



THE PPM FRAMEWORK: A NEW WAY OF AUDITING ALGORITHMIC SYSTEMS

A Conceptual Introduction to Machine Learning

Kareem Saleh, Founder and CEO of FairPlay
January 19, 2022



+ What is Machine Learning?

Traditional Programming



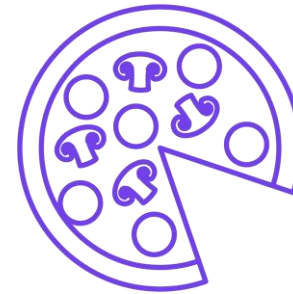
Ingredients

+



Recipe

MAKES
=



Outcome

+ But what if you don't have a recipe?

Traditional / Programming



Ingredients

+



Recipe

MAKES
=



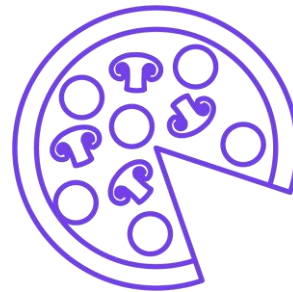
Final Product

Machine Learning Algorithm



Ingredients

+



Outcome

FIGURES OUT



Recipe

+ How is Machine Learning Being Used in Housing?



TENANT SCREENING



MORTGAGE UNDERWRITING



HOME APPRAISAL

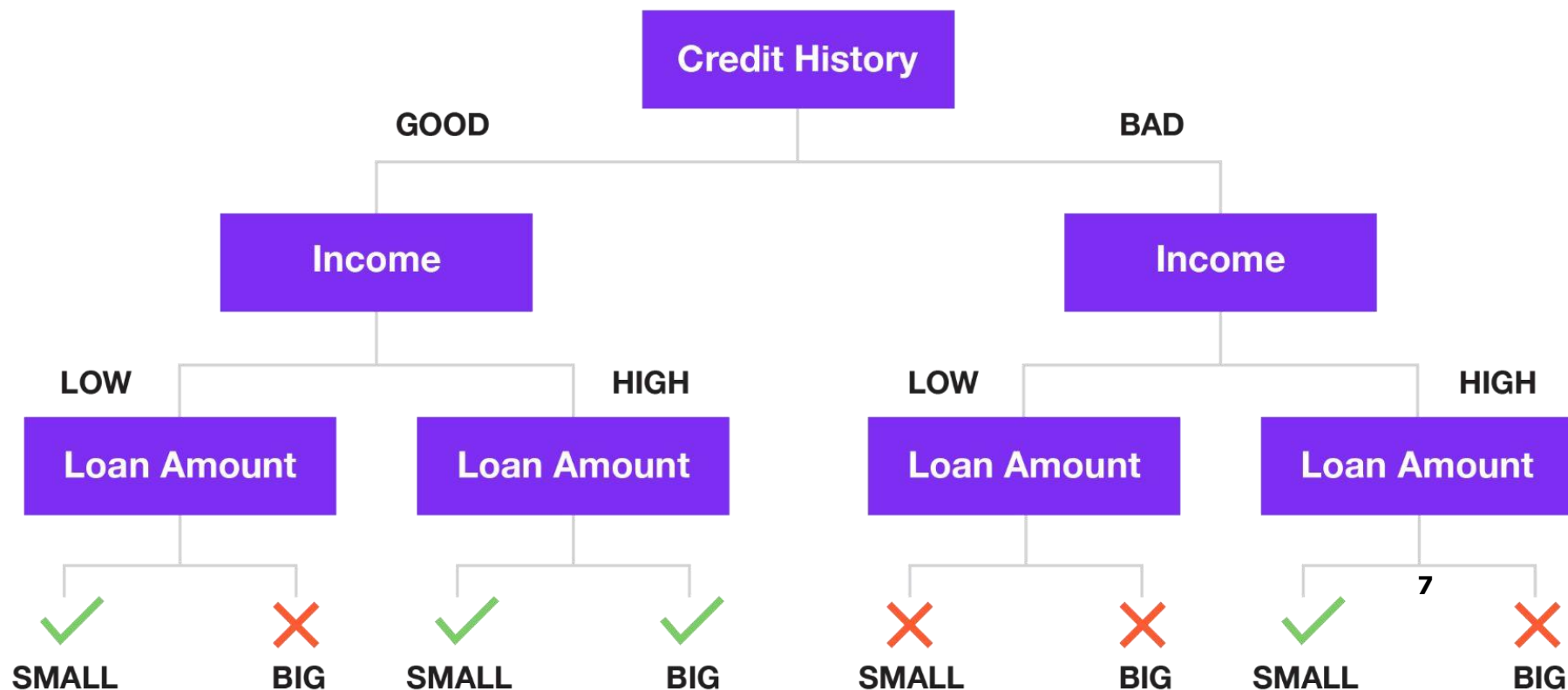
- + **Lending decisions used to be made by humans who tried to assess an applicant's creditworthiness**



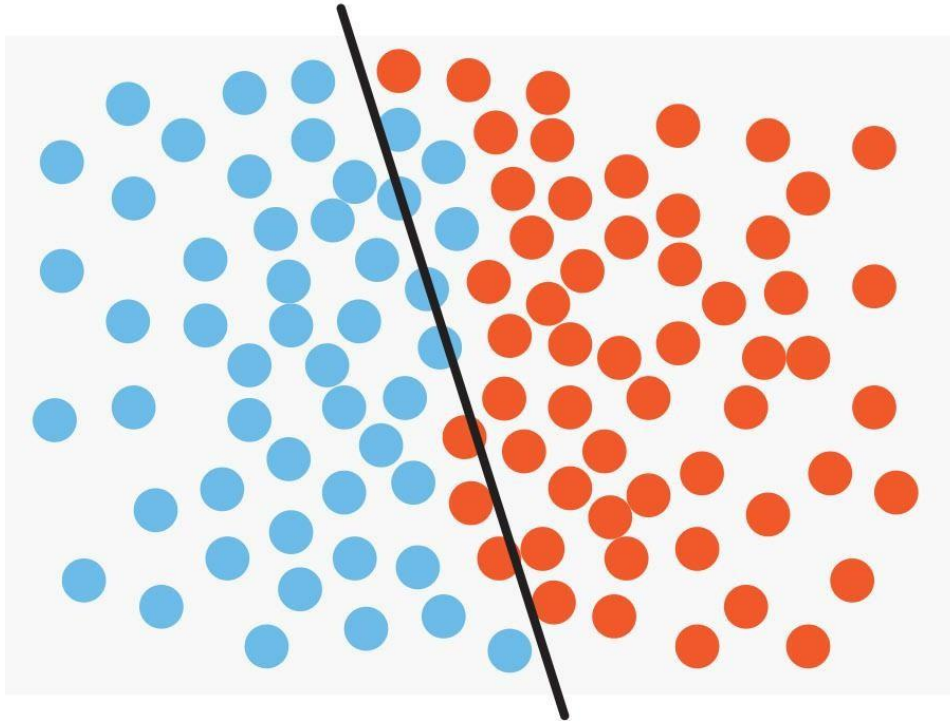
LOAN OFFICER

- + In the 1980s we started using math instead of humans to make lending decisions because it was seemingly more “neutral” and “objective”

