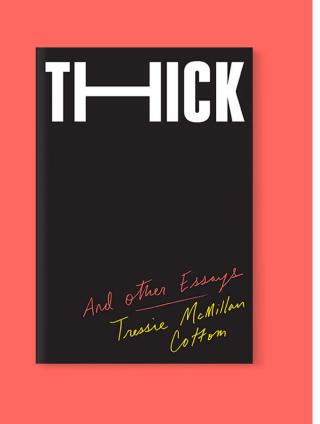
ML and Society

Feb 15, 2023

Passcode: Tressie McMillan Cottom

• A preeminent public scholar on race, digital media, and higher ed If you want to see science writing done right, go read one of her books





Sit with your team!

Team 1	Afzal	Cole	Navid	Tim	
Team 2	Aishwarya	Herman	Mads	Melvin	
Team 3	Daphkar	Juliana	Ibtida	Monica	
Team 4	Joe	Ken	Vedant	Zach	
Team 5	Chaitanya	Evan	Hitesh	Sushanth	
Team 6	Hannah	Harinee	Gabriella	Suradhya	
Team 7	Alex	Connor	Gopi	Shane	Thanh
Team 8	Aditi	Connor	Jason	Mitali	
Team 9	Botsalano	Niharika	Vedang	Yunmei	
Team 10	Dhiraj	Frank	Kashyap	Michael	

Who does my machine learning model serve? How do I know? What can I do about it?

So who is right?



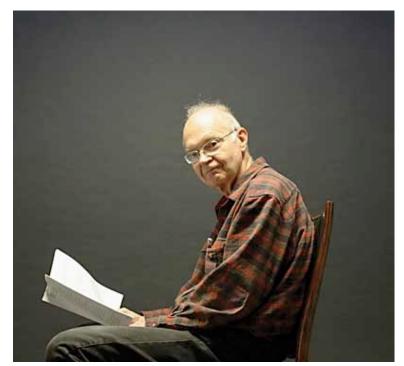
"Algorithms can be biased"



"Algorithms are based on math and hence cannot be biased"

Knuth's Definition of an algorithm

An algorithm is a finite, definitive, effective procedure with some input and some output



Main Steps in Traditional Algorithm Design

Problem Statement Real world problem Problem definition **Problem Definition** Precise mathematical def Algorithm Algorithm design "Implementation" **Data Structures** Analysis Correctness/Run time

Two main points about traditional algorithms

Problem is defined BEFORE algorithm is designed

Can prove algorithm is correct for ALL imuts

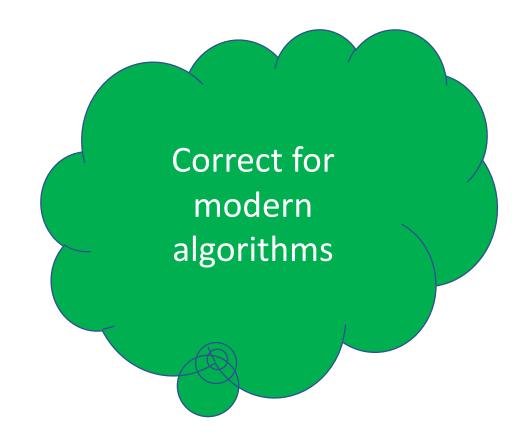
Though really you would be wrong there as well since the problem statement itself could be biased...

One *could* consider such algorithms not biased

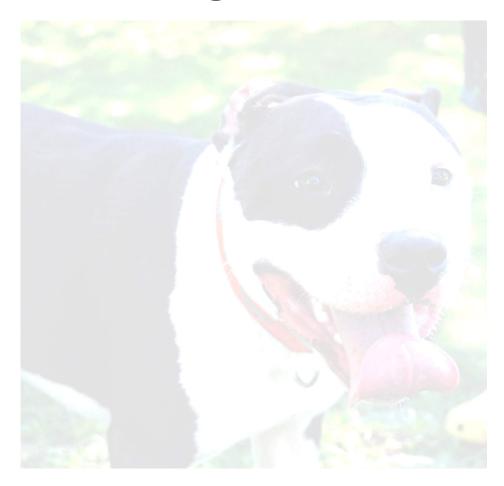
Today

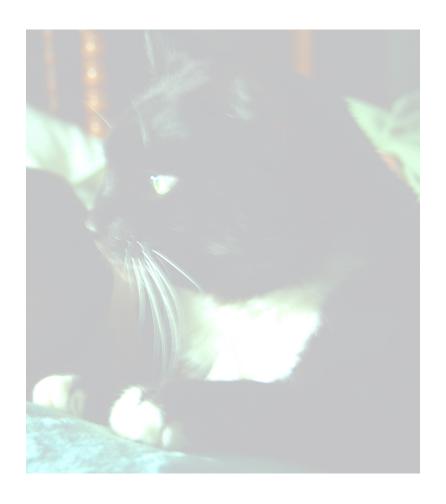


"Algorithms can be biased"

