#### Introduction

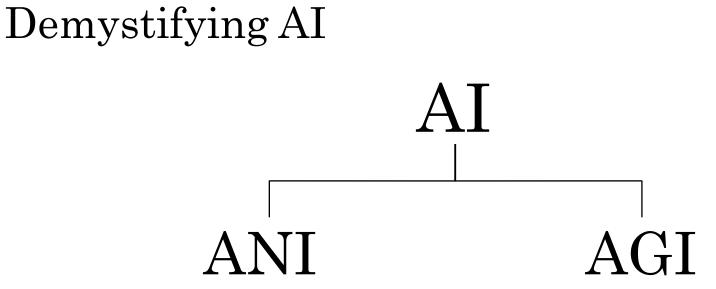
AI value creation by 2030 \$13 trillion

| Retail                              | \$0.8T |
|-------------------------------------|--------|
| Travel                              | \$480B |
| Transport & Logistics               | \$475B |
| Automotive & Assembly               | \$405B |
| Basic Materials                     | \$300B |
| Advanced Electronics/Semiconductors | \$291B |
| Healthcare Systems and Services     | \$267B |
| High Tech                           | \$267B |
| Telecom                             | \$174B |
| Oil & Gas                           | \$173B |
| Agriculture                         | \$164B |

[Source: McKinsey Global Institute.]







(artificial narrow intelligence)

E.g., smart speaker, self-driving car, web search, AI in farming and factories (artificial general intelligence)

Do anything a human can do



## What you'll learn

- What is AI?
  - Machine Learning
  - Data
  - What makes an AI company
  - What machine learning can and cannot do
  - Optional: Intuitive explanation of Deep Learning
- Building AI projects
- Building AI in your company
- AI and society



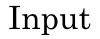


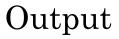
**Andrew Ng** 

## What is AI

Machine Learning

#### 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   |
|                   |                        |                     |

| Wh          | y Now?         |                   |
|-------------|----------------|-------------------|
|             |                | large neural net  |
|             |                | medium neural net |
| nance       |                | small neural net  |
| performance |                | traditional AI    |
|             |                |                   |
|             | amount of data | →                 |





Andrew Ng

## What is AI

What is data

### Example of a table of data (dataset)

| house<br>(square feet)                     | # of<br>bedrooms | price (1000\$)    |
|--|------------------|-------------------|
| 523  | 1                | 100               |
| $\begin{array}{c c} 645\\ 708 \end{array}$ | $\frac{1}{2}$    | $\frac{150}{200}$ |
| 1034                                       | $\frac{2}{3}$    | 300               |
| 2290                                       | 4                | 350               |
| 2545                                       | 4                | 440               |

| image | label   |
|-------|---------|
| C. C. | cat     |
| O O   | not cat |
|       | cat     |
|       | not cat |

"Google cat"



#### Acquiring data

- Manual labeling









- From observing user behaviors

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

| machine  | temperature<br>(°C) | pressure<br>(psi)        | machine<br>fault |
|--|---------------------|--------------------------|------------------|
| $\begin{array}{c} 17987 \\ 34672 \\ 08542 \end{array}$ | $60 \\ 100 \\ 140$  | $7.65 \\ 25.50 \\ 75.50$ | N<br>N<br>V      |
| 98536  | 140 $165$           | 125.00                   | Ŷ                |

- Download from websites / partnerships

#### Use and mis-use of data

Don't throw data at an AI team and assume it will be valuable.



Andrew Ng

#### Data is messy

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

images, audio, text

| house<br>(square feet) | # of<br>bedrooms | price<br>(1000\$)                           |
|------------------------|------------------|---|
| $523\\645$             | 1                | $\begin{array}{c} 100 \\ 0.001 \end{array}$ |
| 708                    | unknown          | 200   |
| 1034<br>unknown        | $\frac{3}{4}$    | unknown<br>350                              |
| 2545                   | unknown          | 440   |





**Andrew Ng** 

## What is AI

The terminology of AI

### Machine learning vs. data science

| Home   | size  | # of          | # of      | newly     | price                                     |
|--------|---|---------------|-----------|-----------|---|
| prices | (square feet)                               | bedrooms      | bathrooms | renovated | (1000\$)                                  |
| P      | $\begin{array}{c} 523 \\ 645 \end{array}$   | 1<br>1        | $2 \\ 3$  | N<br>N    | $\begin{array}{c} 100 \\ 150 \end{array}$ |
|        | 708   | 2             | 1         | N         | 200                                       |
|        | 1034  | 3             | 3         | Y         | 300                                       |
|        | $\begin{array}{c} 2290 \\ 2545 \end{array}$ | $\frac{4}{4}$ | 4 5       | N<br>Y    | $\frac{350}{440}$                         |

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

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

Newly renovated homes have a 15% premium.



