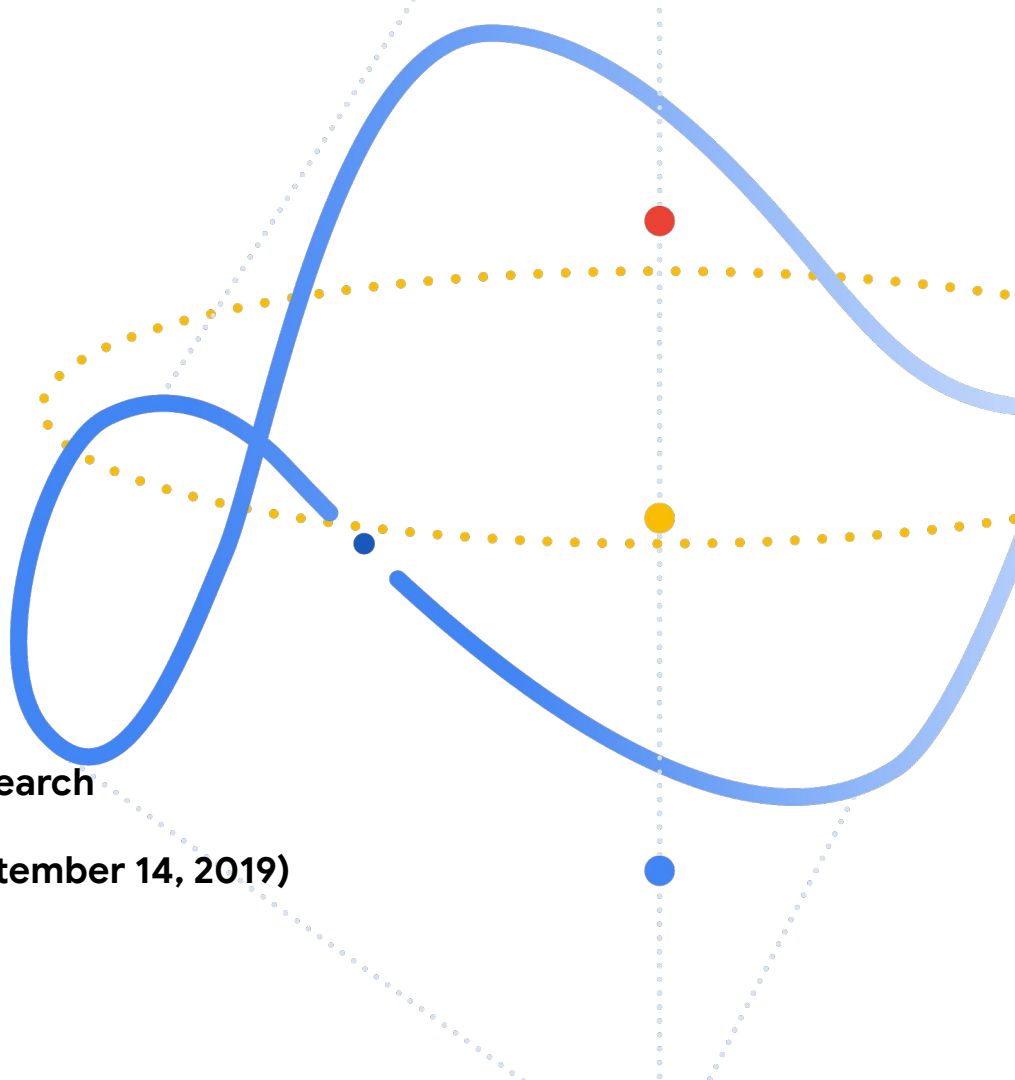


How to find data set and fairness practices

Sayak Paul

Deep Learning Associate at PylmageSearch

Explore ML Academy, Hyderabad (September 14, 2019)