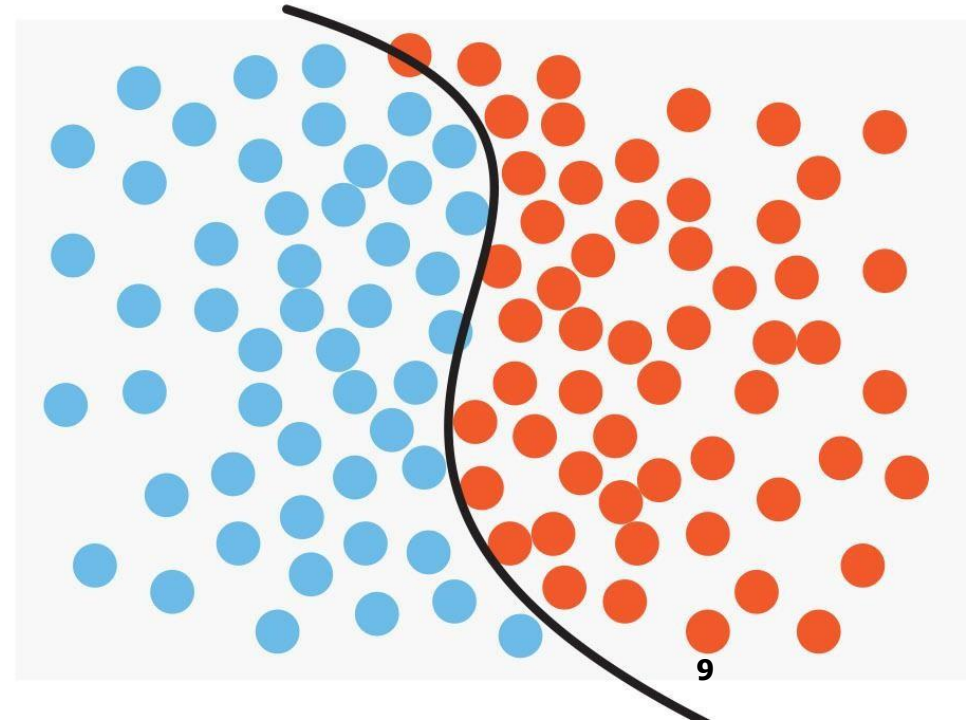
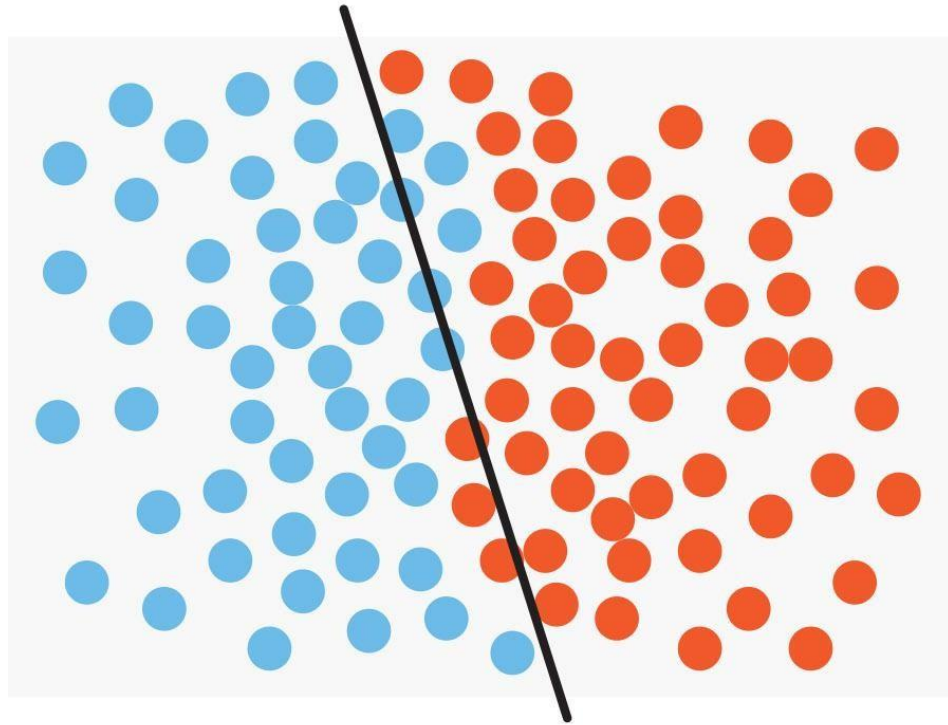
DECISION TREE FOR LOAN APPROVAL



- + **The old math assumes relationships between ingredients (variables) are straight-forward...**



- + **The old math assumes relationships between ingredients (variables) are straight-forward, but the world is actually complex and nonlinear**



+ Imagine we tried to build a model that predicts sex

CAN WE DETERMINE GENDER USING HEIGHT?

+ Men, on average, are taller than women.



+ Height is somewhat but not perfectly predictive of sex

CAN WE DETERMINE SEX USING HEIGHT?

- + Men, on average, are taller than women.
- + But there are tall women and short men, and many people that are the same height.
- + ***Using height alone isn't very accurate.***



+ Including weight adds predictive power, but the model still isn't perfect

CAN WE DETERMINE SEX USING HEIGHT + WEIGHT?

+ Because men are, on average, heavier than women of the same height, accuracy would improve.



+ Using height and weight to predict gender causes kids to be classified as women

CAN WE DETERMINE SEX USING HEIGHT + WEIGHT?

- + Because men are, on average, heavier than women of the same height, accuracy would improve.
- + ***But children would mostly be misclassified as women.***



+ Is birthdate predictive of Sex?

CAN WE DETERMINE SEX USING HEIGHT + WEIGHT + BIRTHDATE?

- + Knowing people's age eliminates the misclassification of children, and improves the model's accuracy of determining gender across all age groups.
- + *But the idea that birthdate can help determine sex is not obvious*



- + **When there are a lot variables, the recipe connecting them to an outcome can be so complex no one can understand it:**



- + **When there are a lot variables, the recipe connecting them to an outcome can be so complex no one can understand it: a black box**



- + The problem with the black box is the risk it will be biased in ways we don't understand.



