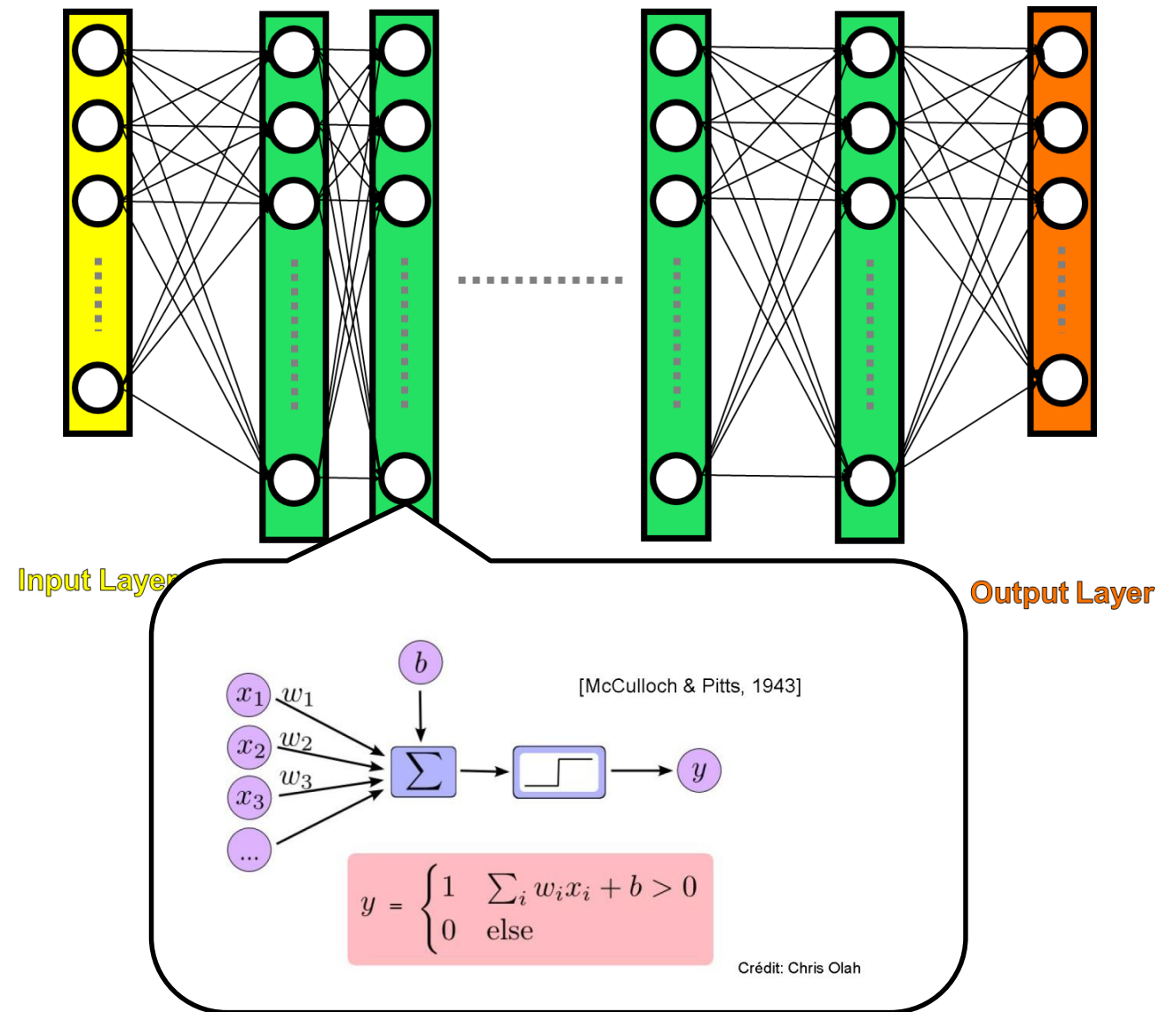
Arriva, Keolis, Transdev, and more

What is deep learning?