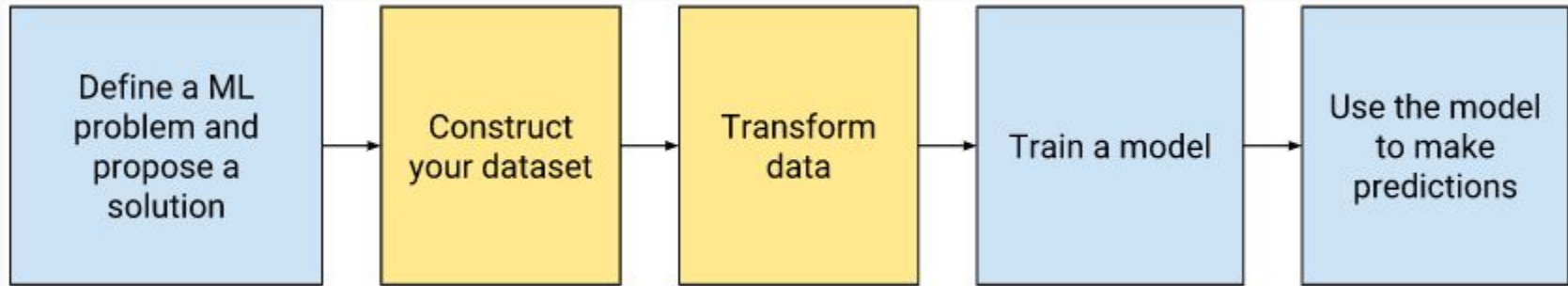




Agenda

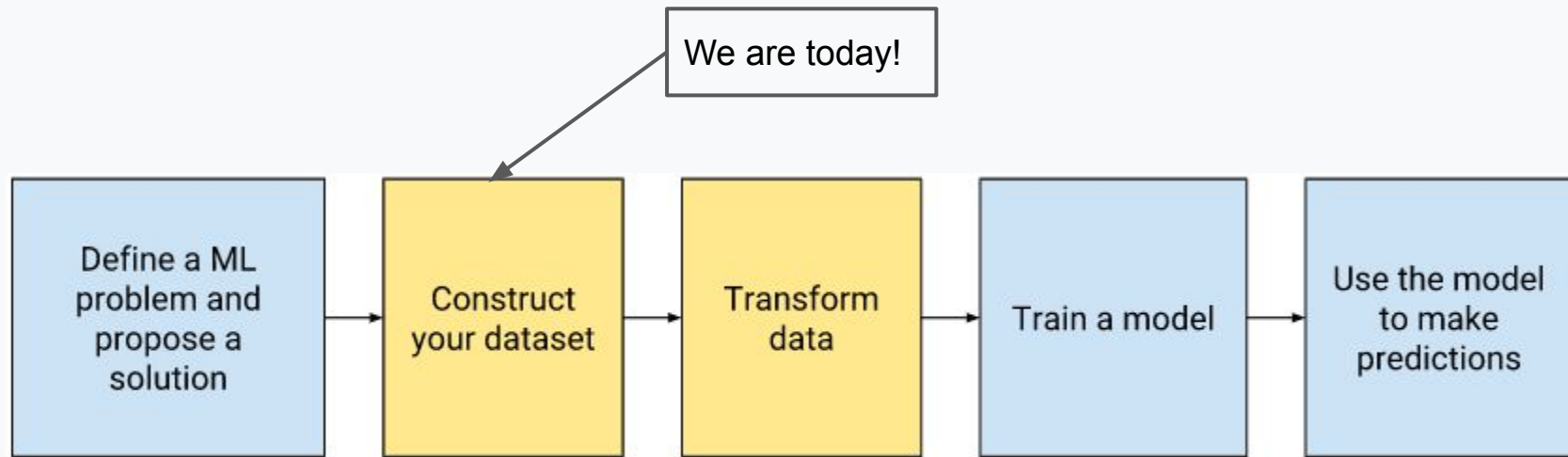
- Why is collecting a good dataset important?
- The process of data preparation
 - Collect raw data
 - Identify features and label sources
 - Select sampling strategy
 - Data splitting
- Incorporating fairness practices
- Guiding lights

The *typical* machine learning workflow

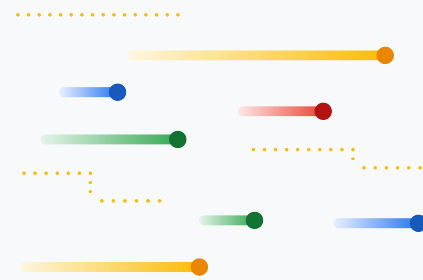


[Source](#)

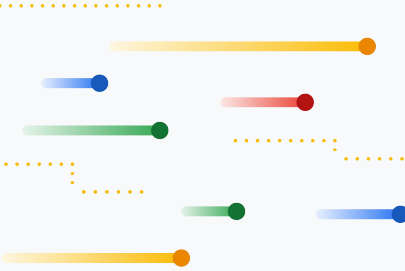
The *typical* machine learning workflow



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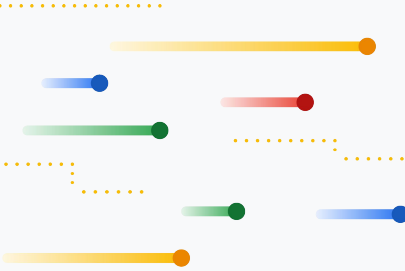