+ If you're buying a high mileage car in Nevada there's a big probability you're a person of color



1
8

- + **ML algorithms relentlessly refine their “recipe” to achieve the best outcome (aka they adjust to better accomplish their target or objective)**

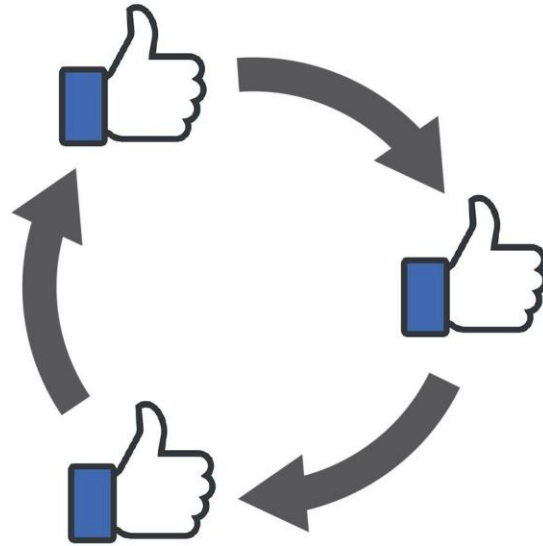


**Every algorithm must
be given a target**

+ Social media algorithms seek to maximize their target: engagement



**Every algorithm must
be given a target**



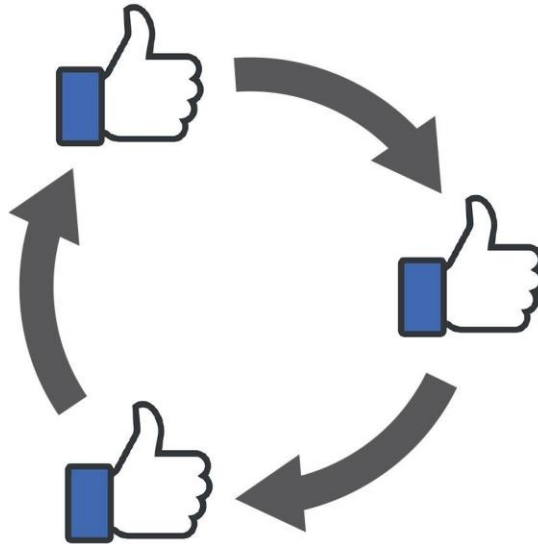
**Social media target:
Maximize engagement**

2
0

- + **The algorithm single mindedly focuses on engagement regardless of whether it's good for your health or good for society**



**Every algorithm must
be given a target**



**Social media target:
Maximize engagement**



**Without regard for
societal harm**

- + **Giving an algorithm one target is problematic: imagine a self-driving car whose only target was to get you from point (a) to point (b)**



**Target: Get from
point (a) to point (b)**

2
2

+ Self-driving cars have a second target: Safety



Target: Get from point (a) to point (b)



Second Target: Safety (obey traffic laws; avoid accidents with cars, pedestrians, cyclists, etc)

+ **We can do this in financial services: Target a low risk of default . . .**



**Target: Low risk of
default**

2
4

+ We can do this in financial services: Target a low risk of default while also targeting fairness