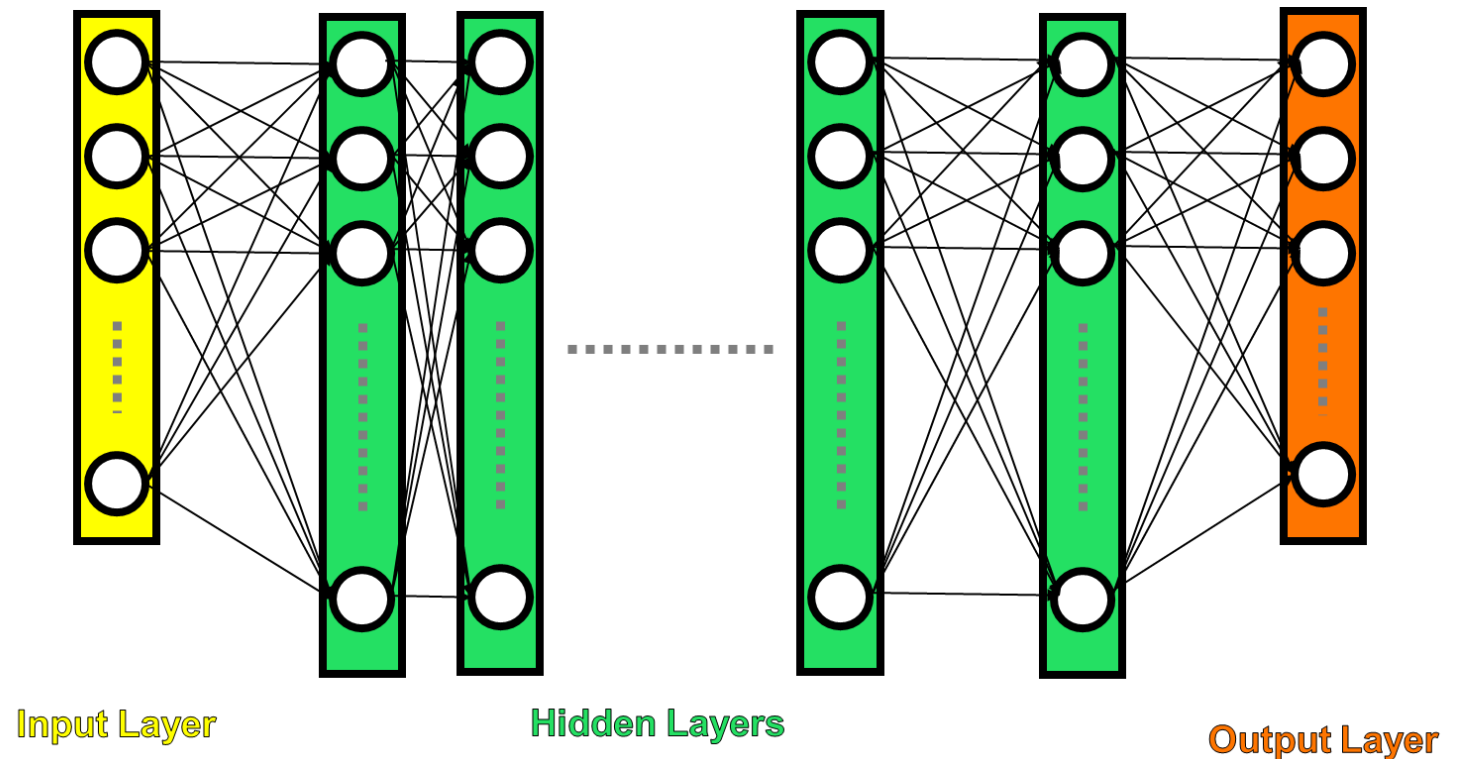
What is deep learning?

- ▶ Inspired by the human brain!
- ▶ A type of machine learning that uses deep levels of artificial neural network layers