**Andrew Ng** 

### Machine learning vs. data science

Machine learning

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

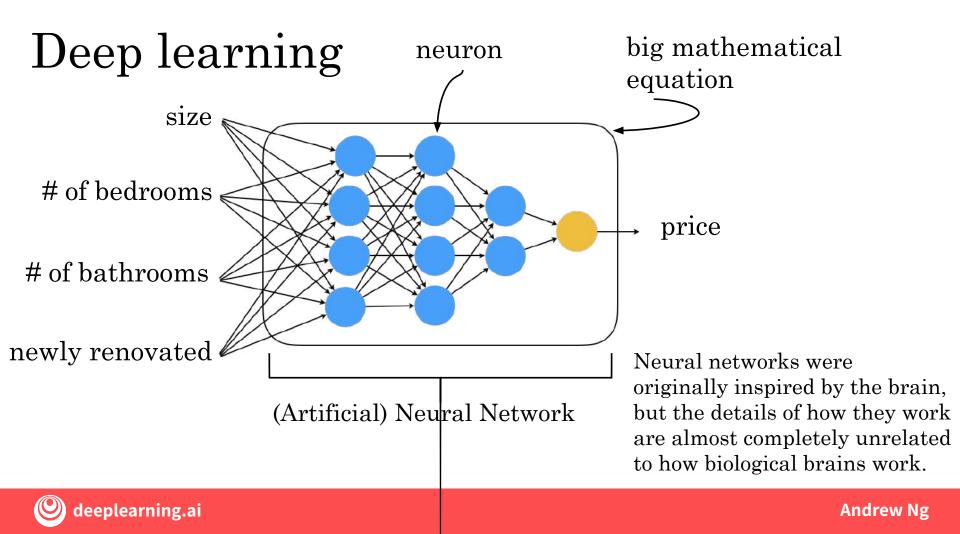
-Arthur Samuel (1959)

Data science

Science of extracting knowledge and insights from data.







### AI has many tools

- Machine learning and data science
- -Deep learning / neural network
- -Other buzzwords: Unsupervised learning, graphical models, planning, knowledge graph, ...





## What is AI

What makes an AI company?

**Andrew Ng** 

### A lesson from the rise of the Internet

#### <u>Internet Era</u>

#### Shopping mall + website ≠ Internet company

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

#### <u>AI era</u>

Any company + deep learning ≠ AI company

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

#### **AI** Transformation

- 1. Execute pilot projects to gain momentum
- 2. Build an in-house AI team
- 3. Provide broad AI training
- 4. Develop an AI strategy
- 5. Develop internal and external communication





**Andrew Ng** 

## What is AI

What machine learning can and cannot do

### Supervised Learning

| Input (A)         | Output (B)             | Application         |
|-------------------|------------------------|---------------------|
| email             | spam? (0/1)            | spam filtering      |
| audio             | text transcripts       | speech recognition  |
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| 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 today 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"

Input text Refund/ Support/ Shipping

Oh, sorry to hear that. I hope your niece had a good birthday. Yes, we can help with....



### What happens if you try?

<u>Input (A)</u> User email <u>Output (B)</u> 2-3 paragraph response

1000 examples

"My box was damaged."

"Where do I write a review?"

"What's the return policy?"

"When is my box arriving?"

Thank you for your email.

Thank you for your email.

Thank you for your email.

Thank yes now your....



Andrew Ng

#### What makes an ML problem easier

1. Learning a "simple" concept

2. Lots of data available







Andrew Ng

## What is AI

More examples of what machine learning can and cannot do

#### Self-driving car

#### Can do



#### Cannot do







#### $\operatorname{stop}$

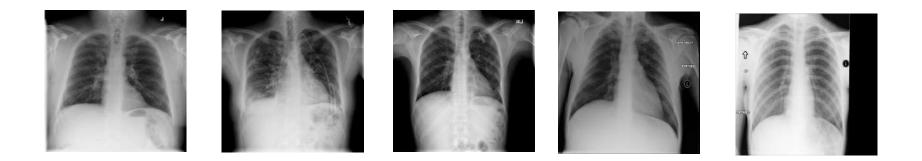
hitchhiker

bike turn left signal

Data
 Need high accuracy



#### 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 machine learning

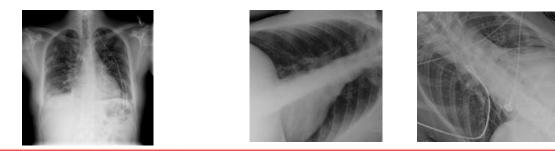
ML tends to work well when:

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

ML tends to work poorly when:

1. Learning complex functions from small amounts of data

2. It is asked to perform on new types of data that it learned from





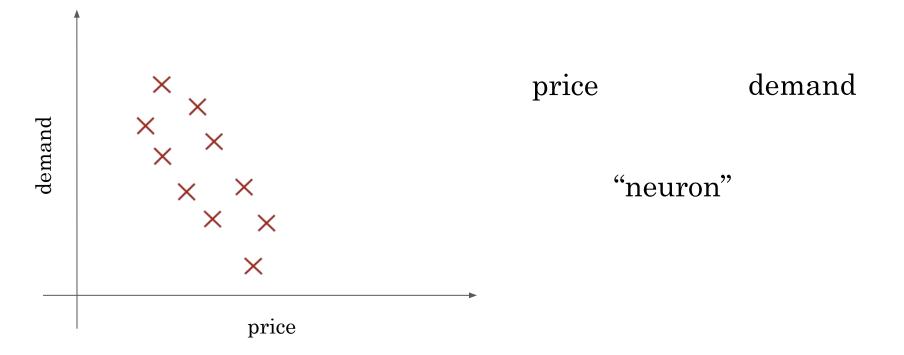


Andrew Ng

## What is AI

Non-technical explanation of deep learning I (optional)