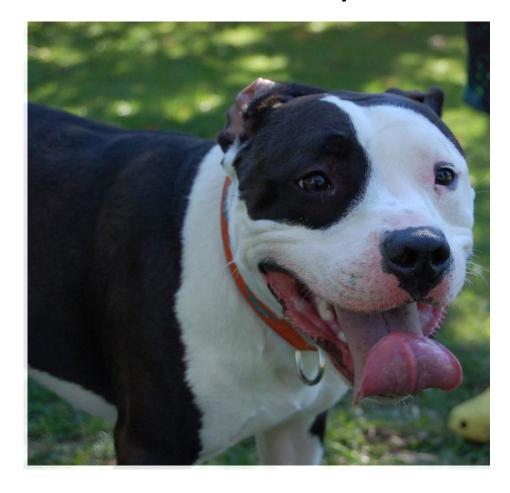


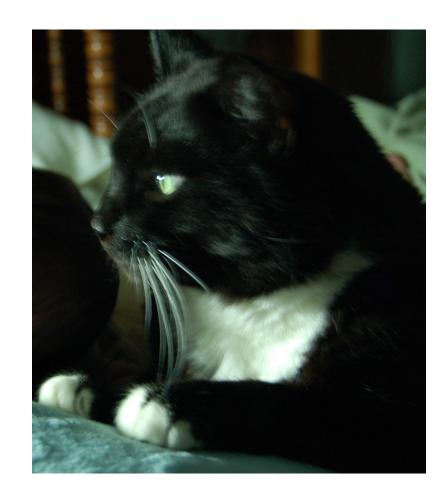
Cat vs. Dogs



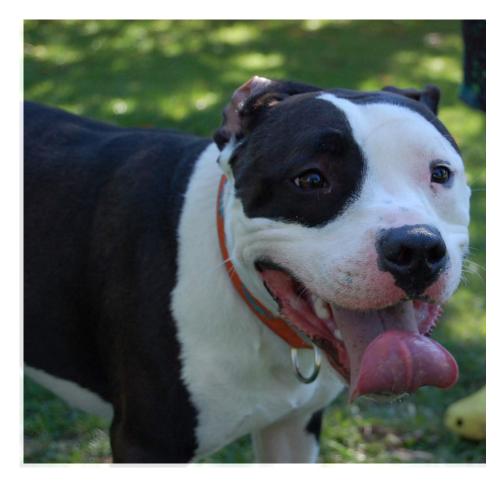


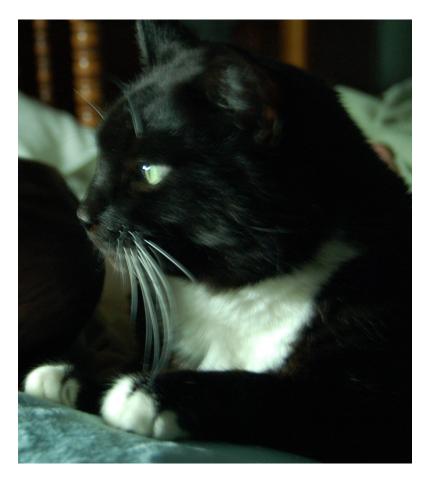
Warren and Billy





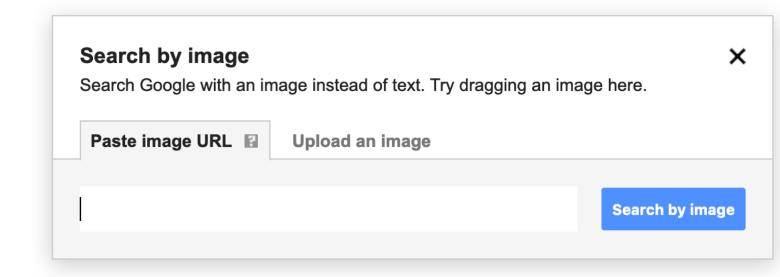
How do you "define" a dog vs cat image?





Google Images has "solved" this problem

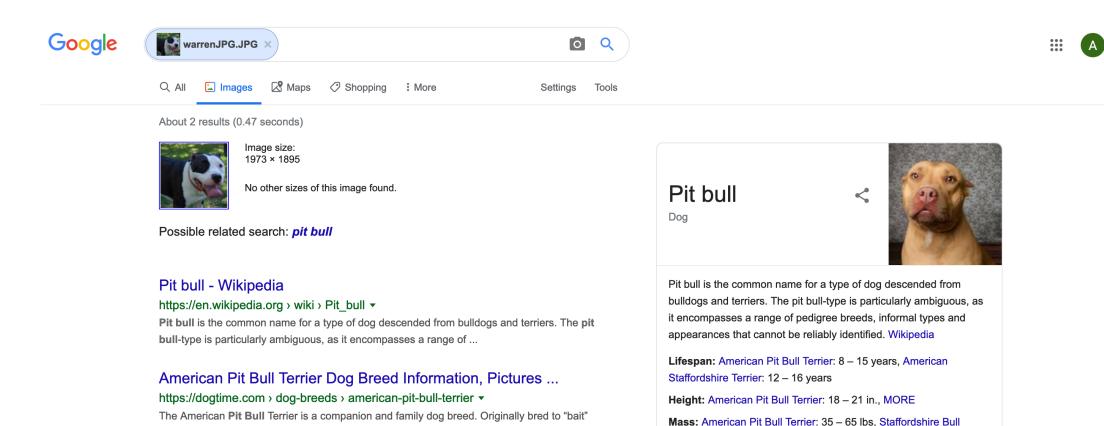




My result for Warren (Spring 20)

bulls, the breed evolved into all-around farm dogs, and later moved ...

Visually similar images

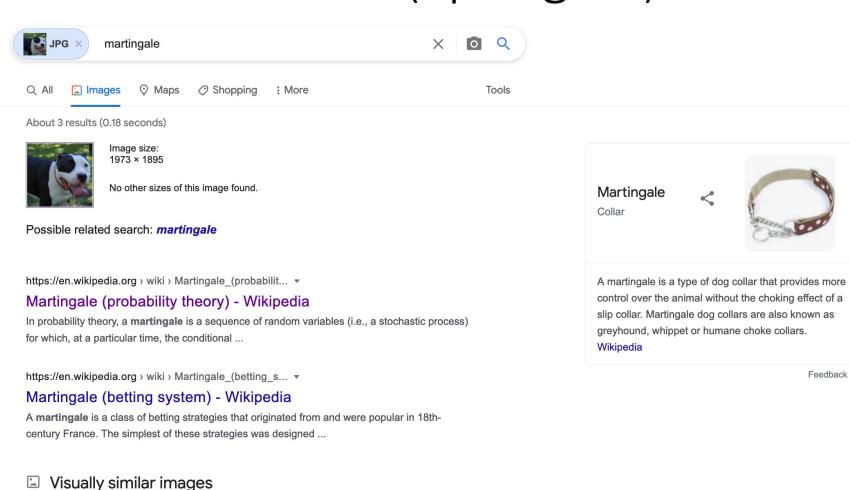


Terrier: 29 – 37 lbs

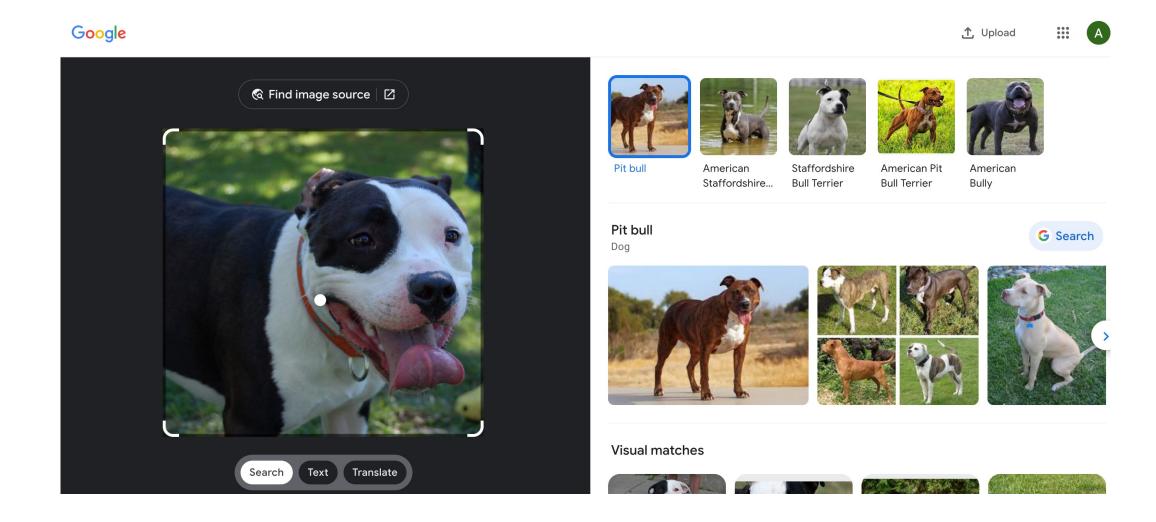
ranges from 10 to 12 years. petwave.com

Life span: The average life span of the American Pit Bull Terrier

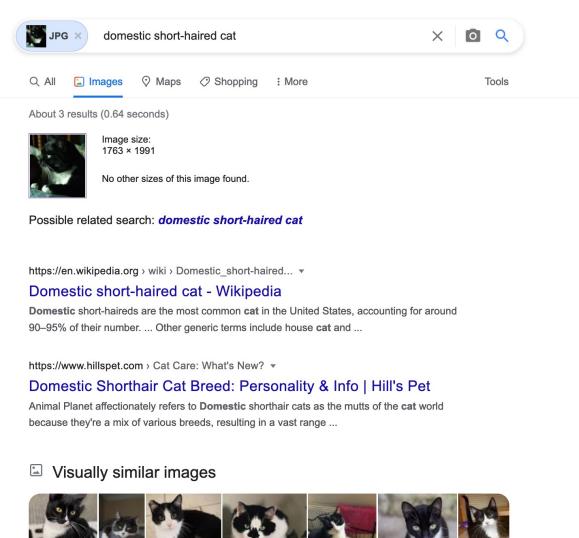
My result for Warren (Spring 22)



My result for Warren (Spring 23)