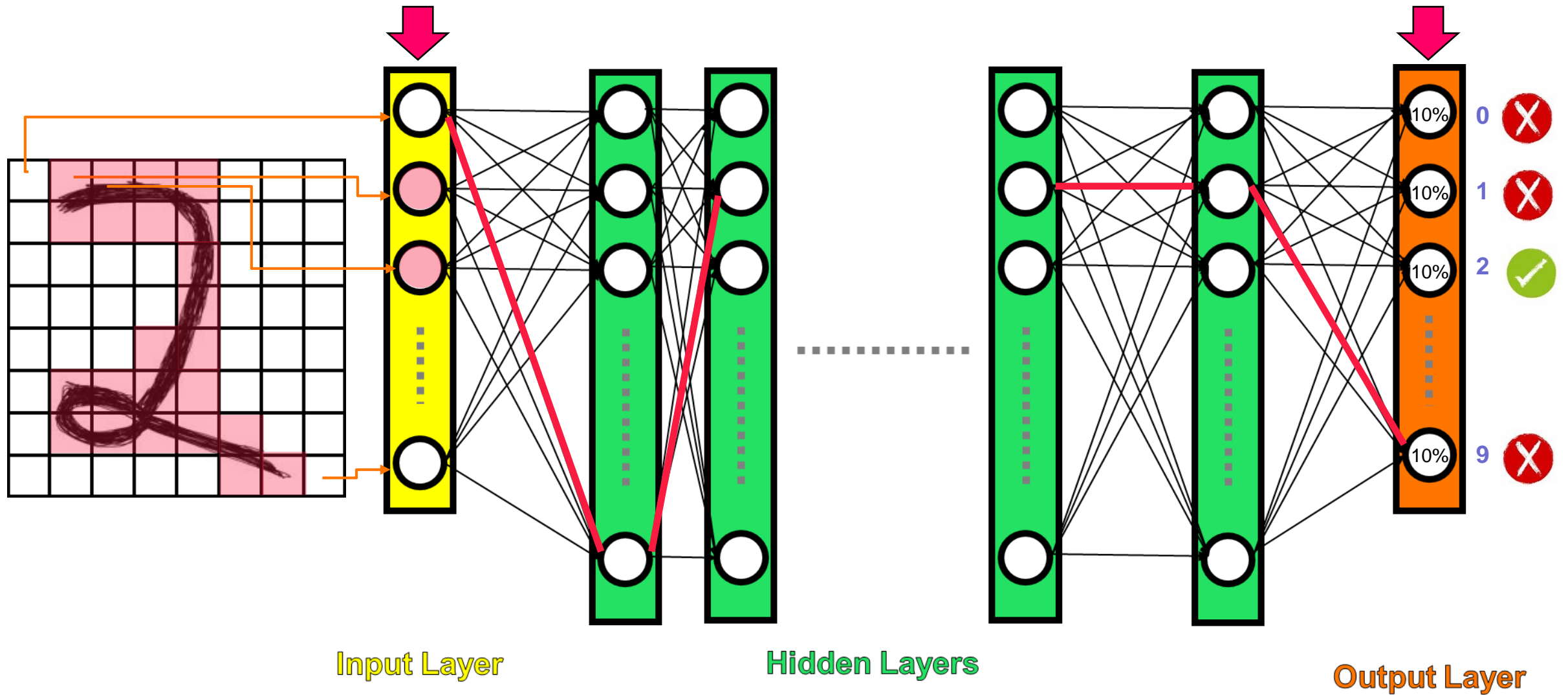


Two phases

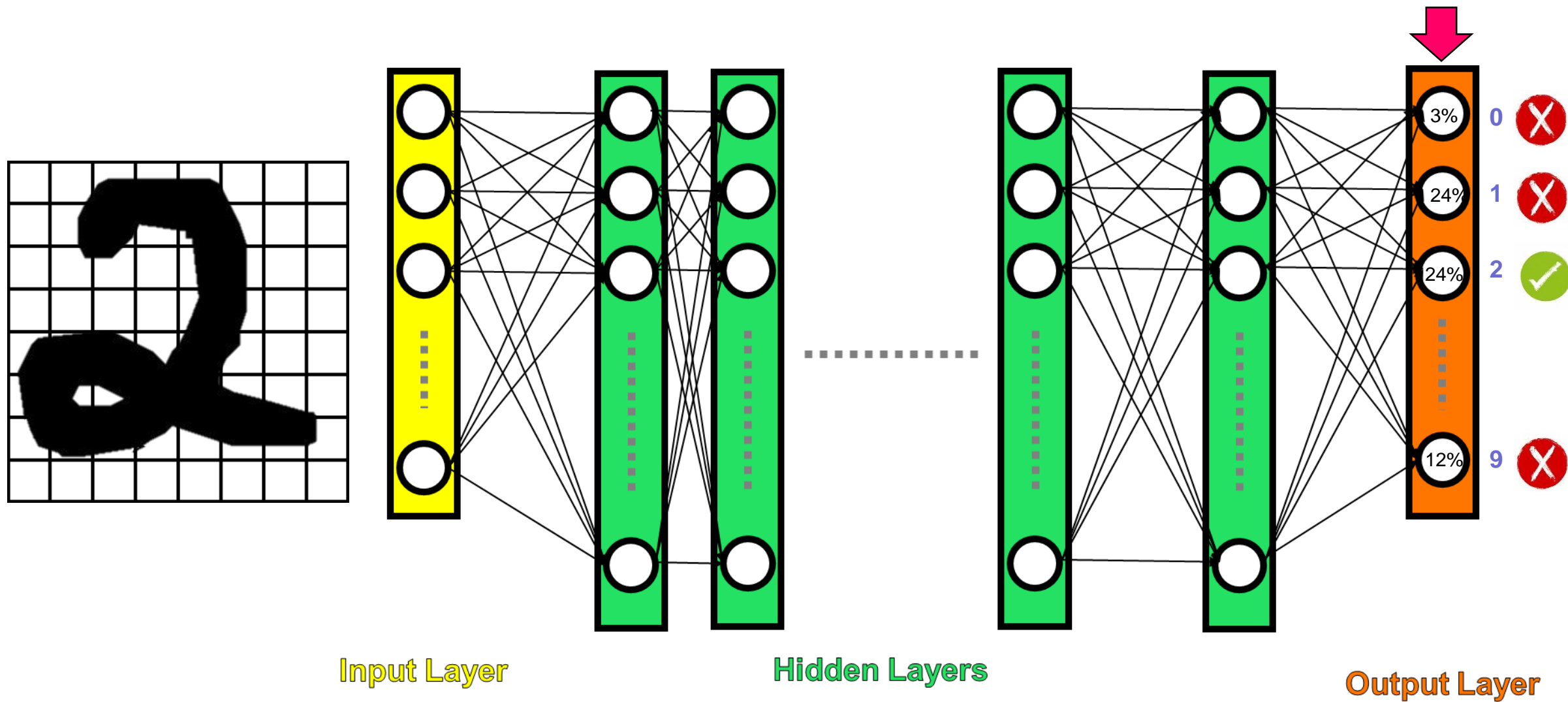
- ▶ Training (supervised)
 - ▶ A lot of data
 - ▶ A lot of computer power (GPU and now TPU processors), allowing for deep networks
 - ▶ The longer of the two phases
- ▶ Prediction
 - ▶ Fast