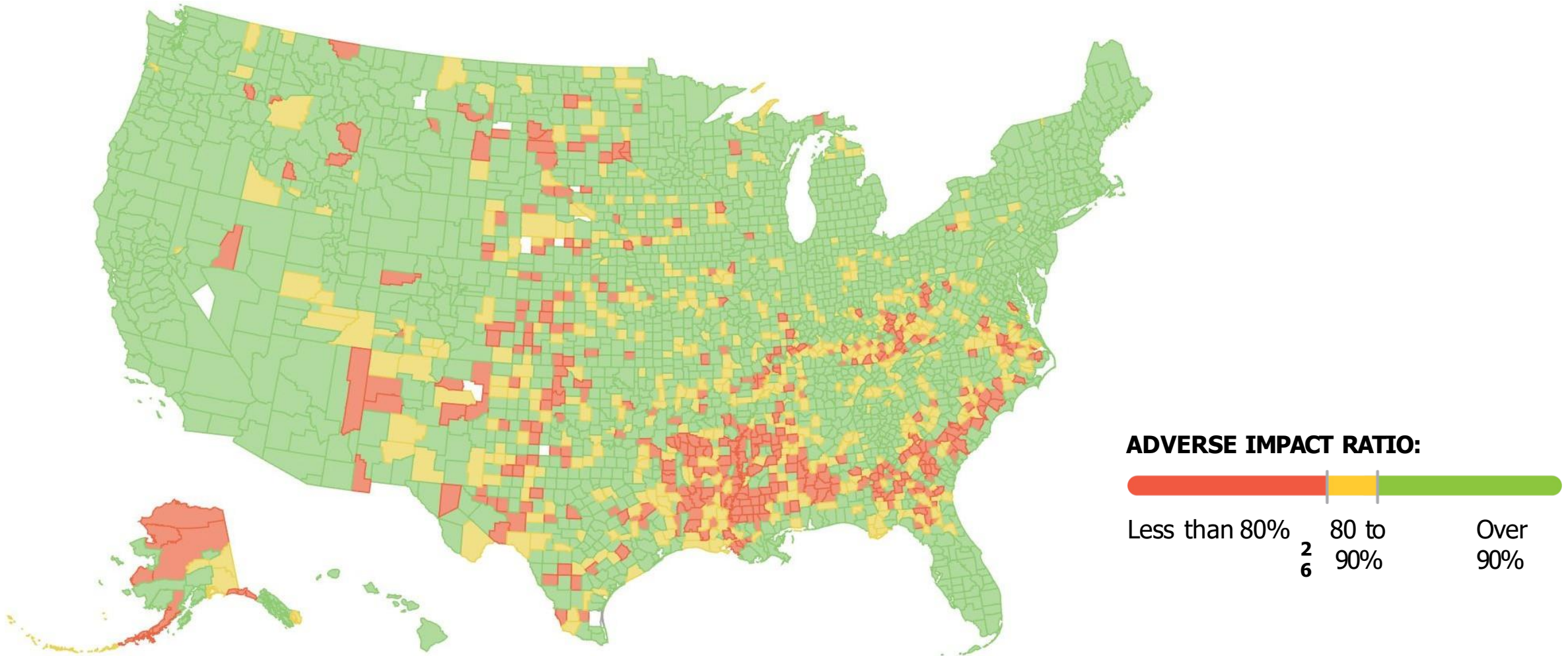


Target: Low risk of default

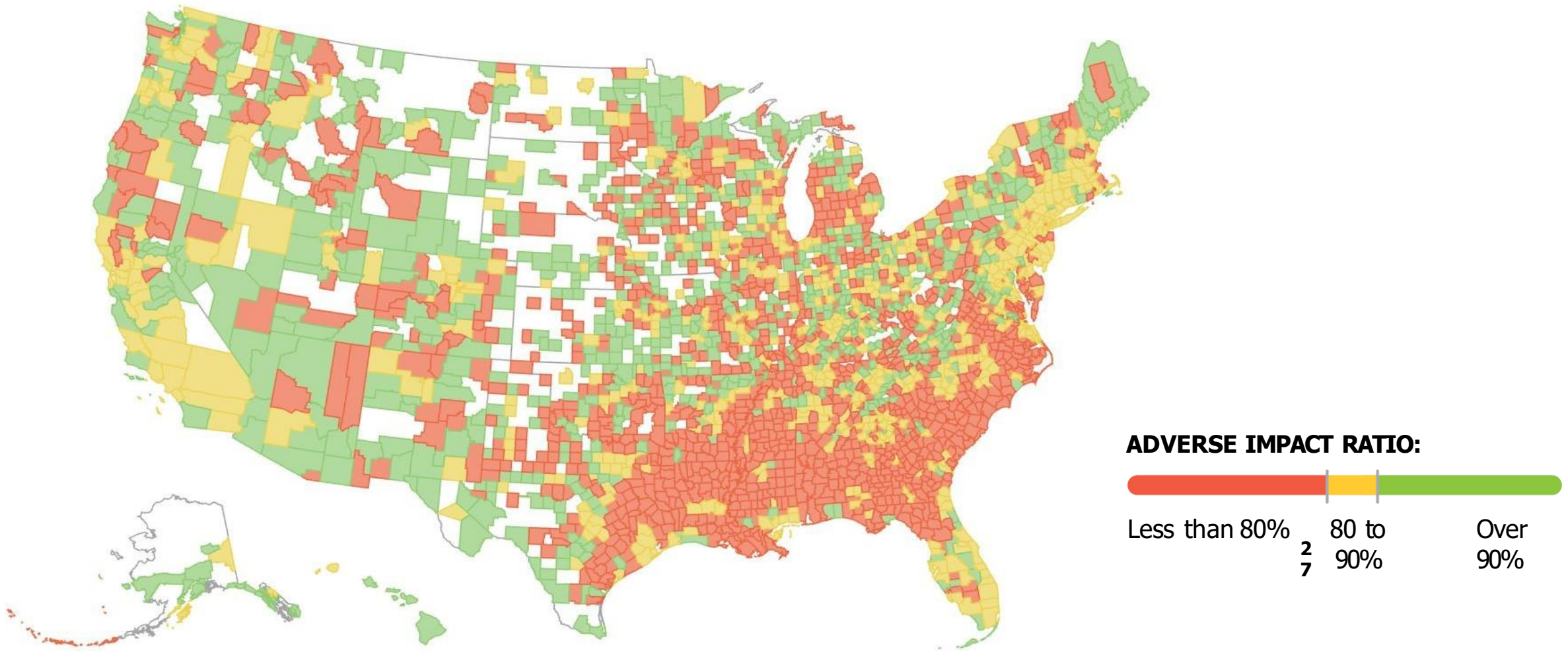


Second Target: Fairness

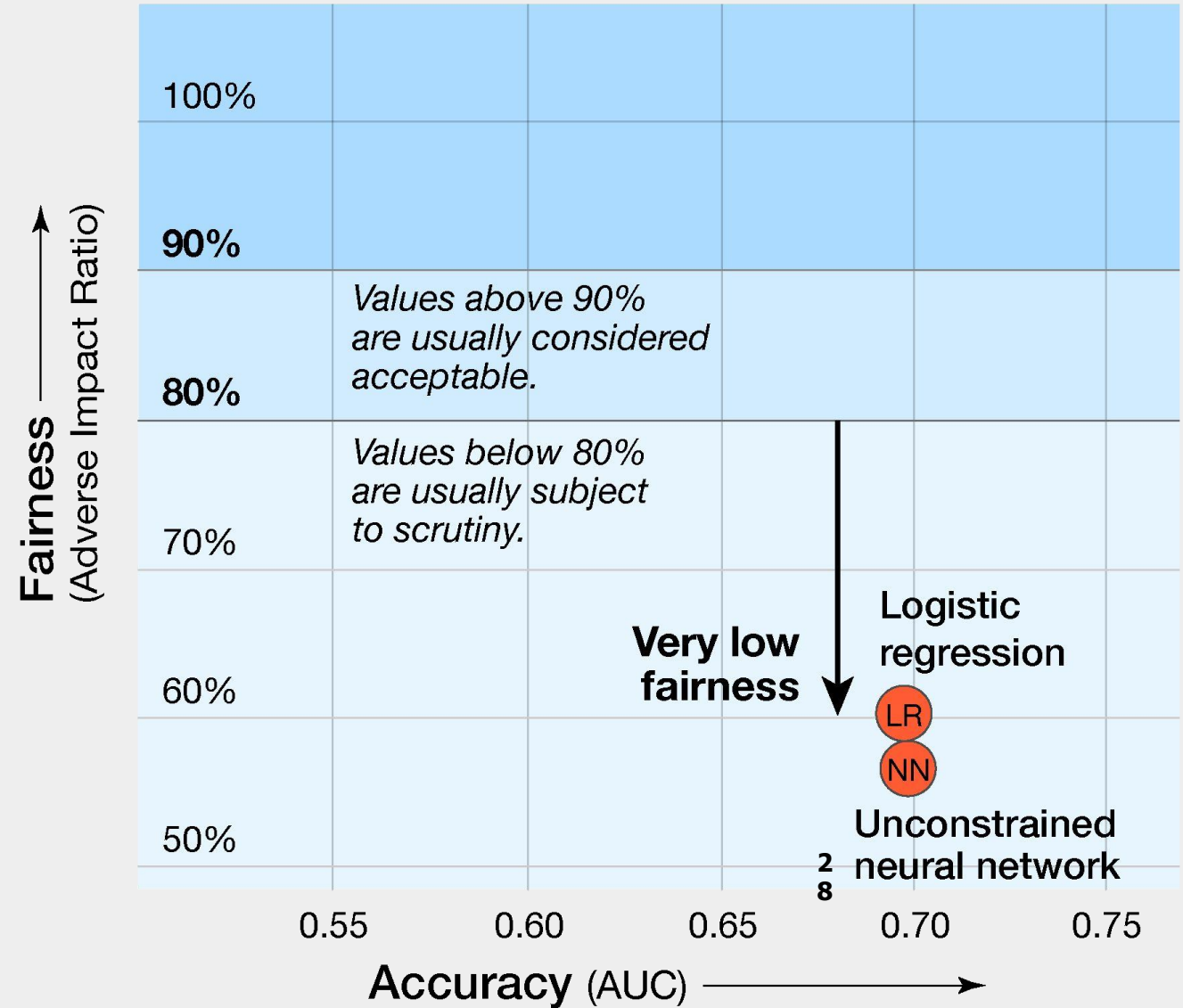
+ U.S. Mortgage Fairness in 2020: Female