My result for Billy (Spring 20+22)



Domestic short-haired cat



A domestic short-haired cat is a cat of mixed ancestry-thus not belonging to any particular recognised cat breed—possessing a coat of short fur. In Britain they are sometimes colloquially called moggies. Wikipedia

Shorthair cat breeds



Shorthair



Shorthair



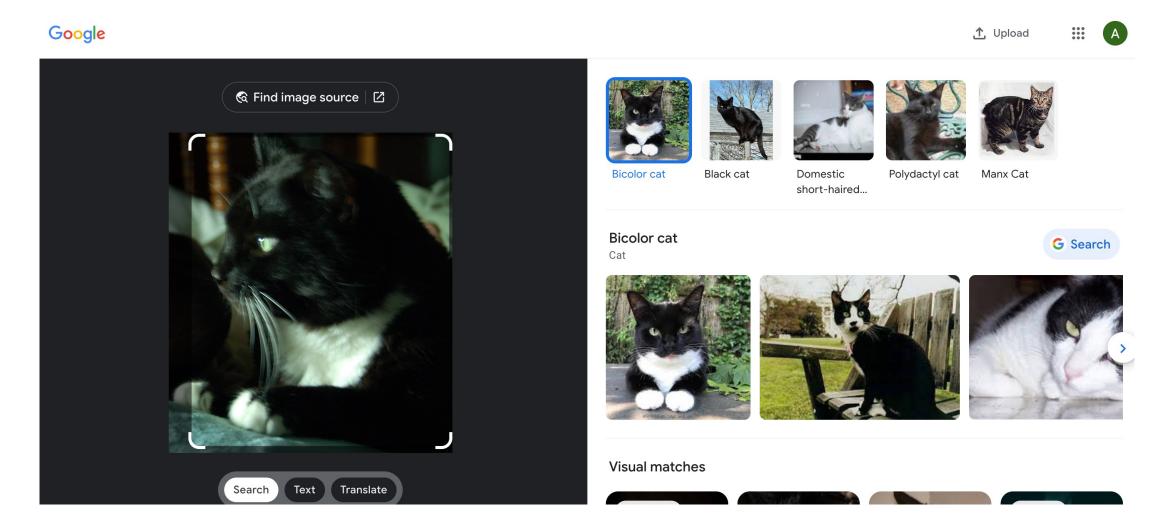


View 45+ more

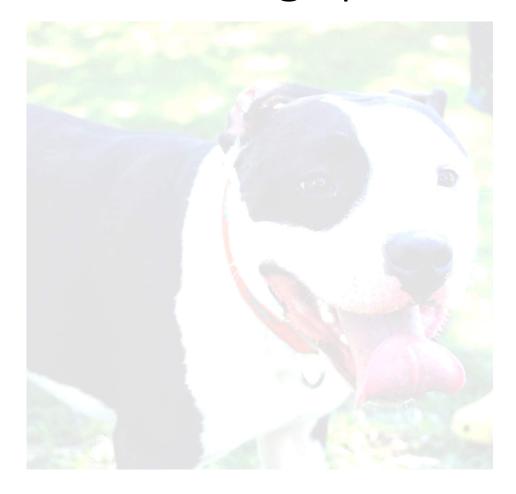
Maine Coon

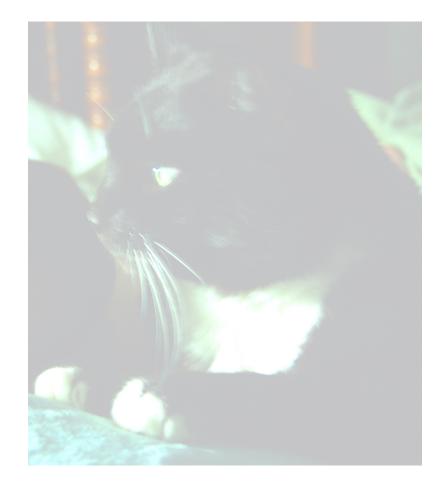
Feedback

My result for Billy (Spring 23)



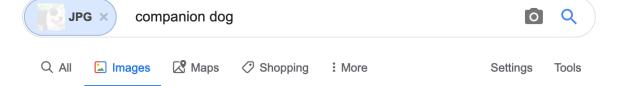
So cats vs dogs problem solved?





My result for modified Warren (Spring 20)





About 2 results (2.32 seconds)



Image size: 1973 × 1895

No other sizes of this image found.

Possible related search: companion dog

Companion dog - Wikipedia

https://en.wikipedia.org > wiki > Companion dog ▼

A companion dog is a dog that does not work, providing only companionship as a pet, rather than usefulness by doing specific tasks. Many of the toy dog breeds ...

Best Companion Dog Breeds | Purina

https://www.purina.com > Dogs > Dog Breeds > Collections ▼

Whether you want a friendly face to come home to or the best companion dog breed for an elderly parent, get the complete list here.

Visually similar images

Companion dog



A companion dog is a dog that does not work, providing only companionship as a pet, rather than usefulness by doing specific tasks. Many of the toy dog breeds are used only for the pleasure of their company, not as workers. Wikipedia

