Supervised Learning

Session 3

Slides Link: https://teachla.uclaacm.com/classes/ml



ACM AI + ACM TeachLA





What are you going to be for Halloween?







- Artificial Intelligence The theory and development of computer systems able to perform tasks that normally require human intelligence.
- Machine Learning A type of AI that provides computers with the ability to learn without being explicitly programmed.
- **Deep Learning** An ML learning method based on artificial neural networks.



ΑΙ

- if-else statements
- Decision Trees + Bayesian decision-making
- data mining

ML

- supervised learning
- unsupervised learning
- reinforcement learning
- deep learning
 - neural networks
- natural language processing



Supervised Learning?

It's one of the most widely used forms of machine learning!



helped



Let's imagine a basketball

game...



...with 3 seconds left, down by 1!







Our captain (**Mr. Supervisor**) has the ball (**input**)!









Mr. Supervisor passes to Mr. Model who can score!



However, Mr. Model actually doesn't

know how to play basketball 😳

...he was given the ball (input data) but...

...doesn't know what to do (output)







Mr. Model needs a supervisor to point him in the right direction...

Mr. Supervisor is here to tell him the output (what he should do)













In other words...

The **model** uses input *and* labeled output data <u>to map</u> the **input** to the **output**

Now that **Mr. Model** has an idea of the <u>end result</u> (**output**)...

Mr. Model can use the ball (input) to score!



labeled output



Supervised Learning:

The **model** (function) uses input *and* labeled output data to <u>map</u> the **input** to the **output**





f(x) = y

Normal/Al outside ML

- **Given:** input and defined relationship / function
- **Find:** the output



Machine Learning

- **Given:** input and output
- Find: the relationship / function



Supervised Learning:

The **model** (function) uses input *and* labeled output data to <u>map</u> the **input** to the **output**



It's "supervised" because we *tell* the model the labeled output *beforehand*

take a moment to let that sink in!





- A Machine Learning *model* is a big function
- A Supervised Machine Learning model uses <u>labeled</u> <u>output</u> data, hence the name "supervised"
 - Labeled: output data we help *give* the model
- Supervised Learning maps the relationship of: input \rightarrow labeled output







Training Models



Not these types of models!



Mapping? Learning?

output)

- Supervised machine learning is all about **"mapping"** the relationship of input to output
- But how exactly does it **"map?"** (given input and



Mapping? Learning?

- Supervised machine learning is all about **"mapping"** the relationship of input to output
- But how exactly does it "map?" (given input and output)

Let's take a look at our basketball example again...



Recall that **Mr. Model** doesn't know how to play basketball

He doesn't know how to:

- shoot
- dribble
- dunk!

He has to **<u>train</u>** to learn

those skills!





So in order to **map** input to output consistently...





So in order to **map** input to output consistently...

Mr. Model must train so he gets good at
scoring baskets
(input → output)











A Model in Training

- A model "training" means it's *learning* ways to correctly *map* input data to output data
- Hence "machine *learning*"
- "Training" and "Learning" are interchangeable terms



A "Trained" Model

So after hours of training... **Mr. Model** is now <u>trained</u> and able to score hoops!

Training a machine learning model is a long, hefty mathematical process

We'll learn how it specifically works in the coming weeks!



Me: *uses machine learning* Machine: *learns* Me:





How to Effectively Train Models



During preseason, in preparation for the NBA season **Mr. Model** trains *a lot*

He trains by playing against the B

team, his **training data**







A **training dataset** is any dataset that allows our model to *train* (map input -> output)



Let's take a closer look at our B Team data!



B Team

In our case, the <u>B Team</u> is our <u>training dataset</u>: **Mr. Model** trains on them in order to learn how to play bball



The average height of the B Team is 4 feet...

So Mr. Model just dunks on them everytime

It's so effective, it becomes the <u>only</u> thing **Mr. Model <u>learns</u> how to do

Testing our training abilities

Now that preseason is over, it's time to *test* Mr.

Model's new bball skills...

...on the NBA!







Test Dataset

When we want to test to see if our model is well trained, we

test our model on a test dataset





Test Data (NBA)

The **test dataset** is an example of what the model will see in the real world... it's a true test of its abilities!



As we test **Mr. Model** plays against the <u>tall</u>, <u>skilled</u> players of the NBA we almost forgot...

...that he only knows how to dunk!



He's only able to dunk on people that are 4 feet tall! (B Team)



Did **Mr. Model** have good training? Especially if he plans to play against the <u>tall</u>, <u>skilled</u> players of the NBA?





8 feet!

(In)effective Training (cont'd)

Did **Mr. Model** have good training? Especially if he plans to play against the tall, skilled players of the NBA?





Mr. Model has become so accustomed to dunking on the short players of the B Team that...



...he sucks at playing against anyone taller than 6 feet because dunking won't work on them!

Mr. Model had bad training!



Whenever a model has become too accustomed to a

specific dataset (playing against B team too much), it's called

overfitting.

- The model is too mode-specific
- Not "general" enough to handle real-world randomness
 - Randomness of players: tall AND short, fast AND slow, good AND bad, clueless AND intelligent

Overfitting (cont.)

- **Overfitting** occurs when model becomes too
 - sample-specific (only knows how the B Team plays)
- Generally, a model *should* be trained to perform well on the entire population of data (well-rounded players)
 - Better for Tesla cars to drive <u>all</u> streets decently well

rather than <u>one specific</u> street perfectly

Combatting Overfitting

What are some ways **Mr. Model** could've avoided overfitting?

- Playing against diverse set of teams/players
 - $\circ\,$ Use a diverse input/output dataset for your model
- Train more! Train skills *other* than dunking too

• Using *more* diverse data for your model <u>always</u> helps

Combatting Overfitting (cont.)

What does this tell us about the **training dataset**?

- Our training dataset was NOT diverse enough!
- Ideally, we want the training dataset to fairly represent the *majority* of the population









Avoid training against a small subset of your population, or else you'll get creamed like **Mr. Model in the NBA!



Conversely, if **Mr. Model** is lazy (doesn't have much data to train on) he can be susceptible to **underfitting**



Underfitting: when the model is too *basic*, and doesn't work effectively in most situations

How do we combat this?

Gather more data to train on!

Training (graphically)

• Predictor variable: any input variable used to measure output



Training (graphically)

• Predictor variable: any input variable used to measure output

The black dotted line represents our model.







¡Kahoot!



Homework (optional)



Closing Comments



- 1. Supervised ML Models are just big functions!
- 2. ML Models "train": finding the <u>relation</u> between input \rightarrow output
- 3. Train on <u>A LOT</u> of diverse data to avoid overfitting/underfitting





Thanks!

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