### Demand prediction





Andrew Ng

### Demand prediction

price

shipping cost

marketing

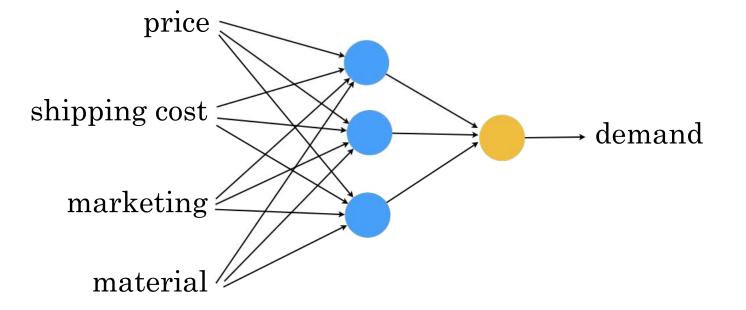
#### material





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### Demand prediction







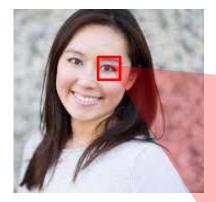


Andrew Ng

## What is AI

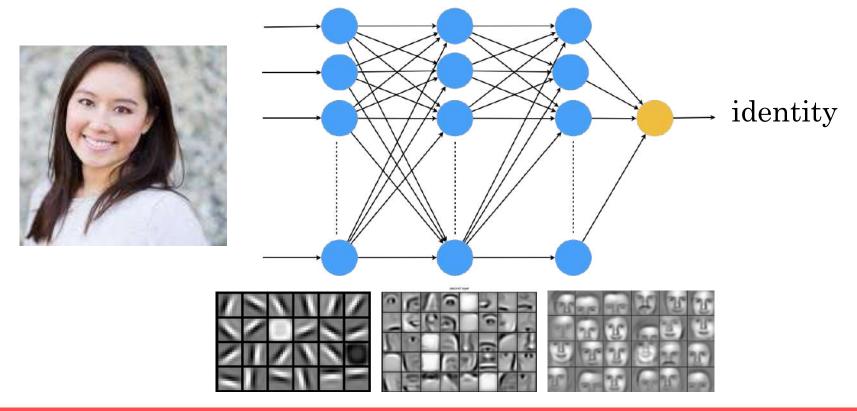
Non-technical explanation of deep learning II (optional)

#### 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 |

#### Face recognition





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Building AI Projects

Starting an AI project

# Starting an AI project

- Workflow of projects
- Selecting AI projects
- Organizing data and team for the projects





Building AI Projects

Workflow of a machine learning project

### Example: Speech recognition



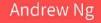






Amazon Echo / Alexa Google *Home*  Apple Siri Baidu *DuerOS* 





### Key steps of a machine learning project

Echo / Alexa

- 1. Collect data
- 2. Train model Iterate many times until good enough
  - 3. Deploy model Get data back Maintain / update model



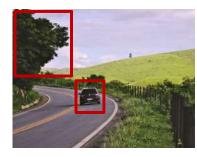
### Key steps of a machine learning project

### Self-driving car

1. Collect data



- 2. Train model Iterate many times until good enough
- 3. Deploy model Get data back Maintain / update model







#### image

#### position of other cars





# Building AI Projects

Workflow of a data science project

### Example: Optimizing a sales funnel

#### Visit website



#### Product page



#### Shopping cart



#### Checkout

| Mugsy   | 1 (E. 1) (B. 1)                                |
|---------|--|
|         | or your purchase!<br>r the order confirmation. |
| Continu | e to homepage                                  |





### Key steps of a data science project

#### Optimizing a sales funnel

- 1. Collect data
- 2. Analyze data

| User ID | Country     | Time            | Webpage     |
|---------|-------------|-----------------|-------------|
| 2009    | Spain       | 08:34:30 Jan 5  | home.html   |
| 2897    | USA         | 13:20:22 May 18 | redmug.html |
| 4893    | Philippines | 22:45:16 Jun 11 | mug.html    |

Iterate many times to get good insights

3. Suggest hypotheses/actions Deploy changes Re-analyze new data periodically



### Key steps of a data science project

| Manufacturing line Final<br>Mix clay Shape mug Add glaze Fire kiln inspection |                   |            | Clay<br>Batch<br># | Supplier                   | Mixing<br>time<br>(minutes)    |
|---|-------------------|------------|--------------------|----------------------------|--------------------------------|
| Mix clay Shape hiug Add glaze Fil   |                   | Inspection | 001                | ClayCo                     | 35                             |
|   |                   |            | 034                | GooClay                    | 22                             |
|   |                   |            | 109                | BrownStuff                 | 28                             |
| 1. Collect data   | Mug<br>Batch<br># | Country    | Humidity           | Temperature<br>in kiln (F) | Duration<br>in kiln<br>(hours) |
| 2. Analyze data   | 301               | Spain      | 0.002%             | 1410°                      | 22                             |
| Iterate many times to get good insight.<br>3. Suggest hypotheses/actions      |                   | USA        | 0.003%             | 1520°                      | 24                             |
| Deploy changes<br>Re-analyze new data periodically                            | 303               | Malaysia   | 0.002%             | 1420°                      | 22                             |
|   |                   |            |                    |                            |                                |



Andrew Ng

# Building AI Projects

#### Every job function needs to learn how to use data



#### Data science



#### Optimize sales funnel

# Red Ceramic Mug

## . Check your email for the order confirmation.

#### Machine learning

| Name   | Title   | Company<br>size | Email  | Priority |
|--------|---------|-----------------|--------|----------|
| Tayler | CEO     | 3050            | tay@a  | high     |
| Janet  | Manager | 230             | jan@b  | medium   |
| David  | Intern  | 30              | dave@c | low      |