Companion dog breeds



Pekingese Yorkshire

Terrier





Spaniel





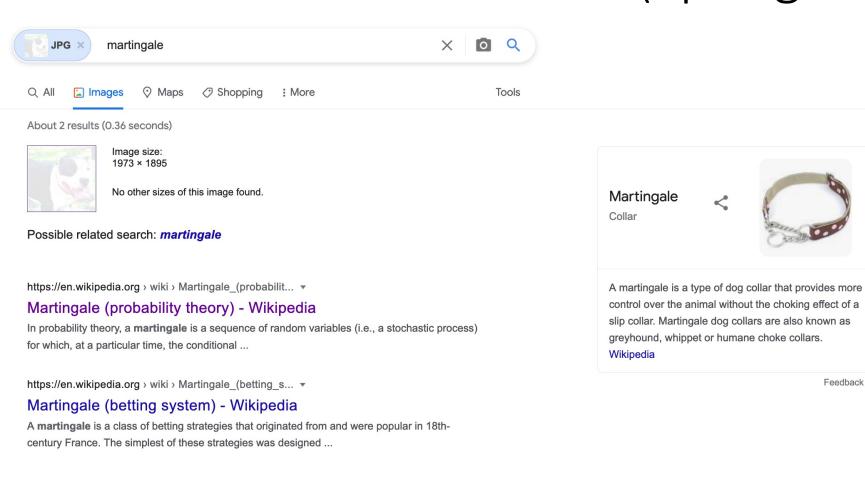


View 4+ more

Valley **Bulldog**

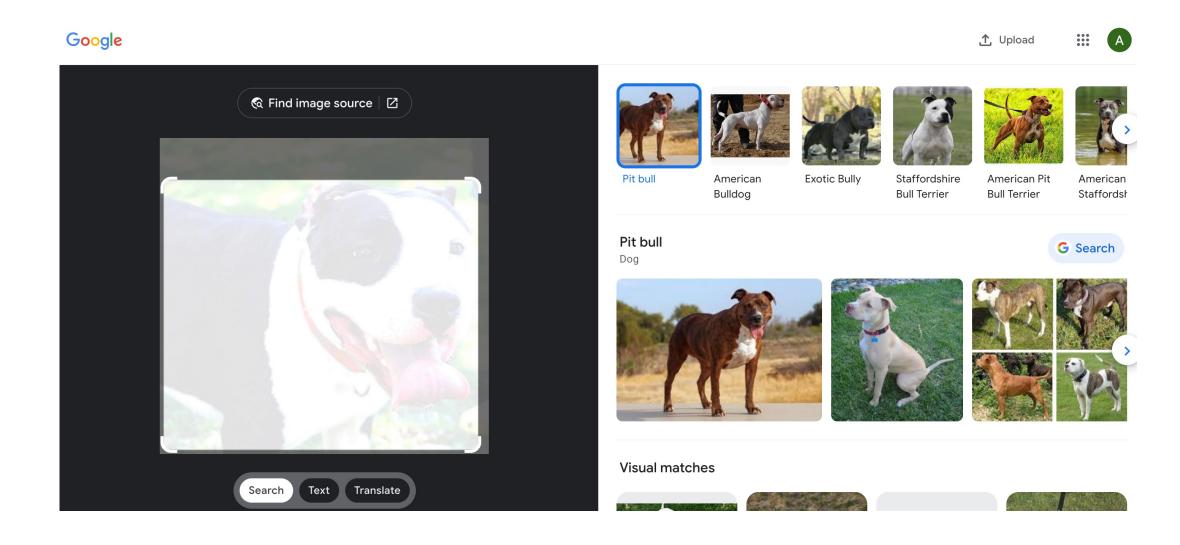
Borador

My result for modified Warren (Spring 22)

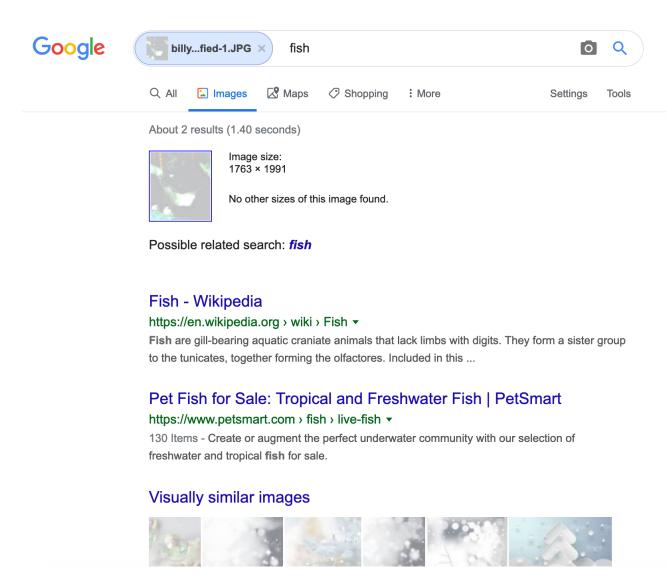


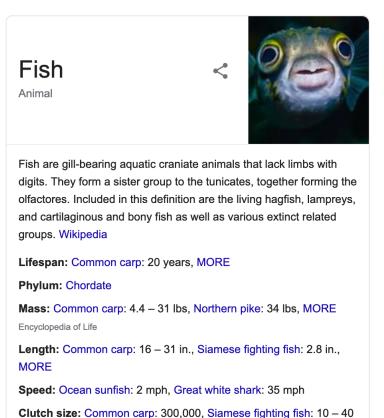
Visually similar images

My result for modified Warren (Spring 23)

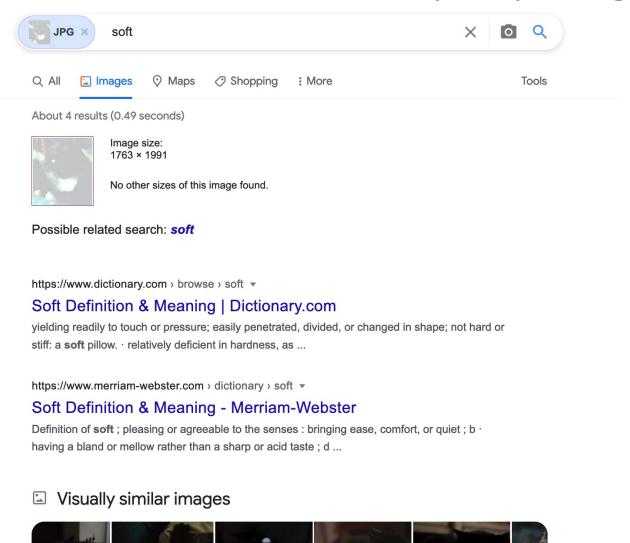


My result for modified Billy (Spring 20)

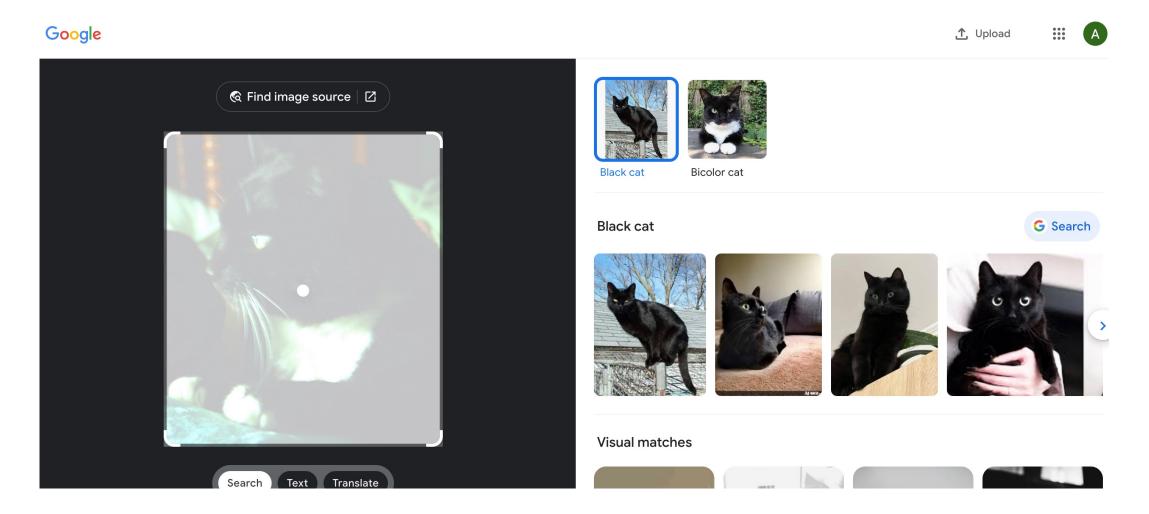




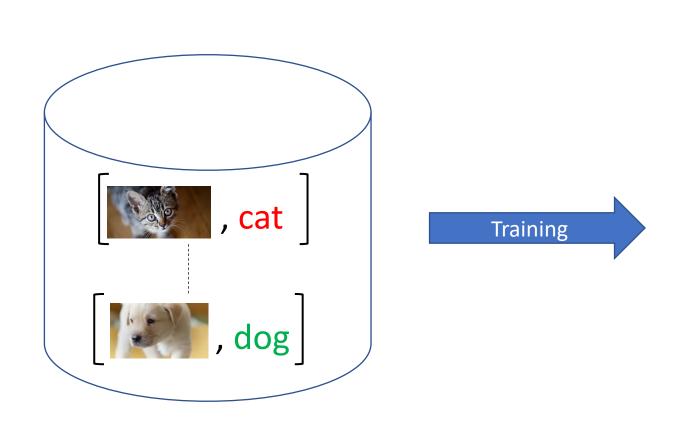
My result for modified Billy (Spring 22)



My result for modified Billy (Spring 23)

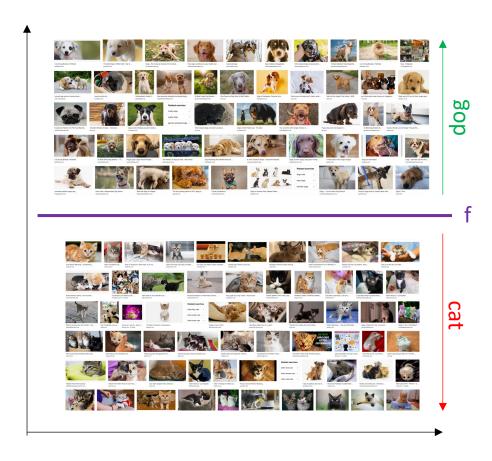


How does Google Images work?