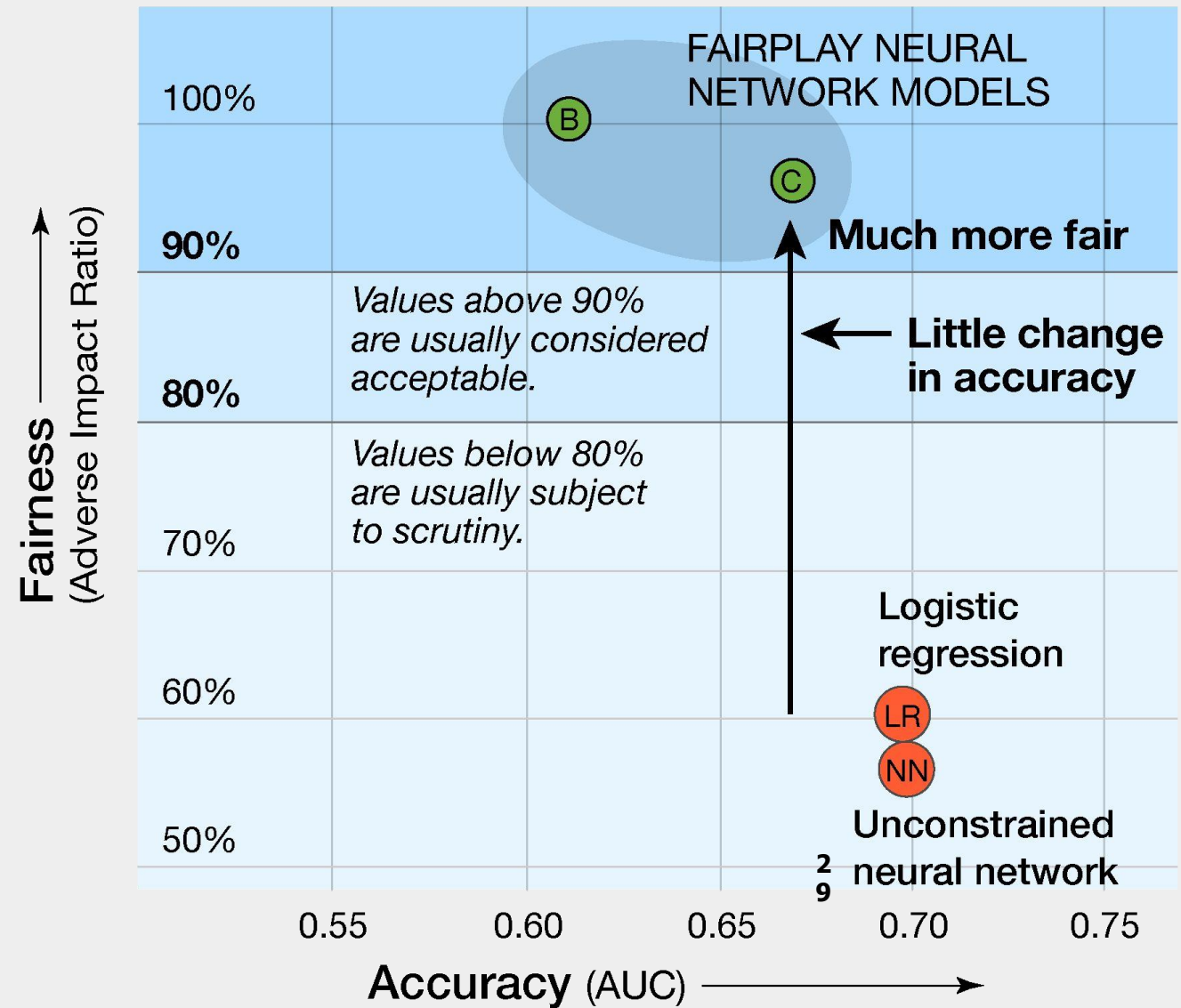
+ U.S. Mortgage Fairness in 2020: Black



+ A fairness-optimized neural network can produce very accurate and considerably fairer mortgage outcomes than a logistic regression or unconstrained neural network

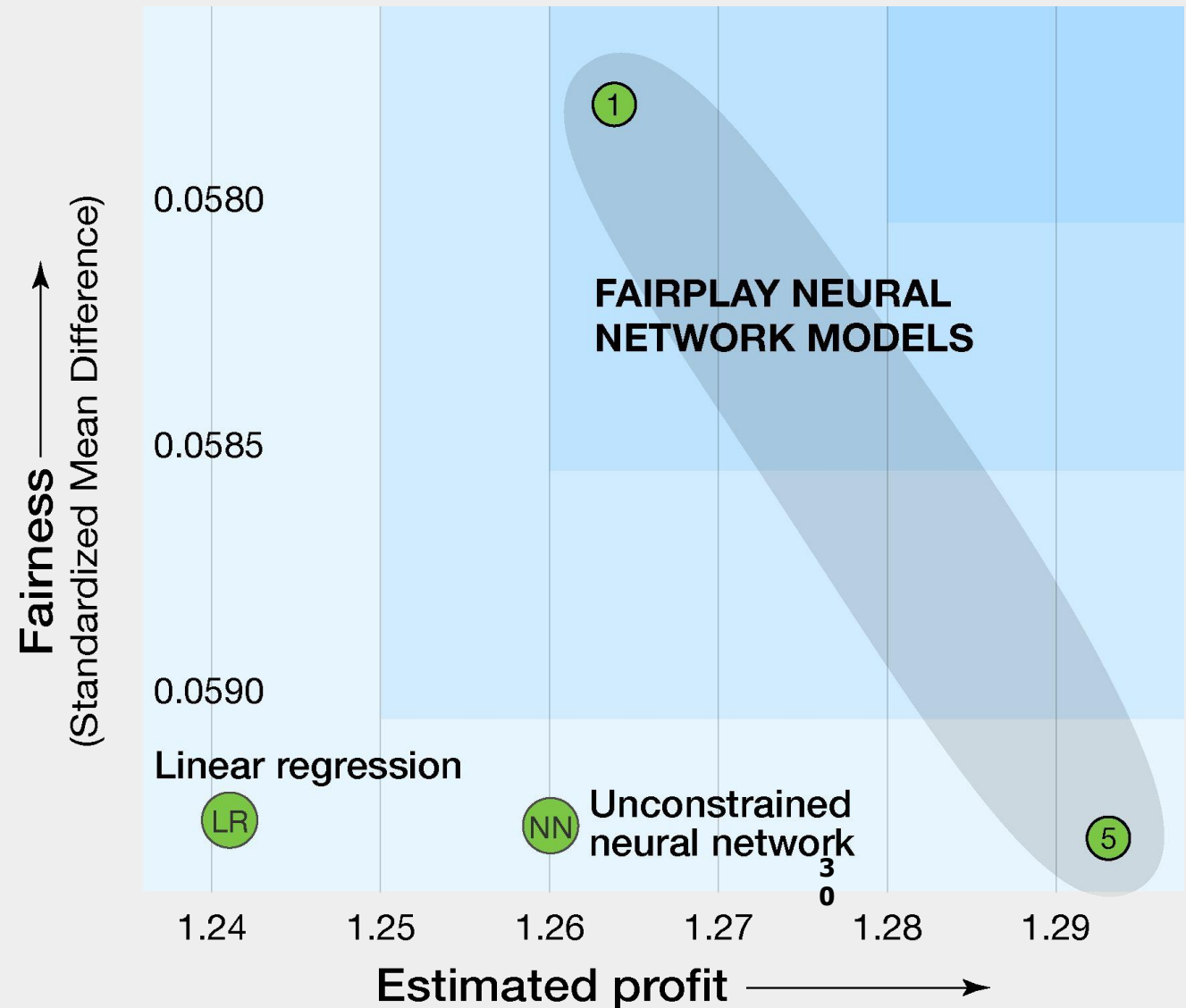


+ A fairness-optimized neural network can produce very accurate and considerably fairer mortgage outcomes than a logistic regression or unconstrained neural network