#### Automated lead sorting



### Manufacturing line manager

#### Data science

Mix clay







Final



#### Optimize sales funnel





inspection





Machine learning



ok

ok

defect

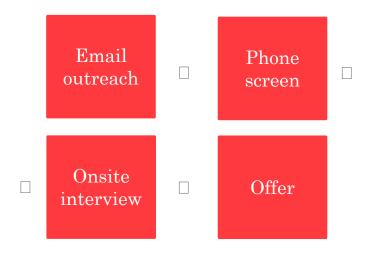
#### Automated visual inspection



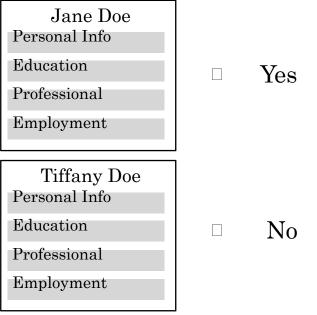
Fire kiln

### Recruiting

#### Data science



### Machine learning



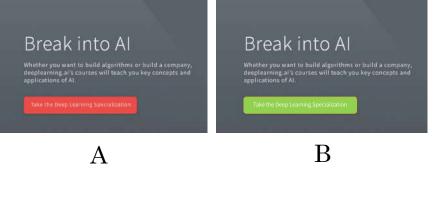
#### Optimize recruiting funnel



Automated resume screening

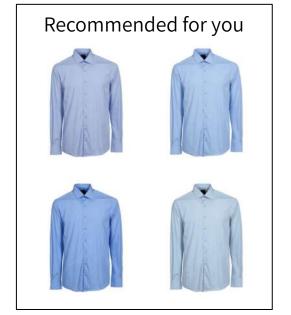
### Marketing

#### Data science



#### A/B testing

#### Machine learning



Customized product recommendation



### Agriculture

#### Data science



Crop analytics

#### Machine learning



#### Precision weed killing







# Building AI Projects

How to choose an AI project I

### AI knowledge and domain knowledge

What AI can do Things valuable for your business





### Brainstorming framework

- Think about optimizing tasks rather than automating jobs. E.g., call center routing, radiologists.
- What are the main drivers of business value?
- What are the main points in your business?



### You can make progress even without big data

- Having more data almost never hurts.
- Data makes some businesses (like web search) defensible.
- But with small datasets, you might still make progress.



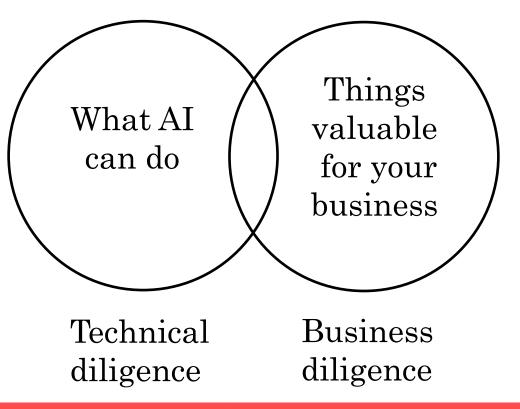




# Building AI Projects

How to choose an AI project II

Due diligence on project





### Due diligence on project

#### Technical diligence

- Can AI system meet desired performance
- How much data is needed
- Engineering timeline

#### Business diligence

- Lower costs
- Increase revenue

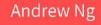
- current business
- Launch new product or new business business



### Build vs. buy

- ML projects can be in-house or outsourced
- DS projects are more commonly in-house
- Some things will be industry standard avoid building those.







Building AI Projects

Working with an AI team

### Specify your acceptance criteria



Goal: detect defects with 95% accuracy

Provide AI team a dataset on which to measure their performance



### How AI teams think about data

#### Training set

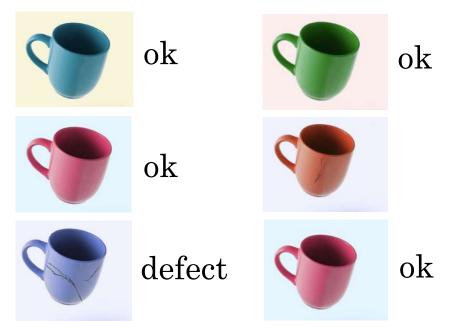


Test set ok ok defect



### Pitfall: Expecting 100% accuracy

#### Test set



- Limitations of ML
- Insufficient data
- Mislabeled data
- Ambiguous label







Andrew Ng

# Building AI Projects

Technical tools for AI teams (optional)

### **Open-source** frameworks

Machine learning frameworks:

- TensorFlow
- PyTorch
- Keras
- MXNet
- CNTK
- Caffe
- PaddlePaddle
- Scikit-learn
- R
- Weka

Research publications

Arxiv

### Open source repositories:

• GitHub



### CPU vs. GPU

CPU: Computer processor (Central Processing Unit)



#### **GPU:** Graphics Processing Unit



#### Cloud vs. On-premises





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Building AI in Your Company

Introduction

### Building AI in your company

- Case studies of complex AI products
- Roles in an AI team
- AI Transformation Playbook
- Taking your first step





# Building AI in Your Company

Case study: Smart speaker

#### Smart speaker









Amazon Echo / Alexa

| Google |
|--------|
| Home   |

Apple Siri

Baidu *DuerOS* 

#### "Hey device, tell me a joke"



## "Hey device, tell me a joke"

Steps to process the command:

- 1. Trigger word/wakeword detection Audio "Hey device"? (0/1)
- 2. Speech recognition Audio
- 3. Intent recognition
- 4. Execute joke

learning.ai



Intent

recognition

joke? time? music? call?