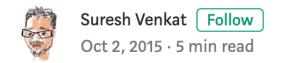
When a new image comes in





When an algorithm isn't...

Cc





Go

The popular press is full of articles about "algorithms" and "algorithmic fairness" and "algorithms that discriminate, (or don't)". As a computer scientist (and one who studies algorithms to boot), I find all this attention to my field rather gratifying, and not a bit terrifying.

)gs

What's even more pleasing is that the popular explanation of an algorithm follows along the lines of the definition we've been using since, well, forever

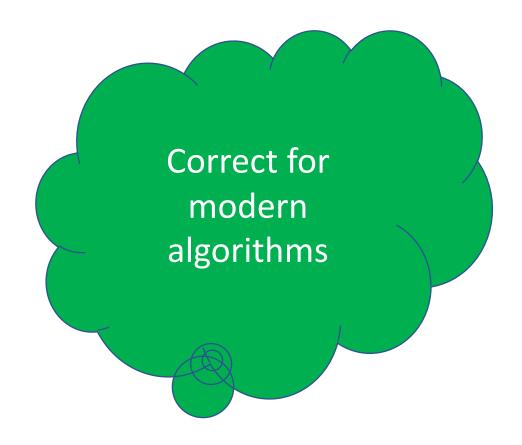
An algorithm is a set of steps (the instructions) each of which is simple and well defined, and that stops after a finite number of these steps.

If we wanted a less intimidating definition of an algorithm, we turn to the kitchen:

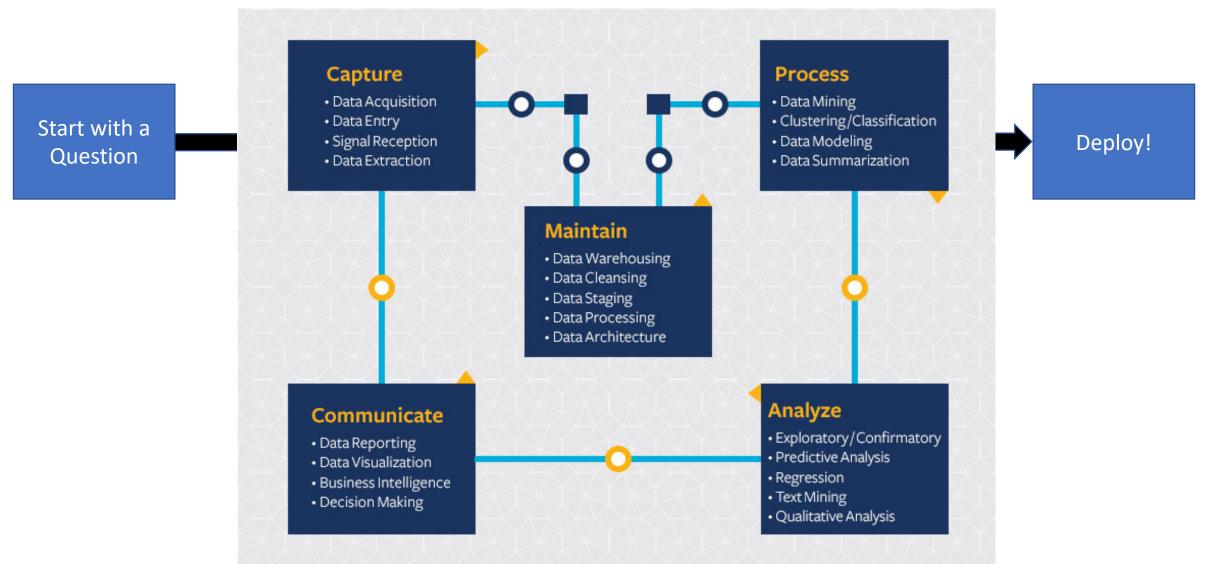
AOC is right!



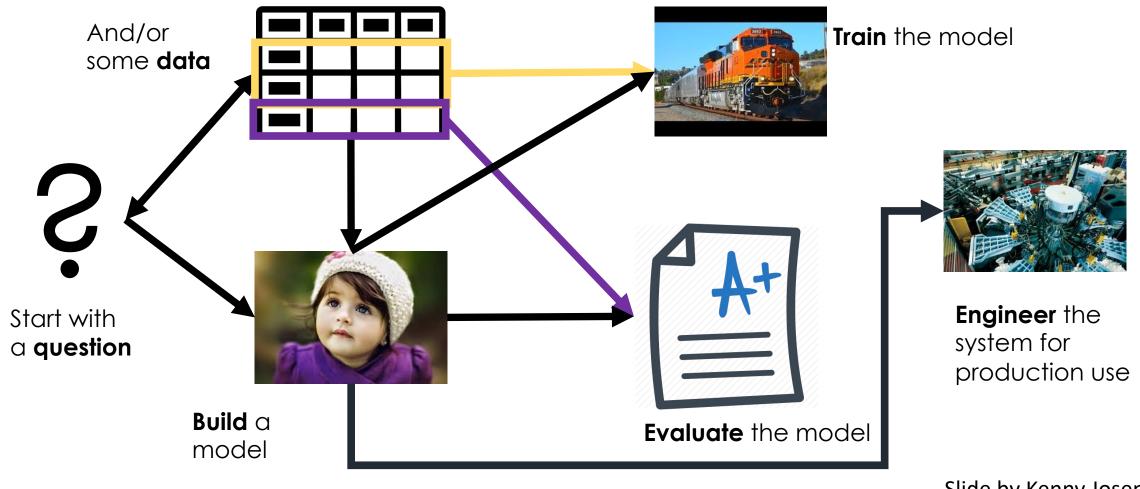
"Algorithms can be biased"



Modern Algorithm = Machine Learning (ML)