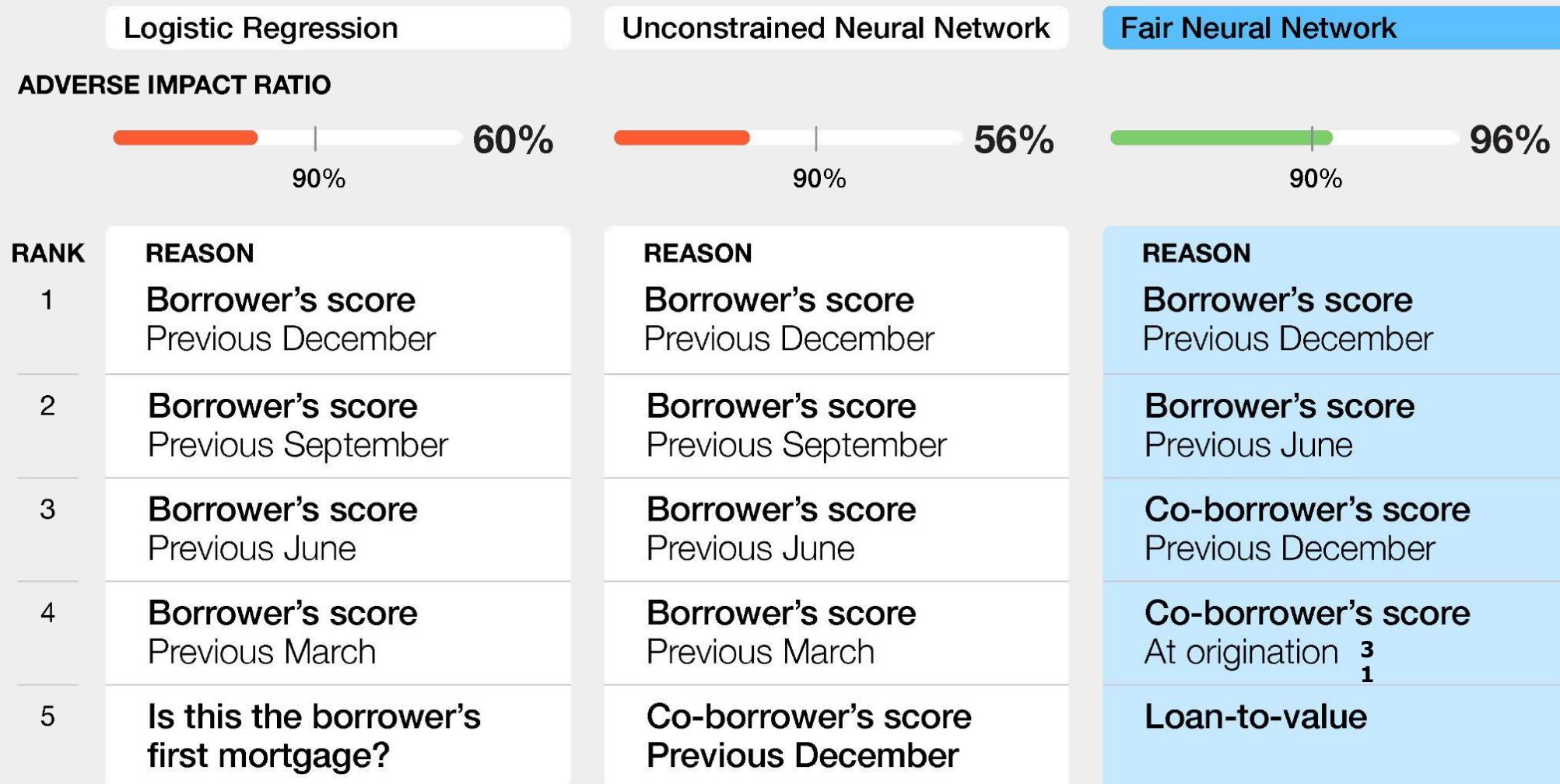


+ A fairness-optimized neural network can produce fairer mortgage pricing than a linear regression or unconstrained neural network

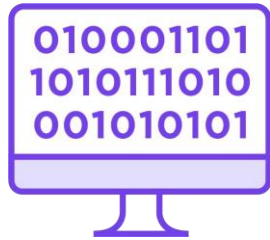
SMD values less than 0.2 are generally considered acceptable; all of the models shown here are well within that range.



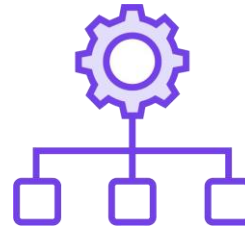
The fairness constrained neural network generates fairer results using the same variables



+ Catalysts driving AI's takeover of high stakes decisions



Data



Advanced Algorithms



Cloud Computing



CONTACT

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BROOKINGS

QUALITY. INDEPENDENCE. IMPACT.

Algorithmic auditing – why and how

Alex C. Engler

Twitter: @aengler

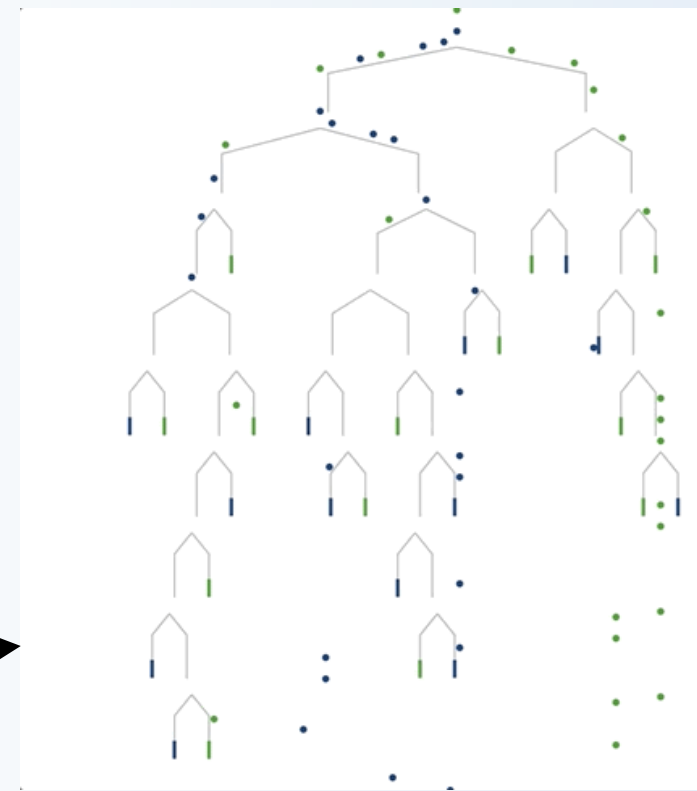
Email: aengler@brookings.edu

Proliferation of algorithms in important services

- “Over 75% of colleges and universities use analytics for enrollment management” – [EduCause Survey, 2015](#)
- “55% of human resources leaders in the United States use predictive algorithms in hiring, at many stages” – [Mercer Global Talent Trends](#)
- “98 percent of mortgages originated by Quicken Loans, the country’s largest lender, used the company’s digital platform, Rocket Mortgage” writes [the NYTimes](#)
- “We found that a category of algorithms that influences health care decisions for over a hundred million Americans shows significant racial bias,” said [Sendhil Mullainathan](#)

Characteristics of algorithms in important services

- Scaled – they often affect many people
- Automated – they often replace an existing process that was done by a person (but not always)
- “Black box” – They may obscure the decision-making process & reduce efficacy of civil liability
- Might be consistent in treatment (but often not in impact) →
- “Learning” means algorithmic behavior may change over time



Concerns with algorithms in important services

- Power centralizing towards institutions (e.g. tuition, ride-sharing)
- Incentivizes privacy violations (e.g. Clearview AI, EverAlbum)
- Opaque process may offer no feedback (e.g. mortgage denial)
- Fraud (different than efficacy)
- Discrimination
- Systemic Risks (e.g. flash crashes)

Proposed solution: Algorithmic audits!

