



Introduction



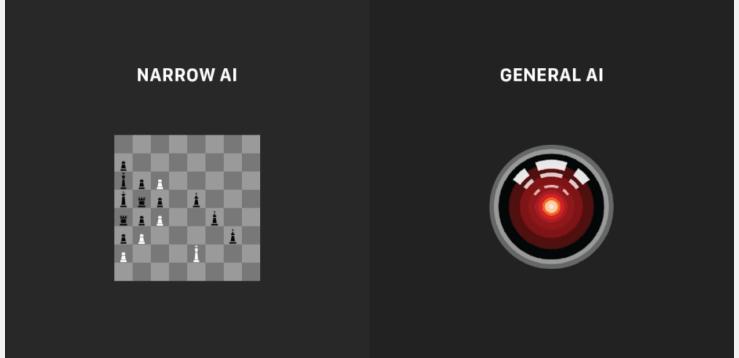
- Al is creating tremendous amounts of value in every industry
- In sectors such as retail, travel, transportation, automotive, materials, manufacturing and so on.





Introduction

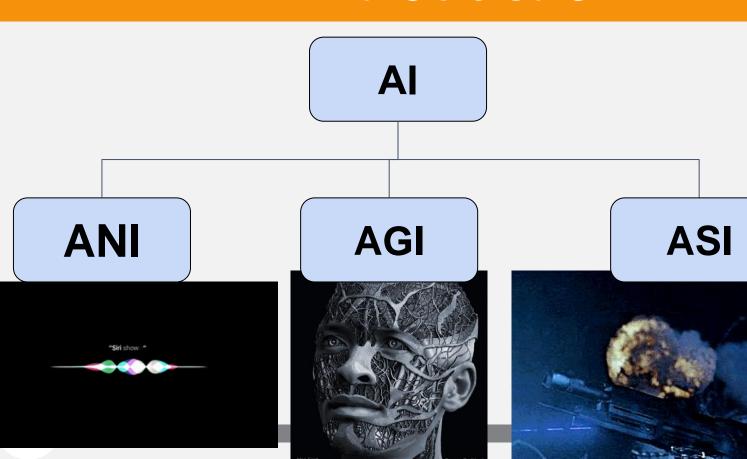






Introduction





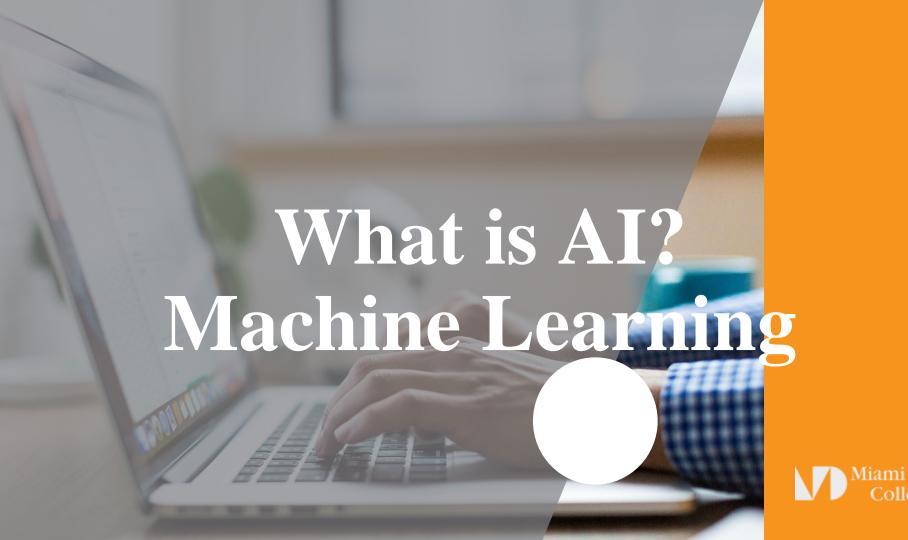
What you'll learn



What is Data Analytics/Data Science

- Machine Learning
- Data
- AI Organizations
- What can you really do with this?
- Deep Learning