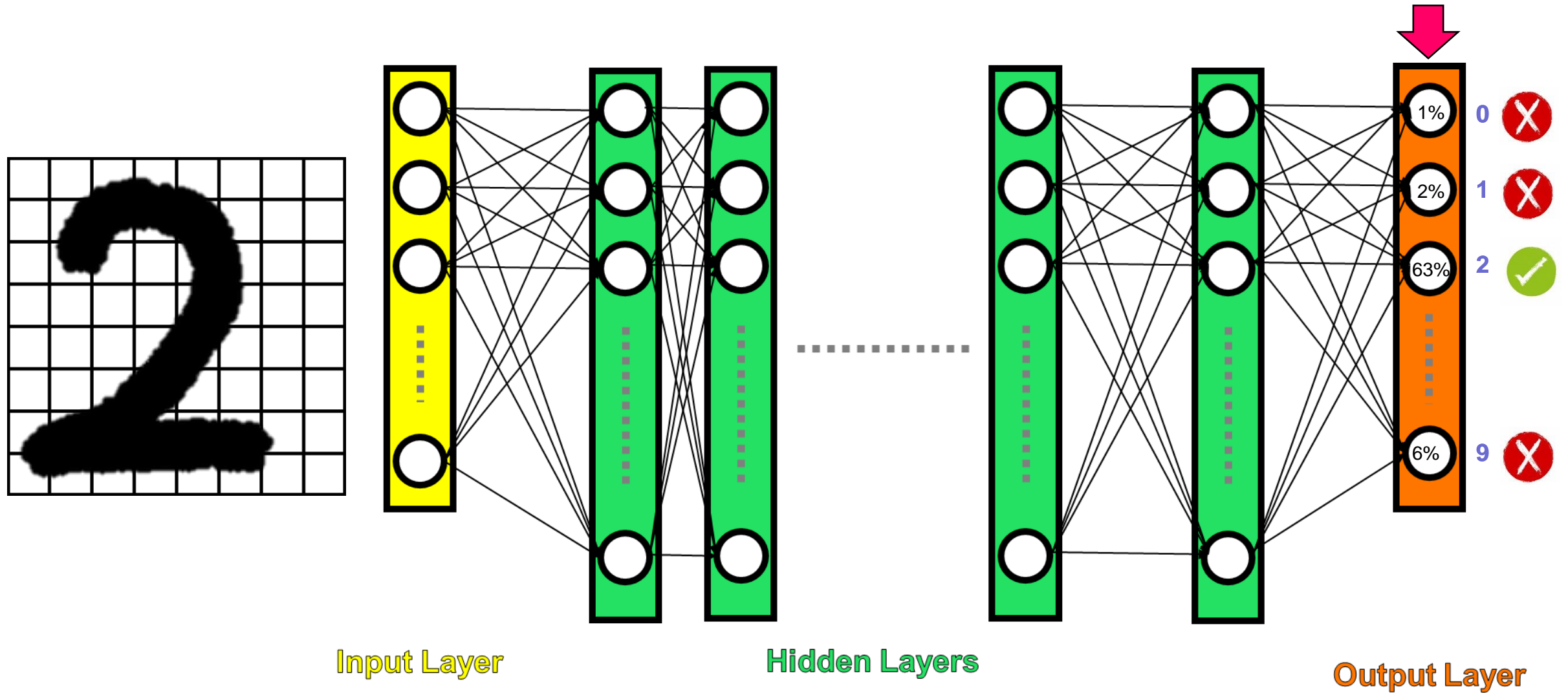
Two phases



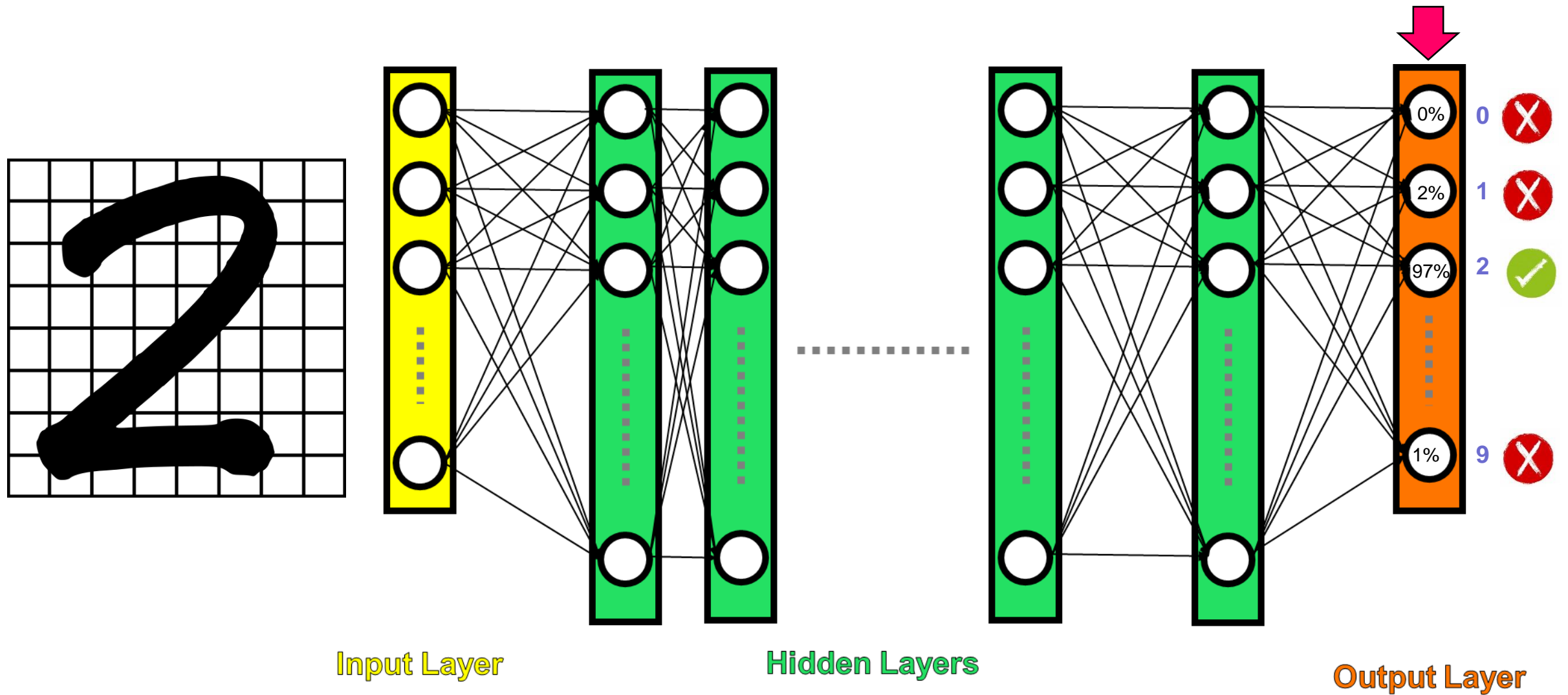
Two phases



Two phases



Two phases



Deep learning challenges

▶ Implementation challenges

- ▶ What is the input layer?
- ▶ What is the output layer?
- ▶ How many hidden layers?
- ▶ How many neurons per layer?
- ▶ What are the weights on the arcs?

▶ Several open source platforms available