- An algorithmic audit is a review of the environment, inputs, functions, outputs, and results of an algorithmic system
- This is what the **NFHA's Purpose, Process, Monitoring Framework** describes, but there are many other examples from health, hiring, education, social media, etc.
- Can be paired with a fine for illegal activity, possibly algorithmic deletion in extreme cases

Do algorithmic audits help with these problems?

- Power centralizing towards institutions (e.g. tuition, ride-sharing)
 - Mostly no, may create public awareness
- Incentivizes privacy violations (e.g. Clearview AI, EverAlbum)
 - Yes, if the privacy violations are illegal
- Opaque process may offer no feedback (e.g. mortgage denial)
 - Mostly no – audits offer global, not individual, explanations.
- Fraud
 - Yes!
- Discrimination
 - Yes!
- Systemic Risks
 - Maybe

What should an audit be?

Process considerations:

- Independent
- Adversarial
- Representative (if there are many models)
- Access to data/code/models

Dependencies & Docs

- Data collection process
- Training data representativeness
- Dependency analysis
- Accurate documentation

Model & outcome review

- Look at outcome/training/hypothetical data
- Consider problem definition
- Consider algorithmic drift, frequency of audit

Who needs to do audits?

- Criteria
 - Audits must be **independent** and **adversarial**.
 - They require extensive access to the data and algorithms
 - This will often mean government action
- Audits do not automatically mean accountability. In accounting, it takes:
 - Market incentives (these don't apply to algorithmic discrimination)
 - Financial auditors (the people!) are legally liable if they commit fraud
 - Plus direct government oversight (SEC)



01-26-21 | POV

Independent auditors are struggling to hold AI companies accountable

Controversial AI company HireVue implied that an external audit showed its algorithms had no bias. But a look at the audit itself tells a different story.

So, Government Audits

- Why:
 - They can be genuinely independent.
 - Usefully, fraud & discrimination are often already against the law.
- How:
 - Develop specific best practices/guidance by model/application type
 - Some legal changes may be necessary
 - Expand administrative subpoenas to include data sets
 - E.g. remove 'predictive validity' from EEOC Uniform Guidelines
 - Hire regulatory data scientists (soon to be a booming field)
 - Start information gathering, enable consumer complaints, and encourage industry whistleblowing

Government Agency Action

- HHS was only agency to detail its [algorithmic regulatory authority](#)
- EEOC launched [Initiative on AI Fairness](#)
- Federal Trade Commission
 - Threatening [blog post on fraud and discrimination](#)
 - Fines and algorithmic deletions for privacy violations (EverAlbum)
 - [Started rulemaking consideration](#)
- Five financial regulators started [information gathering](#) on algorithms, citing Fair Housing + Equal Credit Opportunity Act
 - (DOT OCC, Fed Reserve, FDIC, CFPB, NCUA)
- Housing and Urban Development
 - Reversing the Trump admin change to the [‘discriminatory effects’ rule](#)
 - Has not responded to the two E.O.s on algorithms (E.O. 13960 and E.O. 13859)

Aside: Agencies have algorithms too

- Federal agencies can show the way by documenting and auditing their own algorithms
- Algorithmic documentation required under E.O. 13960
- Eventually, standards for algorithmic auditing will be the norm

Complications of algorithms in important services

- Many algorithmic stages of “funnel processes;” consider hiring:
 - Job recommendation -> Resume analysis -> Questionnaires/Games -> Automated interviews
- Different types of algorithms combined, consider college tuition:
 - First: predict what one student would pay to attend
 - Second: assign scholarships across many prospective students
- Separate algorithmic systems affecting one another (e.g. online markets)
- Human + algorithmic decisions can have unexpected outcomes
 - Felony sentencing by judge [with algorithmic risk assessment](#)

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PPM AUDITING FRAMEWORK

A COMPREHENSIVE NEW FRAMEWORK FOR
AUDITING ALGORITHMIC SYSTEMS

AI AUDITING IN HOUSING

LISA RICE, PRESIDENT AND CEO OF NFHA
MARCH 22, 2022

Amazon built an AI tool to hire people but had to shut it down because it was discriminating against women



Employment

Health

Racial bias in a medical algorithm favors white patients over sicker black patients

+ Add to list

A photograph showing a Black woman with curly hair sitting up in a hospital bed, wearing a light blue hospital gown. She is looking towards a white male doctor with a beard, who is sitting on the edge of the bed and looking back at her. The setting appears to be a hospital room with a window in the background.

Healthcare

Credit scores in America perpetuate racial injustice. Here's how

Sarah Ludwig

Credit reports and scores reflect existing racial inequities in our credit system and economy

Tue 13 Oct 2015 10:14 EDT

503 200

A close-up photograph of a credit report document. A pair of black-rimmed glasses is resting on the document. The document has sections for "CREDIT INFORMATION" and "FIRST MORTGAGE". There are also some smaller text elements like "See Contact Information to contact creditor" and "Account Information".

Credit Scoring

APPLE POLICY TECH

Apple's credit card is being investigated for discriminating against women

Customers say the card offers less credit to women than men

By James Vincent | Nov 11, 2019, 5:57am EST

f t SHARE



Image: Apple

Credit

The Markup

Big Tech Is Watching You. We're Watching Big Tech.

Locked Out

Can Algorithms Violate Fair Housing Laws?