Execution

Trigger word detection Speech recognition

🕖 dee

#### "Hey device, set timer for 10 minutes"

Steps to process the command:

- 1. Trigger word/wakeword detection Audio "Hey device"? (0/1)
- 2. Speech recognition Audio "set timer for 10 minutes"
- **3**. Intent recognition
- 4. a) Extract duration

"Set timer for 10 minutes"

"Let me know when 10 minutes is up"

b) Start timer with set duration

"set timer for 10 minutes" timer

## Other functions

- Play music
- Volume up/down
- Make call
- Current time
- Units conversion
- Simple question

#### Key steps:

- 1. Trigger/wakeword detection
- 2. Speech recognition
- 3. Intent recognition
- 4. Specialized program to execute command

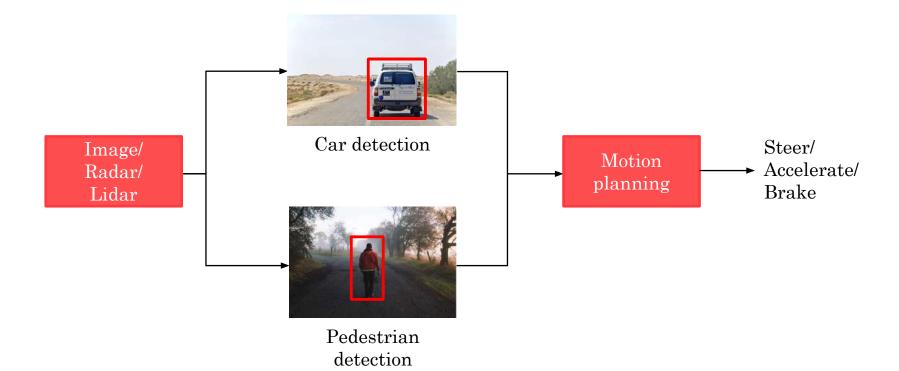




Building AI in Your Company

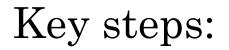
Case study: Self-driving car

#### Steps for deciding how to drive









1. Car detection





2. Pedestrian detection





3. Motion planning

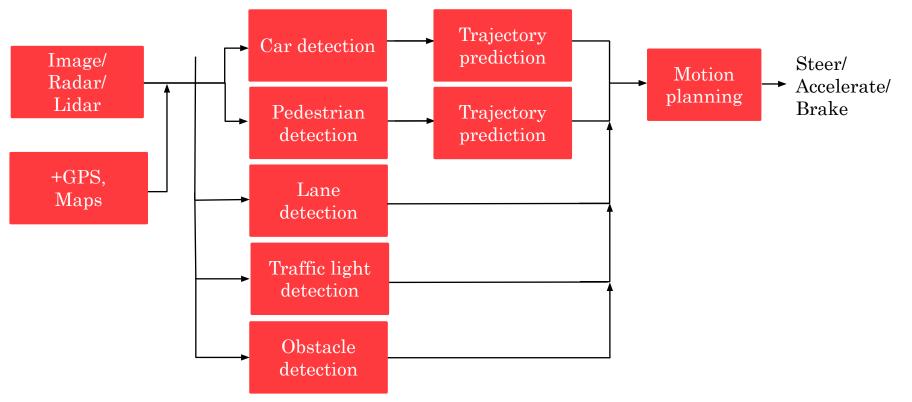








### Steps for deciding how to drive









# Building AI in Your Company

Example roles of an AI team

#### Example roles

- Software Engineer
  - E.g., joke execution, ensure self-driving reliability, ...
- Machine Learning Engineer

#### Applied ML Scientist

- Machine Learning Researcher
  - Extend state-of-the-art in ML





## Example roles

- Data Scientist
  - Examine data and provide insights
  - Make presentation to team/executive
- Data Engineer
  - Organize data

```
1 MB (megabyte)
1,000 MB = GB (gigabyte)
1,000,000 MB = TB (terabyte)
1,000,000,000 MB = PB (petabyte)
```

- Make sure data is saved in an easily accessible, secure and cost effective way
- AI Product Manager
  - Help decide what to build; what's feasible and valuable

#### Getting started with a small team

- 1 Software Engineer, or
- 1 Machine Learning Engineer/Data Scientist, or
- Nobody but yourself





# Building AI in Your Company

AI Transformation Playbook (Part I)

## AI Transformation Playbook

- 1. Execute pilot projects to gain momentum
- 2. Build an in-house AI team
- 3. Provide broad AI training
- 4. Develop an AI strategy
- 5. Develop internal and external communications



#### 1. Execute pilot projects to gain momentum

- More important for the initial project to succeed rather than be the most valuable
- Show traction within 6-12 months
- Can be in-house or outsourced