- ▶ **theano** University of Montreal (MILA/Yoshua Bengio)

- ▶  Facebook

- ▶  Google

Deep learning: does it work?

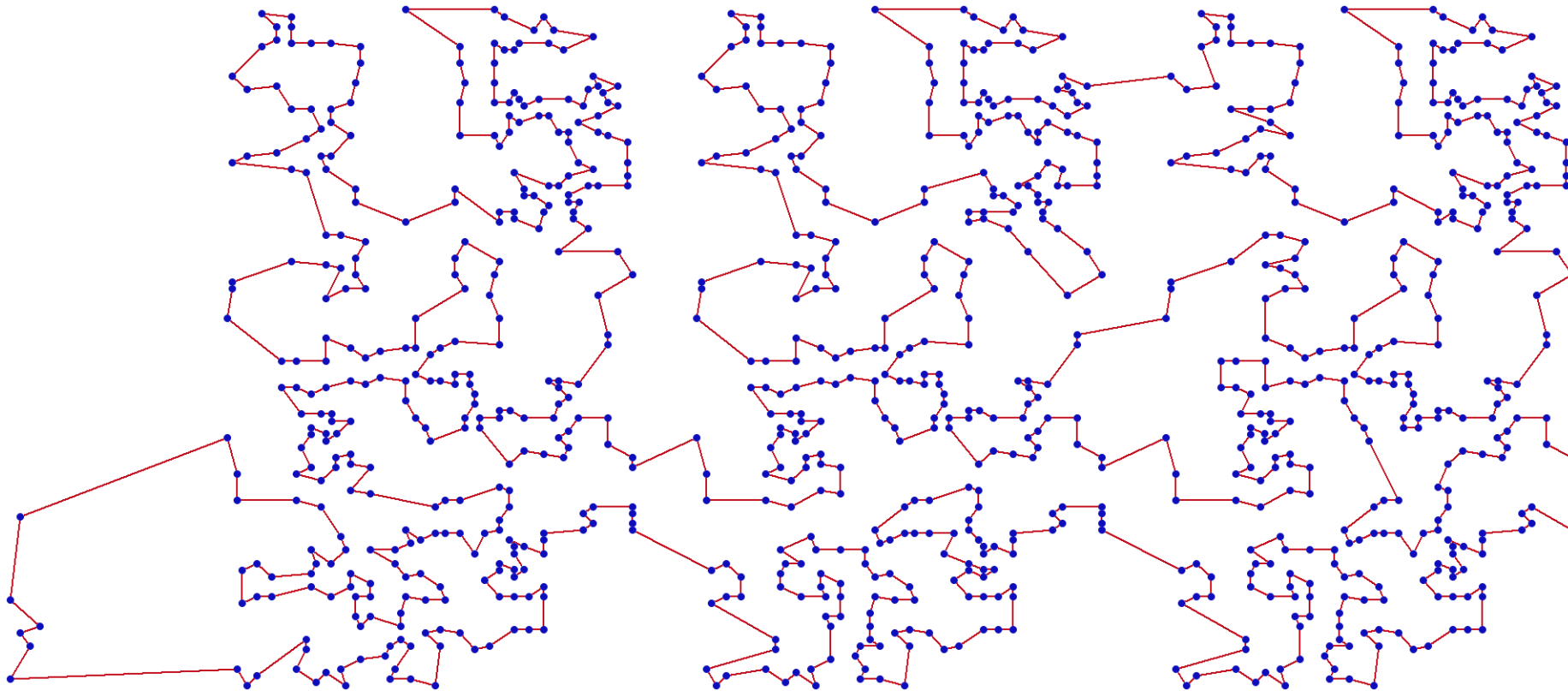
▶ Yes for

- ▶ Spam detection
- ▶ Speech recognition
- ▶ Translation
- ▶ Facial recognition
- ▶ Driving
- ▶ Chess playing
- ▶ Credit card fraud detection
- ▶ Image caption generation
- ▶ ...