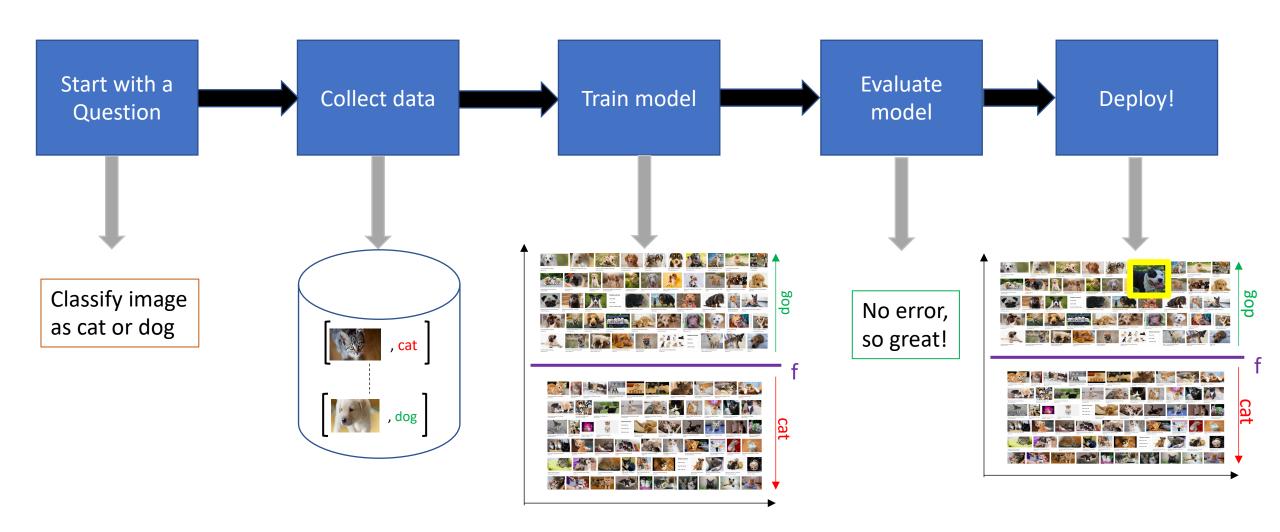


The Machine Learning Pipeline



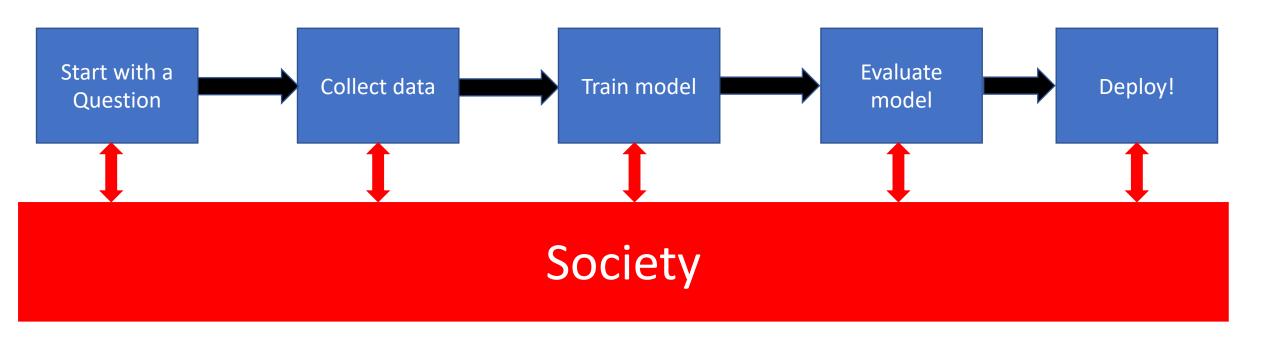
Slide by Kenny Joseph

Back to cats vs. dogs

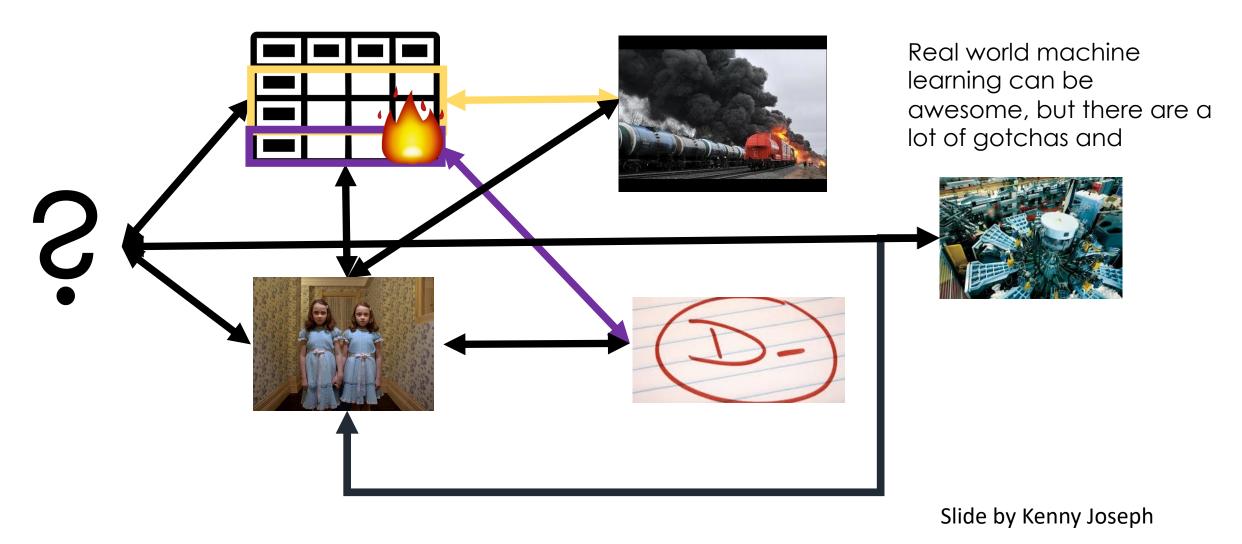




What is missing from this picture?



A real view of the ML Pipeline







Debugging machine learning

I've been thinking, mostly in the context of teaching, about how to specifically teach debugging of machine learning. Personally I find it very helpful to break things down in terms of the usual error terms: Bayes error (how much error is there in the best possible classifier), approximation error (how much do you pay for restricting to some hypothesis class), estimation error (how much do you pay because you only have finite samples), optimization error (how much do you pay because you didn't find a global optimum to your optimization problem). I've generally found that trying to isolate errors to one of these pieces, and then debugging that piece in particular (eg., pick a better optimizer versus pick a better hypothesis class) has been useful.

For instance, my general debugging strategy involves steps like the following:

1. First, ensure that your optimizer isn't the problem. You can do this by adding "cheating" features -- a feature that correlates perfectly with the label. Make sure you can successfully overfit the training data. If not, this is probably either an optimizer problem or natural language processing (NLP), computational linguistics (CL) and related topics (machine learning, math, funding, etc.)

About Me



hal

View my complete profile

Labels

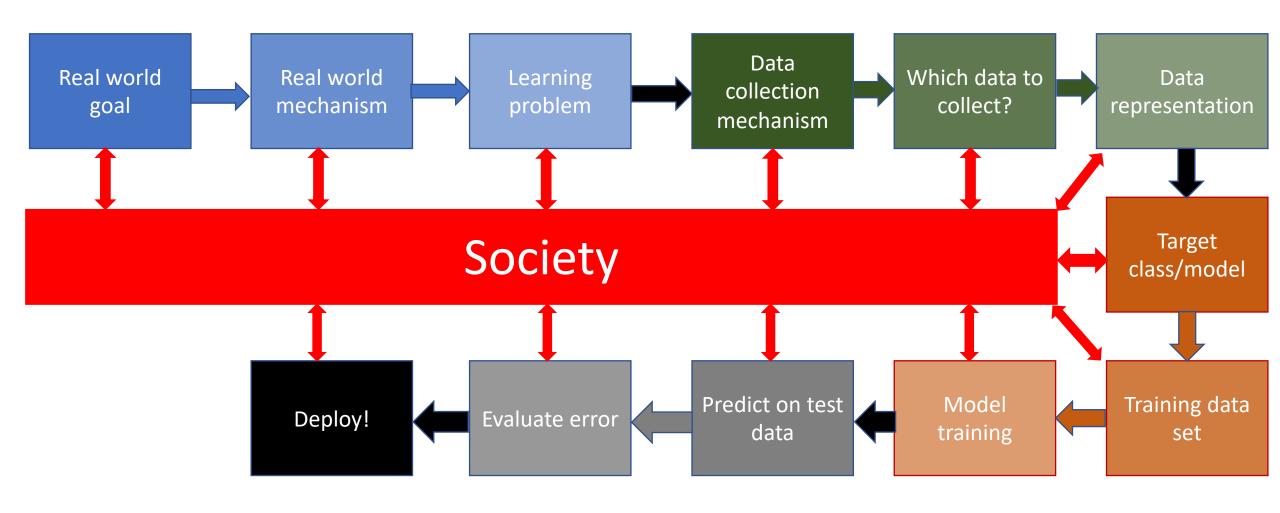
acl (3) ACS (2) advising (1)

Data representation

> Target class/model

Training data set

What is missing?