**Why is collecting a good
dataset important?**



"...one of our most impactful quality advances since neural machine translation has been in identifying the best subset of our training data to use"

- Software Engineer, Google Translate



"...one of our most impactful quality advances since neural machine translation has been in identifying the best subset of our training data to use"

- Software Engineer, Google Translate

"...most of the times when I tried to manually debug interesting-looking errors they could be traced back to issues with the training data." - Software Engineer, Google Translate



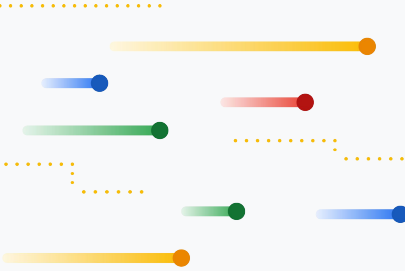
This is a sample from the training data

Lane detection in self-driving cars ([Source](#))



A sample during the time
of inference!

[Source](#)



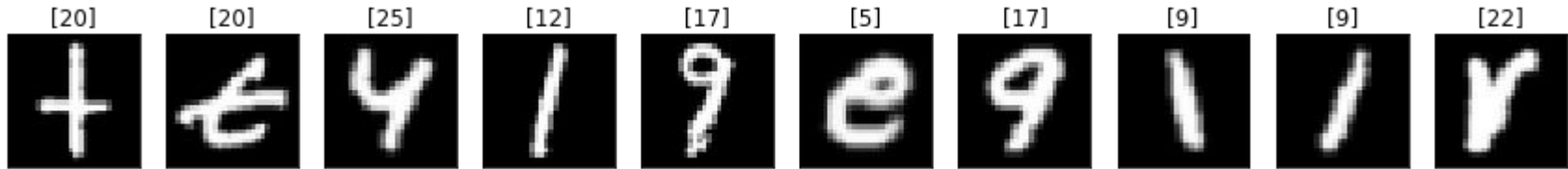
The process of data preparation

Step 1: Collect raw data

- Start with the problem statement
- Identify **input elements** of the problem
- Collect data that closely represents those elements

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[Source](#)

The **size** of a dataset

- Your model should train on **at least an order of magnitude** more data points than trainable parameters
- Simple models on large data sets generally beat fancy models on small data sets

The **quality** of a dataset

The **quality** of a dataset

“Garbage in, garbage out”

The **quality** of a dataset

- Reliability

The **quality** of a dataset

- Reliability
- Feature representation

The **quality** of a dataset

- Reliability
- Feature representation
- Data leakage

Step 2: Identify features and label sources

Step 2.1: Identify features

- Make a hypothesis about a feature(s), test it and then repeat

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- Make a hypothesis about a feature(s), test it and then repeat
- Identify the features such that data leakage is not introduced
- Sometimes more features might lead to Curse of Dimensionality
- Prediction data sources — [online](#) vs. [offline](#)

Step 2.2: Label sources

- Direct label vs. derived labels

Step 2.2: Label sources

- Direct label vs. derived labels
- Human labelling

Step 3: Select sampling strategy

Two situations:


