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



- Workflow of projects
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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







image position of other cars

- 2. Train model

  Iterate many times until good enough
- 3. Deploy model

  Get data back

  Maintain / update model









Workflow of a data science project

## Example: Optimizing a sales funnel



## Key steps of a data science project

#### Optimizing a sales funnel

1. Collect 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

2. Analyze data

Iterate many times to get good insights

3. Suggest hypotheses/actions

Deploy changes

Re-analyze new data periodically

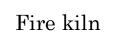
## Key steps of a data science project

#### Manufacturing line

Mix clay S



Add glaze



Final inspection



Clay Batch #	Supplier	Mixing time (minutes)
001	ClayCo	35
034	GooClay	22
109	BrownStuff	28









- 1. Collect data
- 2. Analyze data

  Iterate many times to get good insight
- 3. Suggest hypotheses/actions
  Deploy changes

Re-analyze new data periodically

	Mug Batch #	Country	Humidity	Temperature in kiln (F)	Duration in kiln (hours)
t	301	Spain	0.002%	1410°	22
U	302	USA	0.003%	1520°	24
	303	Malaysia	0.002%	1420°	22



Every job function needs to learn how to use data

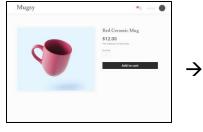
### Sales

#### Data science

#### Visit website



Product page



Shoppingcart



Checkout



Optimize sales funnel

#### Machine learning

Name	Title	Company 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 Shape mug Add glaze

Add glaze

Final Fire kiln inspection



Optimize sales funnel

#### Machine learning



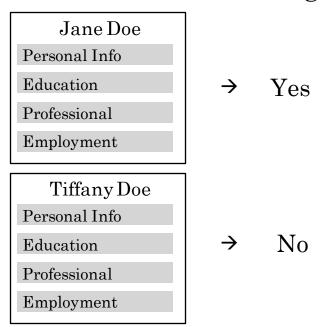
Automated visual inspection

## Recruiting

Data science Email Phone  $\rightarrow$  $\rightarrow$ outreach screen Onsite  $\rightarrow$ Offer interview

Optimize recruiting funnel

#### Machine learning



Automated resume screening

### Marketing

Data science

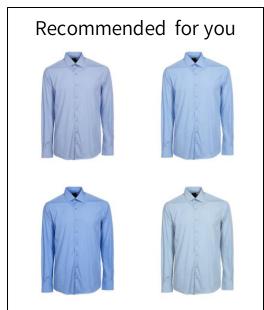




A B

A/B testing

#### Machine learning



Customized product recommendation

## Agriculture

Data science



Crop analytics

Machine learning



Precision weed killing

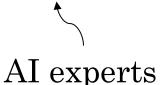


How to choose an AI project I

## AI knowledge and domain knowledge

What AI can do

Things valuable for your business



Domain experts

## 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

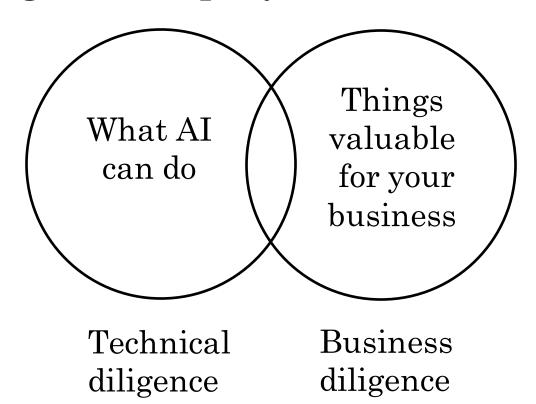
- Having more data almost never hurts.
- Data makes some businesses (like web search) defensible.
- But with small datasets, you might still make progress.





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
  - Launch new product or business

current business

new

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.



Working with an AI team

## Specify your acceptance criteria



ok



ok



defect

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



ok



ok

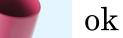


ok

defect







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





Technical tools for AI teams (optional)

## Open-source frameworks

#### Machine learning frameworks:

- PyTorch
- TensorFlow
- Hugging Face
- PaddlePaddle
- Scikit-learn
- R

#### Research publications

Arxiv

Open source repositories:

GitHub

#### CPU vs. GPU

CPU: Computer processor (Central Processing Unit)





GPU: Graphics Processing Unit



Cloud vs. On-premises