#### 2. Build an in-house AI team

#### **BU=** Business Unit

AI function can be under CTO, CIO, CDO, etc. or a new CAIO



# 3. Provide broad AI training

| Role   | What they should learn  |  |  |
|--|---|--|--|
| Executives and senior business<br>leaders      | <ul><li>What AI can do for your enterprise</li><li>AI strategy</li><li>Resource allocation</li></ul>                                    |  |  |
| Leaders of divisions working on AI<br>projects | <ul> <li>Set project direction (technical and<br/>business diligence)</li> <li>Resource allocation</li> <li>Monitor progress</li> </ul> |  |  |
| AI engineer trainees                           | <ul> <li>Build and ship AI software</li> <li>Gather data</li> <li>Execute on specific AI projects</li> </ul>                            |  |  |

The smart CLO knows they should *curate* rather than *create* content





# Building AI in Your Company

AI Transformation Playbook (Part II)

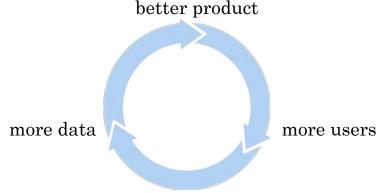
## AI Transformation Playbook

- 1. Execute pilot projects to gain momentum
- 2. Build an in-house AI team
- 3. Provide broad AI training
- 4. Develop an AI strategy
- 5. Develop internal and external communications



### 4. Develop an AI strategy

- Leverage AI to create an advantage specific to your industry sector
- Design strategy aligned with the "Virtuous Cycle of AI"





## 4. Develop an AI strategy

- Consider creating a data strategy

   Strategic data acquisition
   Unified data warehouse
- Create network effects and platform advantages

   In industries with "winner take all" dynamics,

   AI can be an accelerator



#### 5. Develop internal and external communications

- Investor relations
- Government relations
- Customer/user education
- Talent/recruitment
- Internal communications

Detailed AI Transformation Playbook: https://landing.ai/ai-transformation-playbook/





Building AI in Your Company

AI pitfalls to avoid

# AI pitfalls to avoid

#### Don't:

• Expect AI to solve everything

• Hire 2-3 ML engineers and count solely on them to come up with use cases

#### Do:

- Be realistic about what AI can and cannot do given limitations of technology, data, and engineering resources
- Pair engineering talent with business talent and work crossfunctionally to find feasible and valuable projects



# AI pitfalls to avoid

#### Don't:

- Expect the AI project to work the first time
- Expect traditional planning processes to apply without changes
- Think you need superstar AI engineers before you can do anything

#### Do:

- Plan for AI development to be an iterative process, with multiple attempts needed to succeed
- Work with AI team to establish timeline estimates, milestones, KPIs, etc.
- Keep building the team, but get going with the team you have





Building AI in Your Company

Taking your first step in AI

## Some initial steps you can take

- Get friends to learn about AI -This course
  - -Reading group
- Start brainstorming projects -No project is too small
- Hire a few ML/DS people to help
- Hire or appoint an AI leader (VP AI, CAIO, etc.)
- Discuss with CEO/Board possibilities of AI Transformation -Will your company be much more valuable and/or more effective if it were good at AI?



Andrew Ng

# Building AI in Your Company

Survey of major AI application areas (optional)

# **Computer Vision**

• Image classification/Object recognition -Face recognition

register

- Object detection
- Image segmentation
- Tracking









cat



new







## Natural Language Processing

- Text classification
  - Sentiment recognition
- Information retrieval - E.g., web search
- Name entity recognition
- Machine translation

Email

Product description

"The food was good"

"Service was horrible"

Spam/Non-Spam

Product category



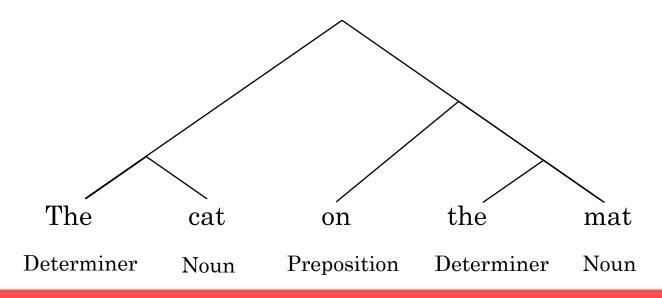
<sup>(Queen Elizabeth II]</sup> knighted Sir Paul McCartney for his services to music at the Buckingham Palace

AIは、新たな電気だ AI is the new electricity



#### Natural Language Processing

• Others: parsing, part-of-speech tagging





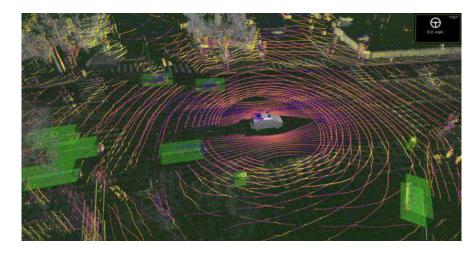


- Speech recognition (speech-to-text)
- Trigger word/wakeword detection
- Speaker ID
- Speech synthesis (text-to-speech, TTS) The quick brown fox jumps over the lazy dog.