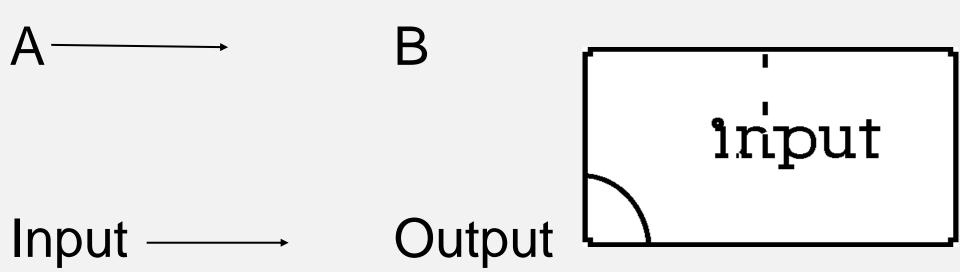


Machine Learning

- The rise of AI has been largely driven by one tool in AI:
 - machine learning.
- In this you'll learn what machine learning is, so that by the end, you will start thinking how machine learning might be applied to a business



Supervised Learning

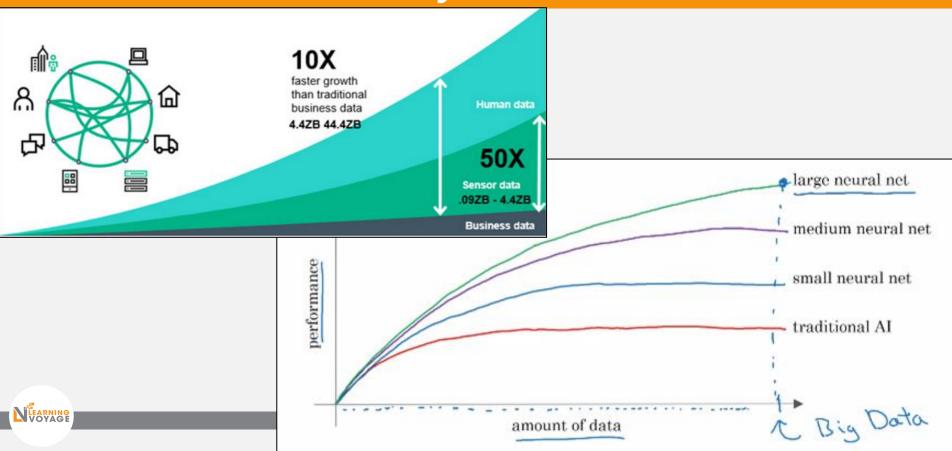




Machine Learning

Input → Process → Output	Use Cases
User Information → MODEL → Click?	Social Media. Online Advertising
email → Model → Spam	Spam Filter
Video → Model → Transcript	Speech Recognition
Shopping History → Model → Offers	Recommender Systems
English → Model → Spanish	Speech Translation
Product Image → Model → Defective	QA
English → Model → Spanish	Speech Translation QA PROCESS O

Why Now

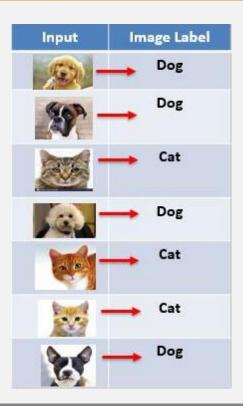






Example of Table Of Data (dataset)

Age	Income	Loan Status		
21	20000	Rejected		
37	55000	Approved		
29	35000	Approved		
23	17000 Rejected			
34	70000	Approved		
47	84000	Rejected		
25	30000	Approved		





Acquiring Data

- Manual labeling







not



cat



ot

- From observing behaviors

user ID	time	price (\$)	purchased	
4783	Jan 21 08:15.20	7.95	yes	
3893	March 3 11:30.15	10.00	yes	
8384	June 11 14:15.05	9.50	no	
0931	Aug 2 20:30.55	12.90	yes	

machine	temperature (°C)	pressure (psi)	machine fault	
17987	60	7.65	N	
34672	100	25.50	N	
08542	140	75.50	Y	
98536	165	125.00	Y	