Not the only ML+society pipeline in town

INSIDE AI

Black-Boxed Politics:

Opacity is a Choice in Al Systems



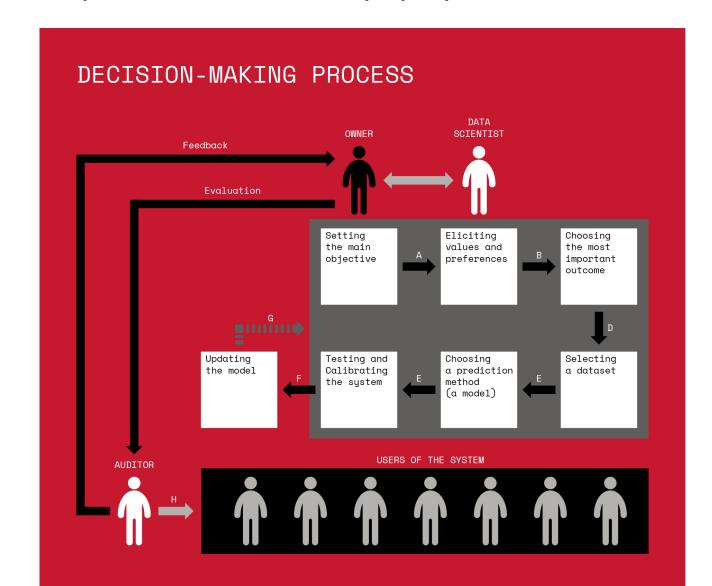


Written by: Agata Foryciarz, Daniel Leufer, Katarzyna Szymielewicz

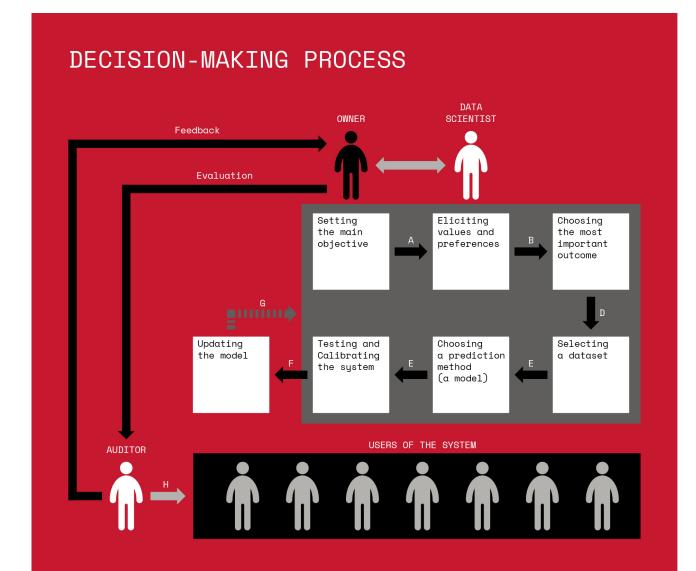
Illustrations by: Olek Modzelewski

Artificial intelligence captures our imagination like almost no other technology: from fears about killer robots to dreams of a fully-automated, frictionless future. As <u>numerous authors have documented</u>, the idea of creating artificial, intelligent machines has entranced and scandalized people for millennia. Indeed, part of what makes the history of 'artificial intelligence' so fascinating is the mix of genuine scientific achievement with myth-making and outright deception

Not the only ML+society pipeline in town



A walkthrough



HUMAN DECISIONS THAT SHAPE AN AI SYSTEM

Setting the main objective

his decision, made by the owner of the system, s likely to be formulated in business or eginning of a longer conversatio

We want to support patients with the greatest health needs by enrolling them in dedicated medical support programs.

We want to identify patients with medical conditions for which treatment is currently under-resourced in our hospital system (eg. eating disorders).

Eliciting values and preferences

This stope can involve convercations with various stakeholders she will be involved with the automated decision support system to understand their needs. Those needs will be balanced against various limitations (budget restrictions, constraints that arise from interpreting the main objective sathwardically), as well as potentially comparing objectives between stakeholders.

A patient advacacy group mants the resources to be allocated to the underfunded eating disorder treatment program, while physiciann insist the program should be used primarily to support diabetics and the elderly.

In the world of data science, one cannot expect the system to achieve multiple diverse objectives at once - rather, a single outcome has to be defined.

At this stage of the conversation the comme and the data scientist will have to decide which one is the most important. This is where trade-offs and dilemmas kick-in.

We want to identify all patients who could

Selecting a dataset

In the real world, access to high-quality and lawfully collected data is limited. At this stage our data exientist will have to compose - from everything that was available (incl. purchases from data brokers) and lawfully collected - a dataset that will be comprehen-

Indeed, in many cases the data that is available will determine data scientist's choice of options for the mathematical interpretation of the objective.

Private electronic health records of the hospital's patients, along with diagnosis codes and medications prescribed - relevant to the task at hand, but there are privacy and data quality concerns.

Choosing a prediction method (a model)

his step involves formulating a loss func-tion - a mathematical formulation of the model's goal. It is composed of an error metric, which typically measures the average error of the model's predictions, and can include additional constraints, which mudge include additional constraints, which madge the model towards desired behaviors and away from worst-case scenarios (e.g. a constraint can be imposed to prevent a situation where errors are not mostly incurred by minority populations.

work in preparing data.

Loss function: Standard cross-entropy loss with L1 regular-ization (pushing the model to drop the least important elements of input completely).

Model: A neural network - can sometimes produce more accurate predictions, requires less data pre-processing, but is not easily interpreta-ble, and can generate counterintuitive predic-tions that are hard to trace in some cases.

a fairness constraint, ensuring equal rates o correct predictions for men and women.

Testing and Calibrating the system

The model has to be tested, using training data. Our data scientist sill now look at the errors that have sourced. Errors con be summarized by multiple metrics. At this stage data scientists choose metrics that are most relevant to evaluate the model. Essed on test results, they decide which errors are acceptable (i.e. how harmful it would be proclument).

Updating the model

If the test results are not satisfactory, this is the signal to redesign one or more components of the system. The decision regarding which component to update and how is most often controlled by the data scien-

Let's change our most preferred outcome. Let's add mathematical constraints in order to change model behavior.

Let's rethink the main objective!