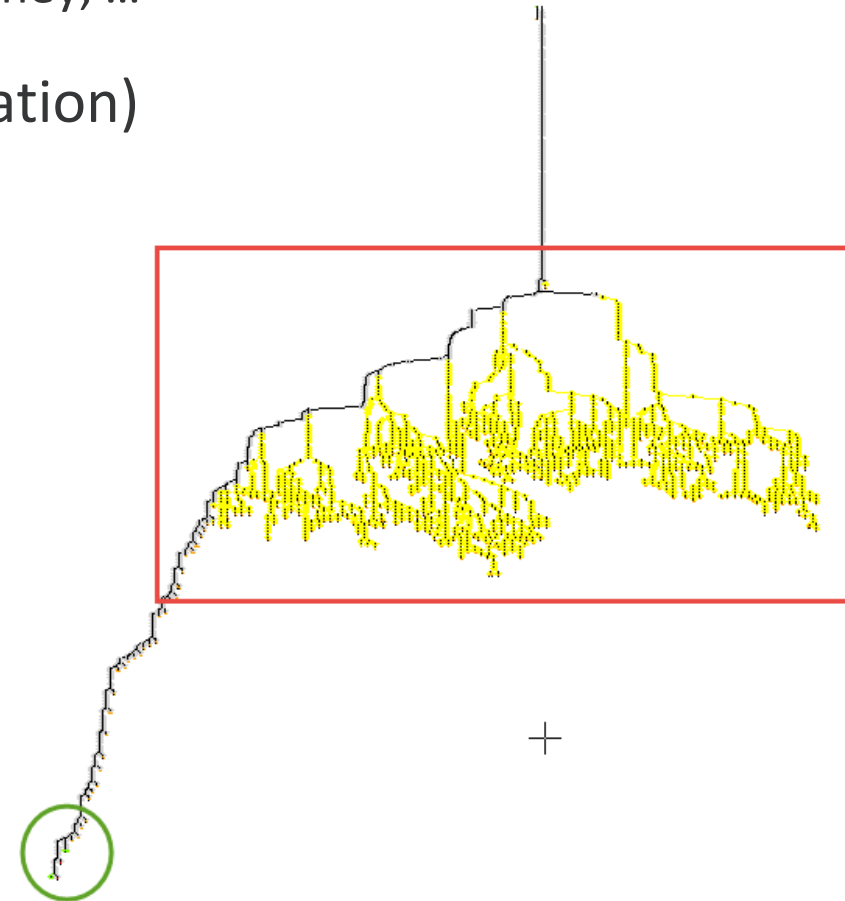
Deep learning and combinatorial optimization

- ▶ Not very efficient when directly applied
- ▶ For example TSP (Traveling Salesman Problem)