Download from websites / partnerships







Data is Messy

- Garbage in, garbage out
- Data problems
 - Incorrect labels
 - Missing values
- Multiple types of data

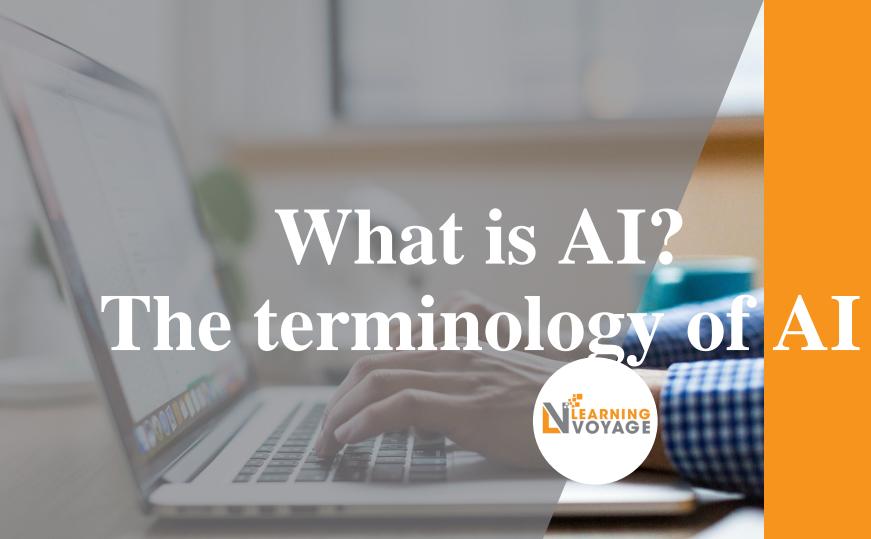
images, audio, text

unstructured

size of house (square feet)	# of bedrooms	price (1000\$)		
523 645 708	1 1 unknown	0.001 210		
1034 unknown 2545	4 unknown	unknown 355 440		







The terminology of AI



- You might have heard terminology from AI, such as machine learning or data science or neural networks or deep learning.
 What do these terms mean?
- In this, you'll see what is this terminology of the most important concepts of AI, so that you will speak with others about it and start thinking how these things could apply in your business.



Machine Learning vs. Data Science

Home prices

size of house (square feet)	# of bedrooms	# of bathrooms	newly renovated	price (1000\$)
523 645 708 1034 2290	1 1 2 3 4	2 3 1 3	N N N Y N	115 150 210 280 355
2545	4	5	Y	440



Running AI system (e.g., websites / mobile app)

Homes with 3 bedrooms are more expensive than homes with 2 bedrooms of a similar size.

Newly renovated homes have a 15% premium.



Machine Learning vs.Data Science

Machine learning

"Field of study that gives computers the ability to learn without being explicitly programmed."

-Arthur Samuel (1959)

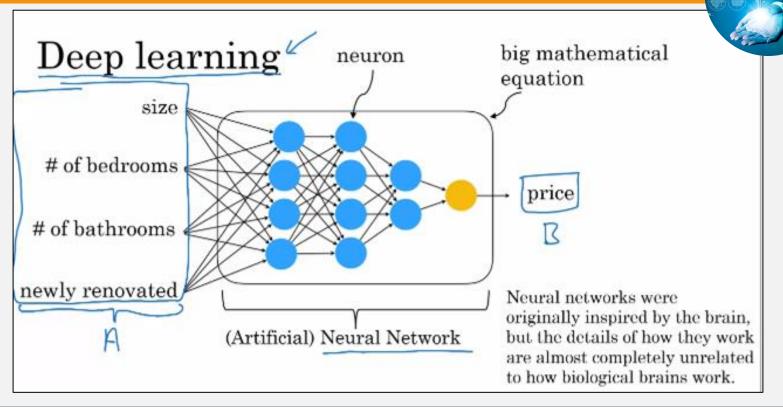
Data science

Science of extracting knowledge and insights from data.