- Need to collect more data :(
- Plenty of data, hence need to use a subset ^_^

Step 3: Select sampling strategy

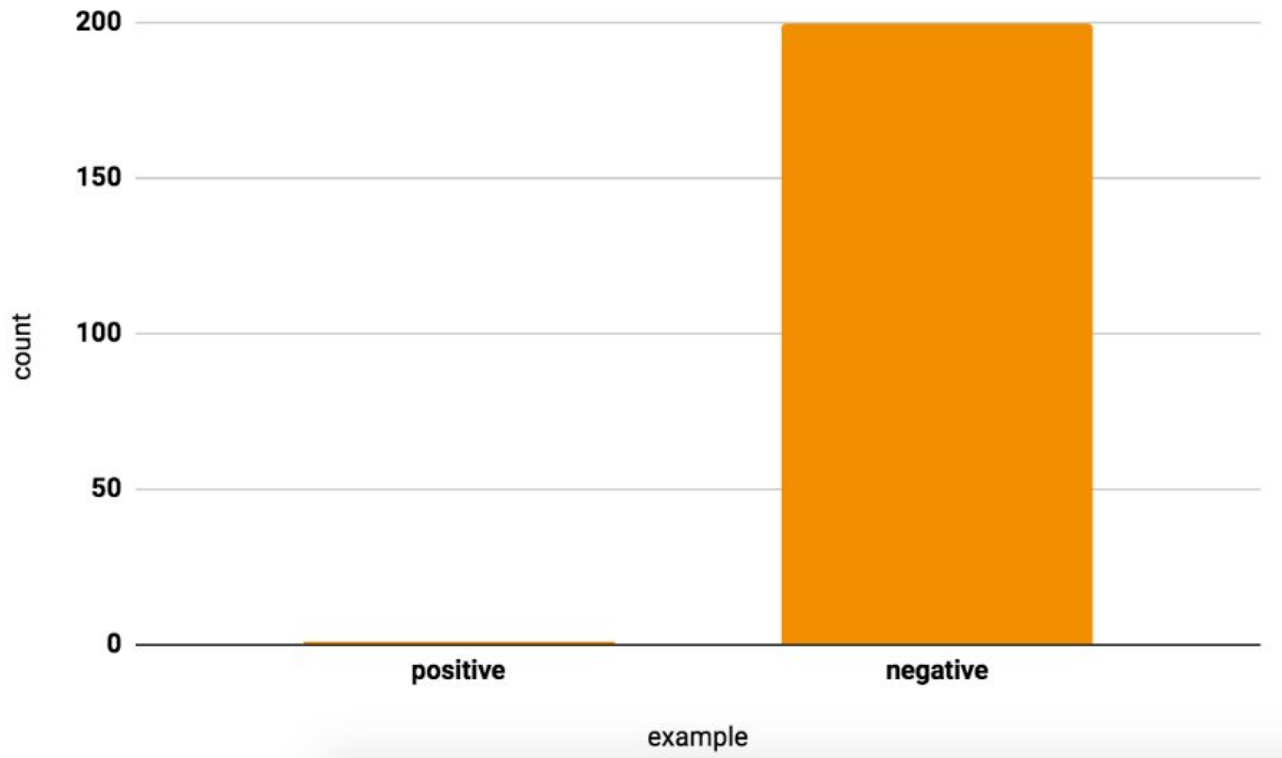
Two situations:

- Need to collect more data :(
- Plenty of data, hence need to use a subset ^_^

How do you do that? :O



The problem of **Class Imbalance**



[Source](#)

Step 4: Data splitting

Three golden splits:

- Training
- Validation
- Testing

Step 4: Data splitting

Two important questions:

- When can we not split the data randomly?
- When can we do that?





Incorporating fairness practices

Why care about it?

Amazon scraps secret AI recruiting tool that showed bias against women

Why care about it?



Amazon scraps secret AI recruiting tool that showed bias against women

 <p>JAMES RIVELLI</p> <p>LOW RISK 3</p>	 <p>ROBERT CANNON</p> <p>MEDIUM RISK 6</p>
<p>JAMES RIVELLI</p> <p>Prior Offenses 1 domestic violence aggravated assault, 1 grand theft, 1 petty theft, 1 drug trafficking</p> <p>Subsequent Offenses 1 grand theft</p> <p>LOW RISK 3</p>	<p>ROBERT CANNON</p> <p>Prior Offense 1 petty theft</p> <p>Subsequent Offenses None</p> <p>MEDIUM RISK 6</p>

Why care about it?

Amazon scraps secret AI recruiting tool that showed bias against women

And many [more](#) ...

 JAMES RIVELLI LOW RISK 3	 ROBERT CANNON MEDIUM RISK 6
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Combating bias

- Identifying bias

Combating bias

- Identifying bias
 - Missing feature values in large numbers
 - Unusual feature values
 - Class imbalance

Combating bias

- Evaluating for bias

Combating bias

- Evaluating for bias
 - Don't fall prey to the **accuracy paradox**
 - Determine the **confusion matrix** of the model
 - Investigate the false predictions made by the model

Wrapping up

- The process of data preparation
 - Collect raw data
 - Identify features and label sources
 - Select sampling strategy
 - Data splitting
- Incorporating fairness practices

References

- [Data Preparation and Feature Engineering in ML](#)
- [How \(and why\) to create a good validation set](#)
- [Becoming One With the Data](#)
- [Data Science from Scratch](#)
- [Machine Learning Fairness](#)

See you next time



Find me here:
sayak.dev

Thank you very much :)



Experts