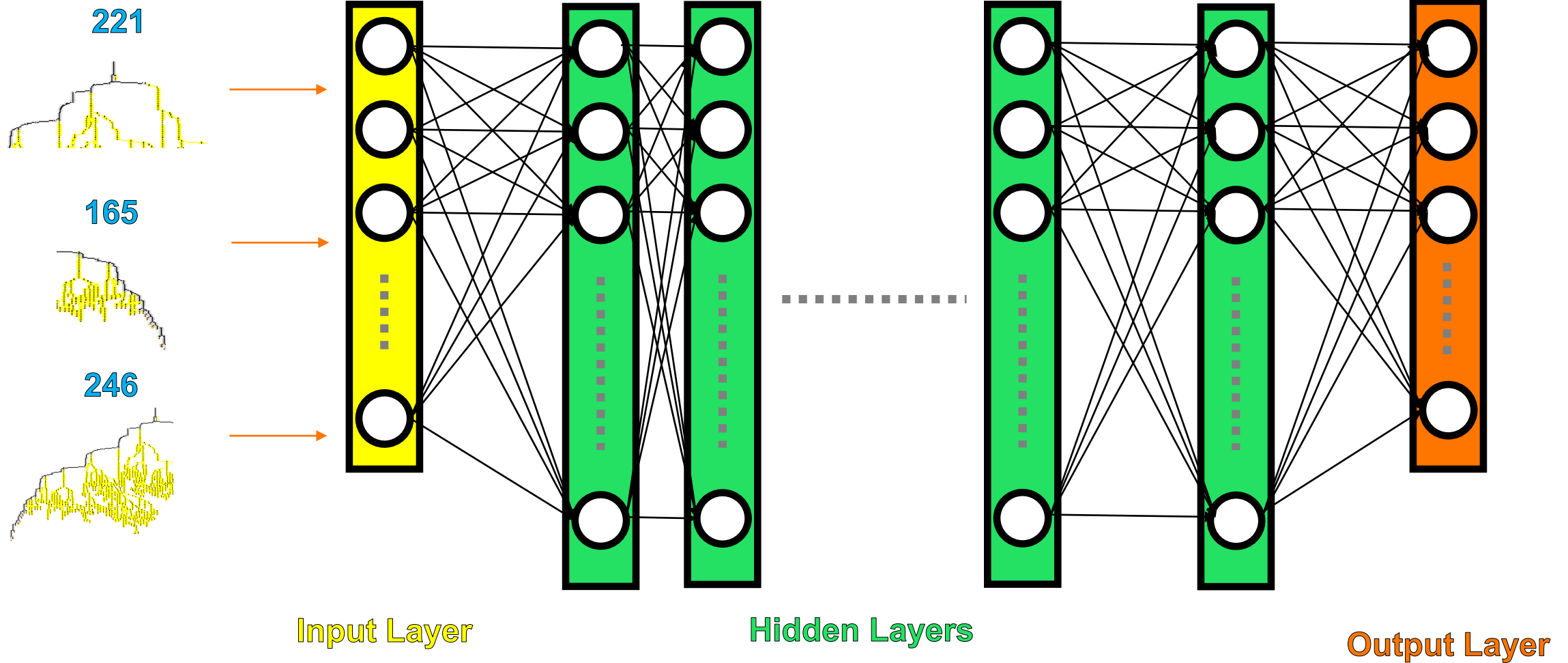


The *HASTUS/CrewOpt* case

- ▶ Optimization algorithm for transit crew scheduling
 - ▶ Installed at 186 sites worldwide
 - ▶ NYC, Chicago, LA, Barcelona, Oslo, Stockholm, Hong Kong, Sydney, ...
- ▶ Based on operations research techniques (column generation)
- ▶ **Involves several heuristic decisions**
 - ▶ Branching
 - ▶ Arc sampling
 - ▶ Resources and dominance
 - ▶ ...
- ▶ Semi-supervised deep learning can help
 - ▶ Learn to make better heuristic decisions
 - ▶ Learn features of good solutions



Deep learning and combinatorial optimization



The *HASTUS/CrewOpt* case

- ▶ Ongoing research project
 - ▶ IVADO partnership (ivado.ca)
- ▶ Expected improvements
 - ▶ Solution cost (\$\$\$)
 - ▶ Running times (once trained)

IVADO

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THE INSTITUTE FOR DATA VALORIZATION: A SCIENTIFIC AND ECONOMIC DATA SCIENCE HUB

Yoshua Bengio named 'Scientist of the Year' by Radio-Canada

For his outstanding contribution to the advancement of artificial intelligence in Canada.

CLOSE TO 1000 SCIENTISTS

#1 IN PUBLICATIONS WORLDWIDE

OVER 40 ACADEMIC PROGRAMS

SOLID PARTNERSHIPS

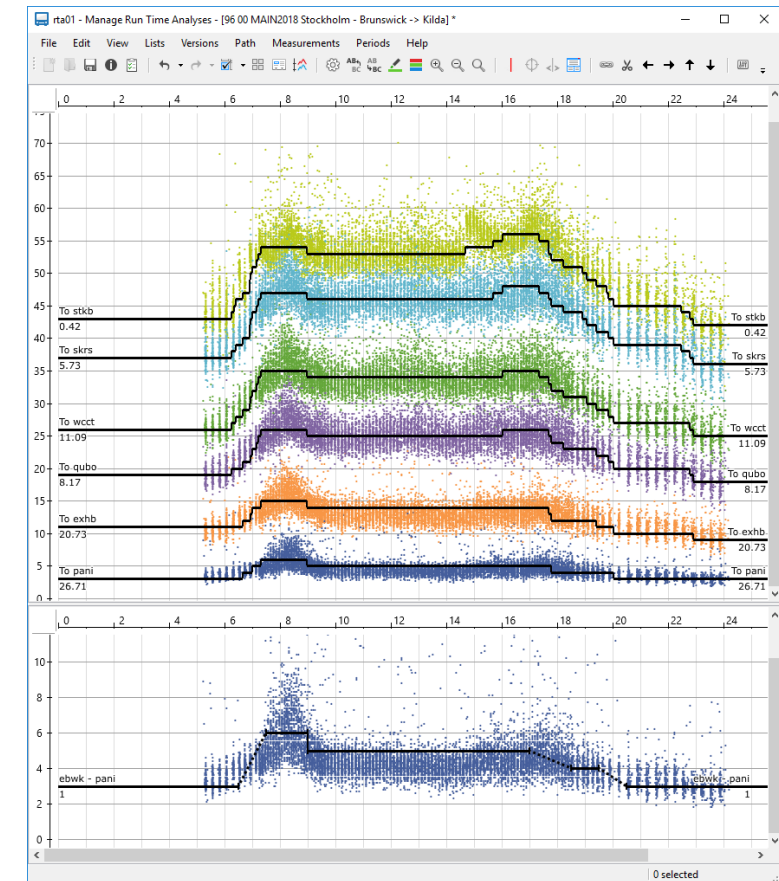
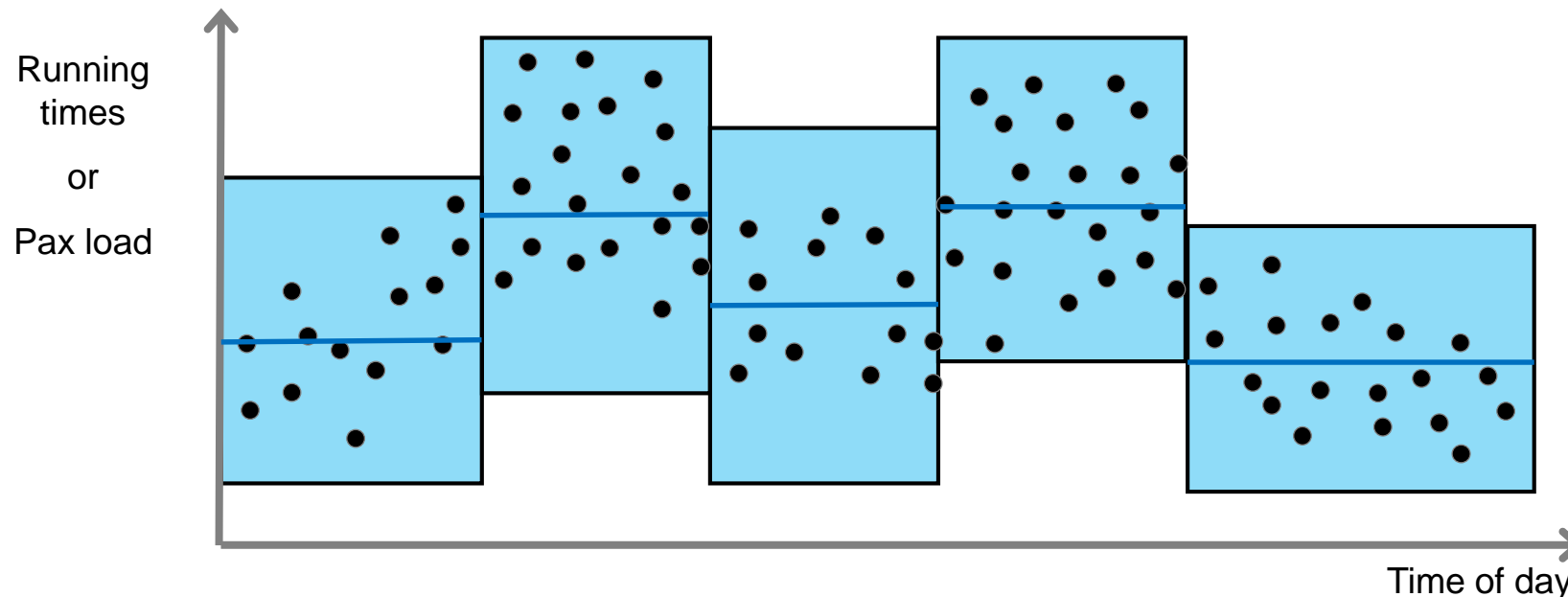
Other deep learning applications