Deep Learning



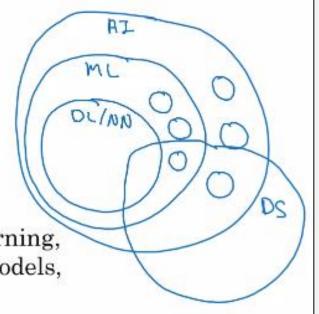


The terminology of AI

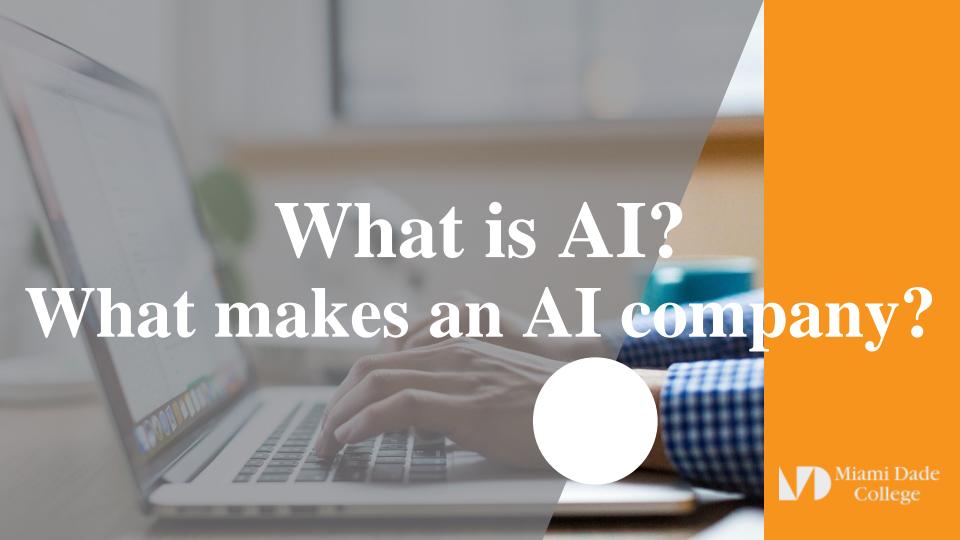
AI has many tools

- Machine learning and data science
- Deep learning / neural network

 Other buzzwords: Unsupervised learning, reinforcement learning, graphical models, planning, knowledge graph, ...







- What makes a company good at AI? Perhaps even more importantly, what will it take for your company to become great at using AI?
- So, what can you do for your company?
- This is the lesson I had learned by watching the rise of the Internet that I think will be relevant to how all of us navigate the rise of AI.