Evaluating before deployment

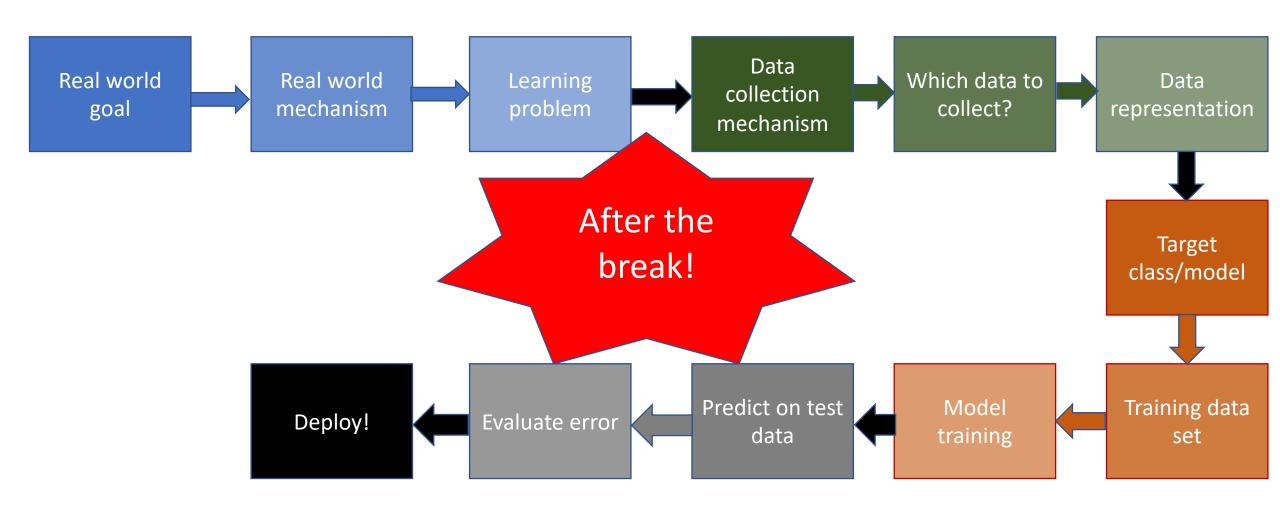
e decision to evaluate the system in

By contrast, in more mature engineering Melds, such as civil engineering, there is a well-defined set of required tests and measurements that have to be reported for

Asking the following questions:
-Does the model perform as well in the new context as it did when we built it on our context as it did when we built it on our ... How does the system perform with regard to various groups of people (e.g. different genders, socioecomouth status, copy)?

-Is there only undersimable shahoufor we howen't anticipated?
-Is there only undersimable shahoufor we howen't anticipated?
-Is the context of the context

A walkthrough



Have you heard of COMPAS?

COMPAS (software)

From Wikipedia, the free encyclopedia

COMPAS, an acronym for Correctional Offender Management Profiling for Alternative Sanctions, is a case manag Equivant (2) used by U.S. courts to assess the likelihood of a defendant becoming a recidivist. [1][2]

COMPAS has been used by the U.S. states of New York, Wisconsin, California, Florida's Broward County, and oth

Contents [hide]

- 1 Risk Assessment
- 2 Critiques and legal rulings
- 3 Accuracy
- 4 Further reading
- 5 See also
- 6 References

Risk Assessment [edit]



Broward County

County in Florida

Broward County is a county in southeastern Florida, US. According to a 2018 census report, the county had a population of 1,951,260, making it the second-most populous county in the state of Florida and the 17th-most populous county in the United States. The county seat is Fort Lauderdale. Wikipedia

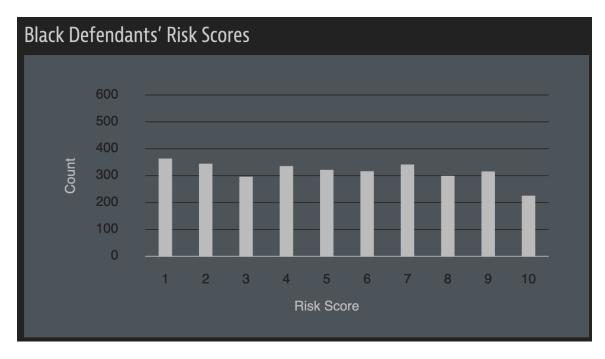
Incorporated cities: 24

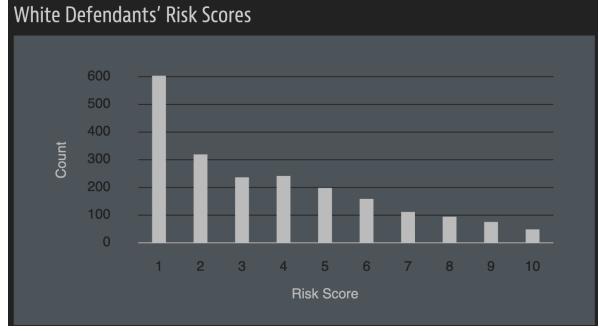
Population: 1.936 million (2017)

Mayor: Mark D. Bogen



A sample of their result





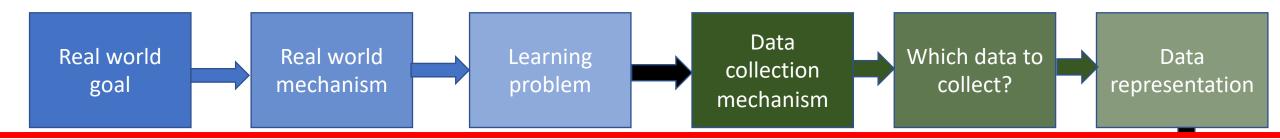
38 FEDERAL PROBATION Volume 80 Number 2

False Positives, False Negatives, and False Analyses: A Rejoinder to "Machine Bias: There's Software Used Across the Country to Predict Future Criminals. And It's Biased Against Blacks."

Anthony W. Flores
California State University, Bakersfield
Kristin Bechtel
Crime and Justice Institute at CRJ
Christopher T. Lowenkamp
Administrative Office of the United States Courts
Probation and Pretrial Services Office

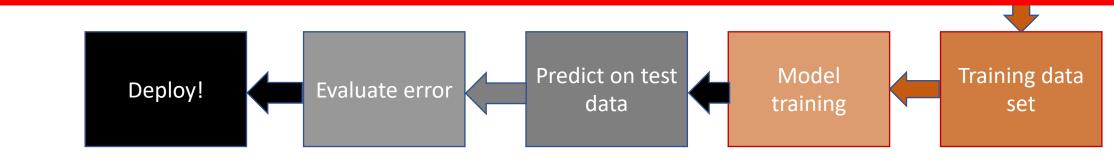


A walkthrough



The Problem

Imagine a situation where the creator of COMPAS had access to the COMPAS dataset. In particular, you are in the team that wants to predict recidivism based on the COMPAS dataset. How would you go about doing it?



Real world goal

Real world goal

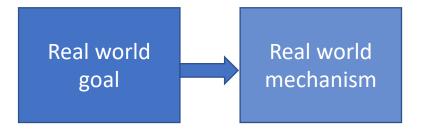
Real world goal

Reduce crime in society.

The Problem

Imagine a situation where the creator of COMPAS had access to the COMPAS dataset. In particular, you are in the team that wants to predict recidivism based on the COMPAS dataset. How would you go about doing it?

Real world mechanism



Real world mechanism

Based on some studies (or not!), your superiors decided that repeat offenders contribute most to crime. This in turn they decided would mean that if one could identify who would commit a crime again in the future, then one could use this information when making judgment on the current crime. Thus, they decided they wanted a system that can identify folks who will re-offend in the future and then promptly handed off the problem to your group to solve it.

The Problem

Imagine a situation where the creator of COMPAS had access to the COMPAS dataset. In particular, you are in the team that wants to predict recidivism based on the COMPAS dataset. How would you go about doing it?

Learning problem





Learning problem

Your group decides on the simplest learning problem: given a defendant *predict* if they will re-offend or not (in other words you are doing *binary classification* (binary because you are "labeling" defendants as either going too re-offend or not going to re-offend and you are doing classification because you are putting people into the two bins-- i.e. giving them a binary label and hence assigning them a "class."

There is another related option (which is what COMPAS \mathbb{Z} : instead of assigning defendants to two scores: they assign a score from 1 (being least likely to re-offend) to 10 (most likely to defends). This range of score (rather than a binary classification) could potentially be more useful to the end user of your system.

However, for our discussion (and indeed for most of the rest of the course), we will focus on binary classification.

The Problem

Imagine a situation where the creator of COMPAS had access to the COMPAS dataset. In particular, you are in the team that wants to predict recidivism based on the COMPAS dataset. How would you go about doing it?