#### Robotics

- Perception: figuring out what's in the world around you
- Motion planning: finding a path for the robot to follow
- Control: sending commands to the motors to follow a path





#### General machine learning

• Unstructured data (images, audio, text)



#### • Structured data

| House size    | # of                                 | Price    | Clay batch | Supplier   | Mixing time |
|---------------|--------------------------------------|----------|------------|------------|-------------|
| (square feet) | bedrooms                             | (1000\$) | #          |            | (minutes)   |
| 523           | $\begin{array}{c}1\\1\\2\end{array}$ | 100      | 001        | ClayCo     | 35          |
| 645           |                                      | 150      | 034        | GooClay    | 22          |
| 708           |                                      | 200      | 109        | BrownStuff | 28          |





# Building AI in Your Company

Survey of major AI techniques (optional)

Unsupervised learning

Clustering Potato chip sales price per packet  $\times \times \times \times$ # of packets

Given data (without any specific desired output labels), find something interesting about the data



Finding cats from unlabeled YouTube videos



## Transfer learning

### Car detection



Golf cart detection



100,000 images

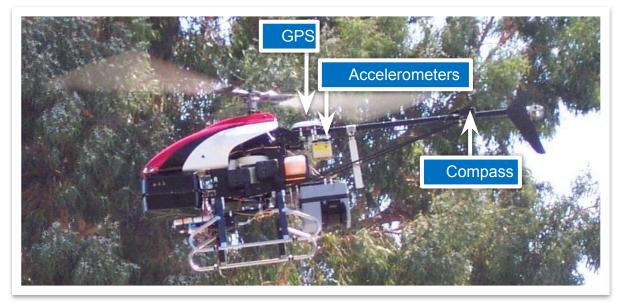
100 images

### Learn from task A, and use knowledge to help on task B





## Reinforcement learning



Use a "reward signal" to tell the AI when it is doing well or poorly. It automatically learns to maximize its rewards.



## **Reinforcement learning**



Use a "reward signal" to tell the AI when it is doing well or poorly. It automatically learns to maximize its rewards.



### GANs (Generative Adversarial Network)

Synthesize new images from scratch



[Source: Karras et al. (2018). Progressive Growing of GANs for Improved Quality, Stability, and Variation]





### Knowledge Graph

Jan 3, 2019 - Ada Lovelace, in full Ada King, countess of Lovelace, original name Augusta Ada Byron,

| All Images Books News Videos More Settings Tool  |  |       |
|--|--|-------|
|  | ols  |       |
| ada lovelace   | \$ a   |       |
| All Images Books Videos News More Settings   | s Tools  |       |
| About 9,260,000 results (0.35 seconds)   |  |       |
| Ada Lovelace - Wikipedia<br>https://en.wikipedia.org/wiki/Ada Lovelace -   |  | 1     |
| Augusta Ada King, Countess of Lovelace * an English mathematician and writer, chiefly k<br>her work on Charles Babbage's proposed mechanical | rknown for   |       |
| Resting place: Church of St. Mary Magdalene, Spouse(s): William King-Noel, 1st Earl o  | of A Contraction of A Contraction  |       |
| Known for: Mathematics, computing  |  |       |
| Charles Babbage - Analytical Engine - William King-Noel, 1st Earl of - Lady Byron  |  |       |
| Ada Lovelace: Founder of Scientific Computing  |  | e in  |
| https://www.sdsc.edu/ScienceWomen/lovelace.html *  | Ada Lovelace   |       |
| ADA BYRON, COUNTESS OF LOVELACE Ada Byron was the daughter of a brief married  |  |       |
| between the Romantic poet Lord Byron and Anne Isabelle   |  |       |
| People also ask  | Augusta Ada King, Countess of Lovelace was an English mait<br>and writer, chiefly known for her work on Charles Babbage's p<br>mechanical general-purpose computer, the Analytical Engine. | ropos |
| What is Ada Lovelace famous for?   | Sorn: December 10, 1815, London, United Kingdom  |       |
|  | Died: November 27, 1852, Marylebone, United Kingdom     Spouse: William King-Noel, 1st Earl of Lovelace (m. 1835-18  |       |
| What did Ada Lovelace invent and what impact it had?   |  | (27   |
| What did Ada Lovelace invent and what impact it had?<br>When did Ada Lovelace invent the computer?   | Children: Anne Blunt, 15th Baroness Wentworth, MORE  | 52)   |
|  |  | 52)   |

10

| Ada Lovelace |  |  |
|--------------|--|--|
| Born         | Dec 10, 1815                           |  |
| Died         | Nov 27, 1852                           |  |
| Bio          | English<br>mathematician and<br>writer |  |

| Northern Rooster Hotel |                   |
|------------------------|-------------------|
| Address                | 45 Rooster St, LA |
| Phone                  | (650) 555-3992    |
| Wifi                   | yes               |
| Pool                   | no                |



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AI and Society

Introduction

## AI and society

- AI and hype
- Limitations of AI

-Bias

-Adversarial attacks

- AI, developing economies, and jobs
- Conclusion



AI and Ethics



Andrew Ng

# AI and Society

A realistic view of AI

### Goldilocks rule for AI

- Too optimistic: Sentient / super-intelligent AI killer robots coming soon
- Too pessimistic: AI cannot do everything, so an AI winter is coming
- Just right: AI can't do everything, but will transform industries

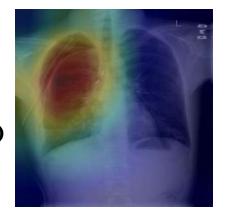


### Limitations of AI

- Performance limitations
- Explainability is hard (but sometimes doable)



Right-sided Pneumothorax (collapsed lung)