A lesson from the rise of the Internet

Internet Era

AI era



A lesson from the rise of the Internet

Internet Era

AI era

Shopping mall + website

≠ Internet company

Any company + deep learning ≠ AI company



A lesson from the rise of the Internet

Internet Era

Shopping mall + website
≠ Internet company

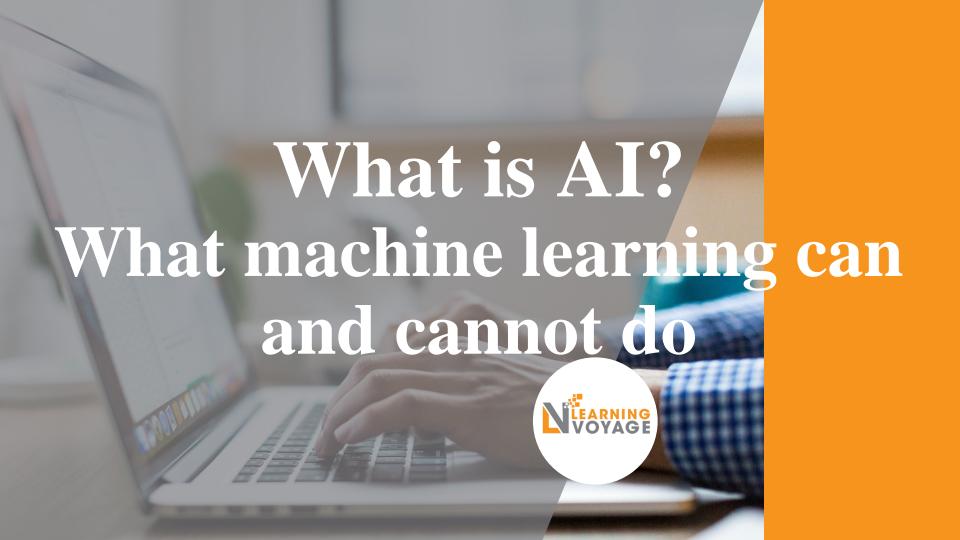
- A/B testing
- Short iteration time
- Decision making pushed down to engineers and other specialized roles

AI era

Any company + deep learning ≠ AI company

- Strategic data acquisition
- Unified data warehouse
- Pervasive automation
- New roles (e.g., MLE) and division of labor





What machine learning can and cannot do

- In these slides and the next slides, I hope to help you develop intuition about what AI can and cannot do. In practice, before I commit to a specific AI project, I'll usually have either myself or engineers do technical diligence on the project to make sure that it is feasible.
- This means: looking at the data, look at the input, and output A and B, and just thinking through if this is something AI can really do.



Supervised learning

Input (A)	Output (B)	Application		
email	spam? (0/1)	spam filtering		
audio	text transcripts	speech recognition		
English	Chinese	machine translation		
ad, user info	click? (0/1)	online advertising		
image, radar info	position of other cars	Self-driving car		
image of phone	defect? (0/1)	visual inspection		

Anything you can do with 1 second of thought, we can probably now or soon automate.



What machine learning can and cannot do

The toy arrived two days late, so I wasn't able to give it to my niece for her birthday.

Can I return it?



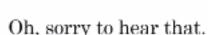
"Refund request"











I hope your niece had a good birthday.

Yes, we can help with....



What Happens If You Try?

Input (A) User email		Output (B) 2-3 paragraph response
1000 examples		
"My box was damaged."		Thank you for your email.
"Where do I write a review?"		Thank you for your email.
"What's the return policy?"		Thank you for your email.
"When is my box arriving?"		Thank yes now your



What makes an ML Problem Easier

1. Learning a "simple" concept

2. Lots of data available



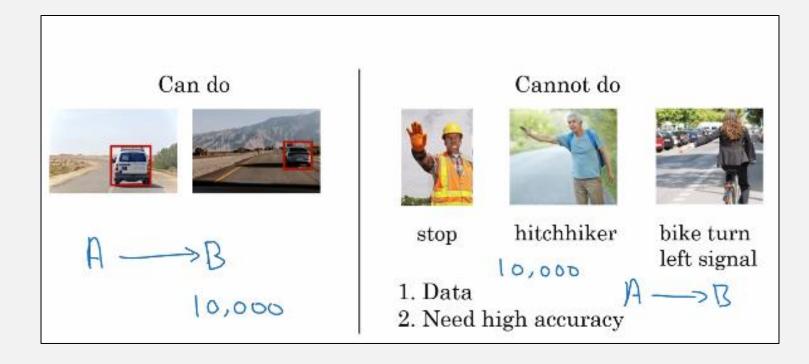


More examples of what machine learning can and cannot do

- One of the challenges of becoming good at recognizing what AI can and cannot do is that it does take seeing a few examples of concrete successes and failures of AI.
- If you work on an average of say, one new AI project a year, then to see three examples would take you three years of work experience and that's just a long time.