► Unsupervised learning

► Useful to identify data clusters

► Example

- *Clustering* to identify higher values for [April 20, May 17]
- Measures should be removed from analyses

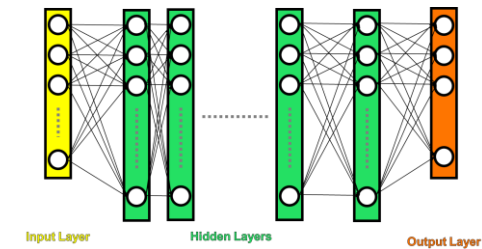


Other deep learning applications

► Forecasting (e.g. absenteeism)

► Training

Input								Output
Date	Day	°F	Start time	End time	Spread	Route	Gender	Absent
6/2/2017	Monday	28	6:47	15:08	8h21	168	Male	1
6/2/2017	Monday	28	7:39	16:02	8h23	51	Female	0
7/2/2017	Tuesday	33	9:20	19:08	9h48	12	Male	0
7/2/2017	Tuesday	33	6:47	15:08	8h21	168	Male	0
8/2/2017	Wednesday	26	14:00	23:58	9h58	121	Female	1
8/2/2017	Wednesday	26	5:30	13:12	7h42	51	Male	0
9/2/2017	Thursday	32	9:00	17:00	8h00	129	Male	0
9/2/2017	Thursday	32	8:37	16:08	7h31	168	Male	0
10/2/2017	Friday	35	6:47	20:08	13h21	51	Male	0



► Prediction

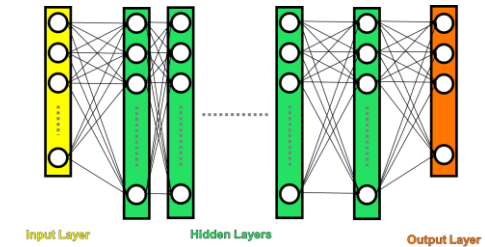
5/2/2018	Monday	27	6:52	15:18	8h26	168	Male	0.21
6/2/2018	Tuesday	38	8:17	14:54	6h37	51	Female	0.03
22/6/2018	Friday	22	6:08	20:02	13h54	168	Male	0.46

Other deep learning applications

► Forecasting (e.g. demand prediction)

► Training

Input									Output
Date	Day	°F	Route	Stop	Rain	Football	Snow (in)	Time	Demand
6/2/2017	Monday	36	51	1012	Yes	No	0	7:12	8
6/2/2017	Monday	36	51	1015	Yes	No	0	8:15	12
7/2/2017	Tuesday	29	121	2103	No	No	5	13:13	10
7/2/2017	Tuesday	29	129	1050	No	No	5	12:14	6
8/2/2017	Wednesday	34	168	1234	No	No	0	20:15	1
8/2/2017	Wednesday	34	12		No	No	0	6:54	9
9/2/2017	Thursday	38	121	0113	No	Yes	0	9:14	3
9/2/2017	Thursday	38	129		No	Yes	0	11:13	2
10/2/2017	Friday	41	51	1124	Yes	No	0	8:00	15



► Prediction

5/2/2018	Monday	27	51	1012		No	8	7:14	7
6/2/2018	Tuesday	31	129	1050		No	0	9:17	4
22/6/2018	Thursday	39	168	1234		Yes	0	16:43	15




Deep learning can be very efficient for tasks executed by humans

For combinatorial optimization problems

- ▶ Can help to improve specialized heuristic algorithms
- ▶ Ongoing work with IVADO

Stay tuned, the future is now!



Thank you!
Any questions?

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Improving efficiency at every turn