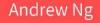
[Rajpurkar et al. (2018). CheXNet: Radiologist-Level Pneumonia Detection on Chest X-Rays with Deep Learning.] [Wang et al. (2017). ChestX-ray8: Hospital-scale Chest X-ray Database and Benchmarks on Weakly-Supervised Classification and Localization of Common Thorax Diseases. IEEE CVPR] [Images source: NIH Clinical Center Image dataset: <u>https://nihcc.app.box.com/v/ChestXray-NIHCC</u>]



### Limitations of AI

- Biased AI through biased data
- Adversarial attacks on AI





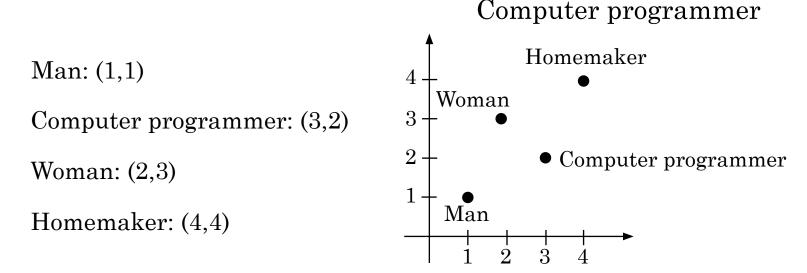


AI and Society

**Discrimination / Bias** 

### AI learning unhealthy stereotypes

- Man : Woman as Father : Mother
- Man : Woman as King : Queen
- Man : Computer programmer as Woman : Homemaker



Bolukbasi et al. (2016). Man is to Computer Programmer as Woman is to Homemaker? Debiasing Word Embeddings.



### Why bias matters

- Hiring tool that discriminated against women
- Facial recognition matching dark skinned individuals to criminal mugshots
- Bank loan approvals
- Toxic effect of reinforcing unhealthy stereotypes



### Combating bias

- Technical solutions:
  - E.g., "zero out" the bias in words
  - Use less biased and/or more inclusive data
- Transparency and/or auditing processes
- Diverse workforce
  - Creates less biased applications







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# AI and Society

Adversarial attacks on AI

### Adversarial attacks on AI



Hummingbird

### Minor perturbation



Hammer



Hare

Minor perturbation



Desk



### Physical attacks



"Milla Jovovich"



#### Fails to see stop sign

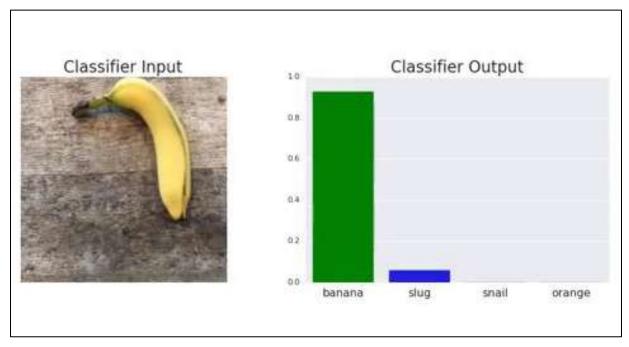


Banana





### Physical attacks



[Sharif et al. (2016). Accessorize to a Crime: Real and Stealthy Attacks on State-of-the-Art Face Recognition.] [Eykholt et al. (2018). Physical Adversarial Examples for Object Detectors.] [Brown et al. (2018). Adversarial Patch.]



### Adversarial defenses

- Defenses do exist, but incur some cost
- Similar to spam vs. anti-spam, we may be in an arms race for some applications







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# AI and Society

Adverse uses of AI

### Adverse uses of AI

- DeepFakes
  - Synthesize video of people doing things they never did
- Undermining of democracy and privacy - Oppressive monitoring of individuals
- Generating fake comments
- Spam vs. anti-spam and fraud vs. anti-fraud





## AI and Society

AI and developing economies

### Developing economies

### "leapfrog"

-Mobile phones -Mobile payments -Online education





### How developing economies can build AI

- US and China are leading, but all AI communities are still immature
- Focus on AI to strengthen a country's vertical industries
- Public-private partnerships to accelerate development





# AI and Society

AI and jobs

### AI's impact on jobs worldwide

### Jobs replaced by 2030

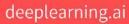
### Jobs created by 2030

## $400\text{--}800\,\mathrm{mil}$

## $555-890 \mathrm{\ mil}$

[Source: McKinsey Global Institute.]

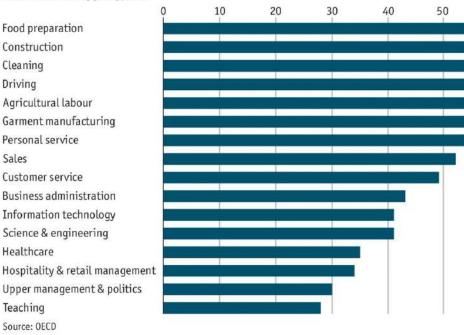




### AI's impact on jobs worldwide

#### Automated for the people

Automation risk by job type, %



[Image credit: Economist.com] [Nedelkoska, L. and G. Quintini. (2018). Automation, skills use and training. OECD Social, **Employment** and Migration Working Papers, No. 202.]

60

Economist.com

Driving

Sales



### Some solutions

- Conditional basic income: provide a safety net but incentivize learning
- Lifelong learning
- Political solutions





AI for Everyone

Conclusion

### What you've learned

- What is AI?
- Building AI projects
- Building AI in your company
- AI and society

### Keep learning!

- Online courses, books, blogs, ...
- deeplearning.ai mailing list