Self Driving Car





More examples of what machine learning can and cannot do

- Say you want to build an AI system to look at X-ray images and diagnose pneumonia. So, all of these are chest X-rays.
 - So, the input A could be the X-ray image and the output B can be the diagnosis.
- Does this patient have pneumonia or not?
 - So, that's something that AI can do.
- Something that AI cannot do would be to diagnose pneumonia from 10 images of a medical textbook chapter explaining pneumonia.



X Ray Diagnosis











Can do

Diagnose pneumonia from ~10,000 labeled images



Cannot do

Diagnose pneumonia from 10 images of a medical textbook chapter explaining pneumonia



Strengths and Weaknesses Of ML

ML tends to work well when:

- 1. Learning a "simple" concept
- 2. There is lots of data available

ML tends to work poorly when:

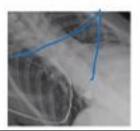
- 1. Learning complex concepts from small amounts of data
- 2. It is asked to perform on new types of data

A->B







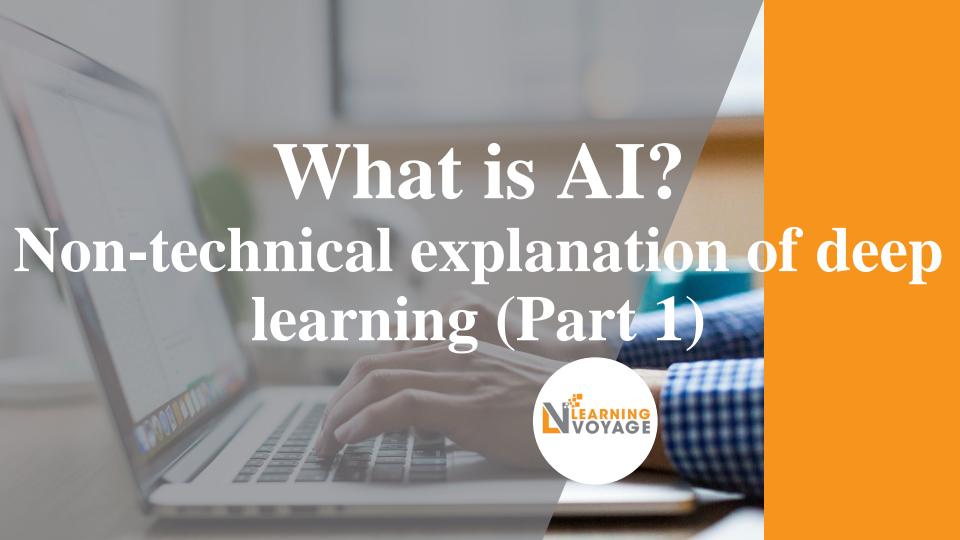




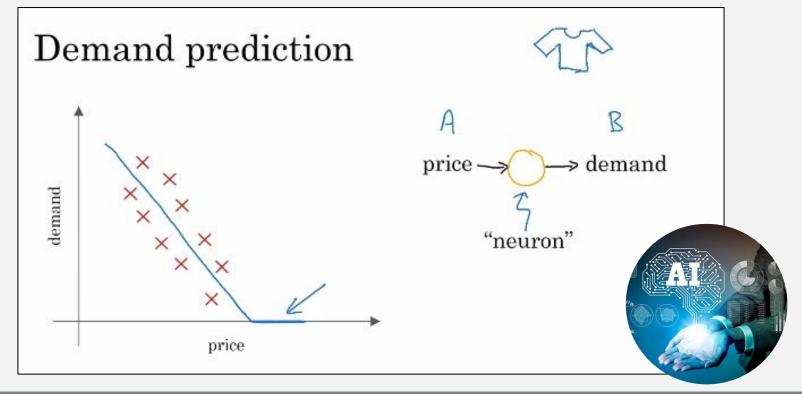
More examples of what machine learning can and cannot do

- I hope these examples are helping you hone your intuitions about what AI can and cannot do. In case the boundary between what it can or cannot do still seems fuzzy to you, don't worry.
- It is completely normal, completely okay. In fact even today, I still can't look at a project and immediately tell is something that's feasible or not.
- I often still need weeks or small numbers of weeks of technical diligence before forming strong conviction about whether something is feasible or not.





Non-technical explanation of deep learning (Part 1, optional)



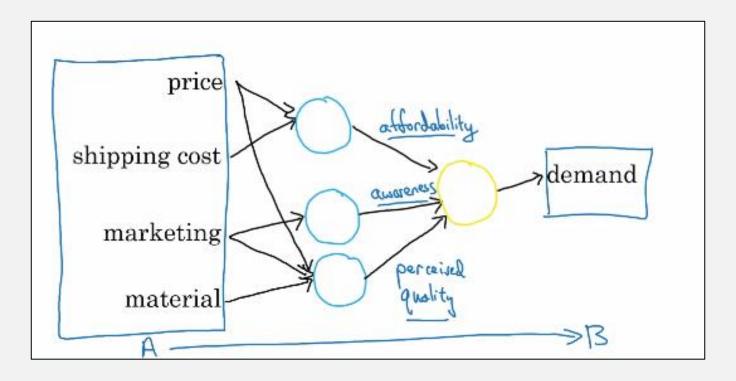


Demand Prediction

- Suppose that instead of knowing only the price of the product, you also have the shipping costs that the customers will have to pay to get the product.
- May be you spend more or less on marketing in a given week, and you can also make the product out of high quality material.



Demand Prediction



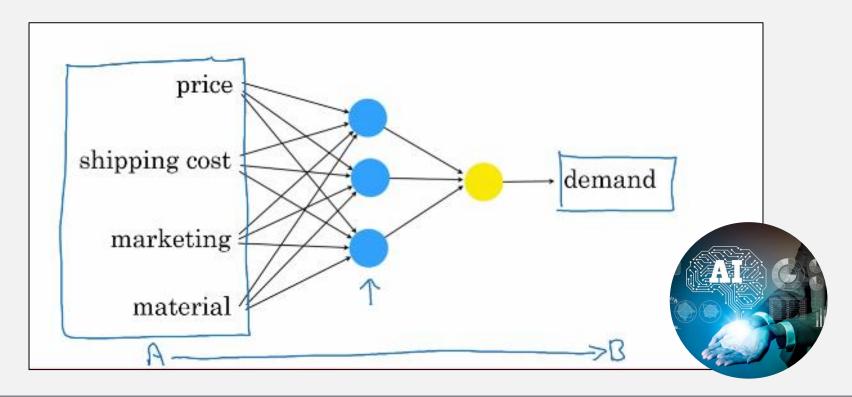


Non-technical explanation of deep learning (Part 1, optional)

- So It learns this input output or A to B mapping.
- This is a fairly small neural network with just four artificial neurons.
- In practice, neural networks used today are much larger, with easily thousands, tens of thousands or even much larger than that numbers of neurons.



Demand Prediction



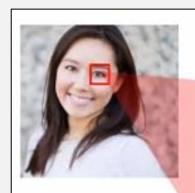


Non-technical explanation of deep learning (Part 1, optional)

- So that's a neural network, is a group of artificial neurons each of which computes a relatively simple function.
- But when you stack enough of them together like Lego bricks, they can compute incredibly complicated functions that give you very accurate mappings from the input A to the output B.
- Now, in this you saw an example of neural networks applied to demand prediction.



Face Recognition



30	32	22	12	10	10	12	33	35	30
12	11	12	234	170	176	13	15	12	12
234	222	220	230	200	222	230	234	56	78
190	220	186	112	110	110	112	180	30	32
49	250	250	250	4	2	254	200	44	.6
55	250	250	250	3	1	250	245	25	3
189	195	199	150	110	110	182	190	199	55
200	202	218	222	203	200	200	208	215	222
219	215	220	220	222	214	215	210	220	220
220	220	220	220	221	220	221	220	220	222



Non-technical explanation of deep learning (Part 2, optional)