Landlords increasingly use screening services to weed out renters. Advocates say both landlords and the algorithms should be accountable when things go wrong

By Lauren Kirchner

September 24, 2020 08:00 ET



Buša Photography/Getty Images

Tenant Screening Services

After Lawsuits, Facebook Announces Changes To Alleged Discriminatory Ad Targeting

npr By Brakhton Booker

Published March 19, 2019 at 1:32 PM CDT

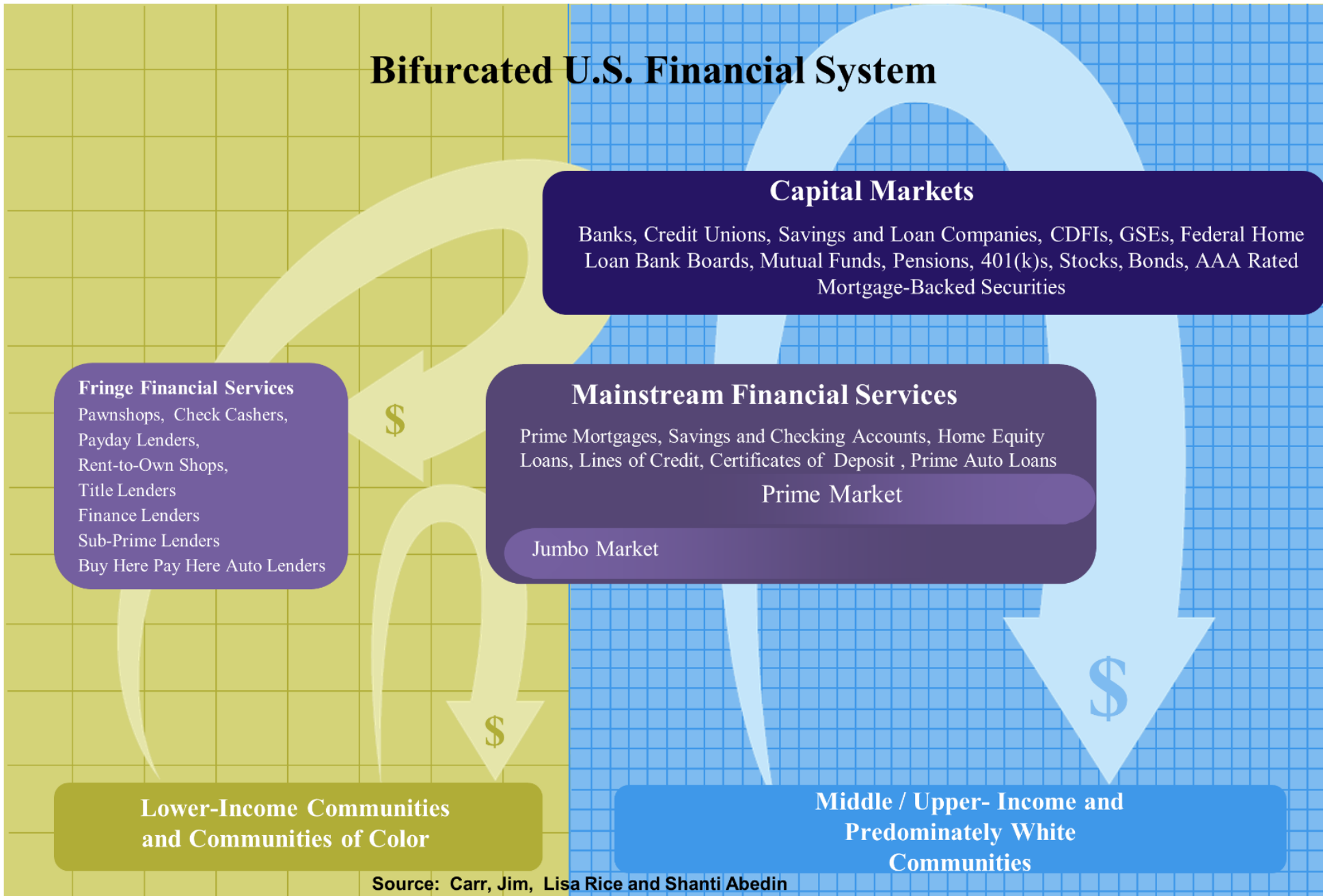
f t in



Marketing Platforms

- Credit Scoring Systems
- Automated Underwriting Systems
 - Automized Verifications Processing
- Risk-based Pricing Systems
- Advertising Systems/Platforms
- Revenue Management Software
- E-loans Processing
 - Video-Conferencing
 - Digital Fingerprinting
- Facial Recognition Systems
- Automated Property Valuation Systems
- Tenant Selection Systems

Bifurcated U.S. Financial System



HOW TECHNOLOGIES MANIFEST BIAS

- Unrepresentative, Insufficient or Flawed Data
- Design Flaws
- Biased Feedback Loops
- Insufficient or No Testing for Bias
- Untrained Designers
- Lack of Diversity

EFFECTIVE AUDITING OF SYSTEMS CAN REDUCE DISPARITIES

- Compelling models to be fairer can lead to the development of more accurate and safer models that minimize or eliminate disparities in housing and lending.
- The adoption of fairer models would expand access to credit for underserved groups and result in less disruption for consumers, families, communities, and industry.
- Fairer models can help reduce the racial wealth and homeownership gaps.
- The development of state-of-the-art policies and protocols would mark the U.S. as a leader in advancing responsible technologies that serve markets, consumers, and the greater society.





OVERVIEW AND PURPOSE ELEMENT OF THE PPM FRAMEWORK

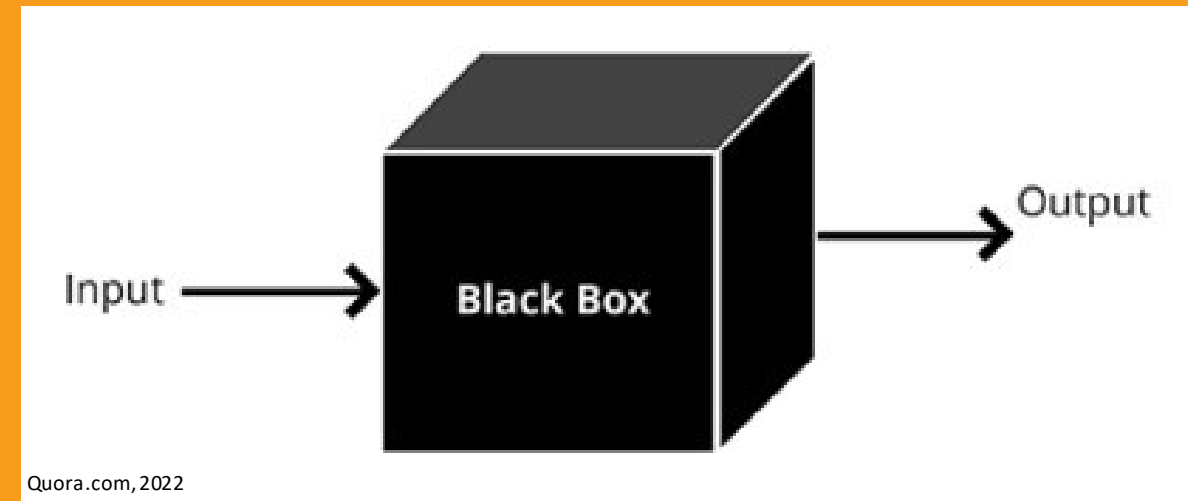
SNIGDHA SHARMA, FAIROPS TEAM LEAD & TECH EQUITY ANALYST, NFHA

OUTLINE

- What job does PPM – Purpose, Process, Monitoring - do?
- The Purpose element of the PPM Framework?

WHAT JOB DOES PPM TRY TO DO?

- Examines every stage of algorithm solution development
- A trust tool regulators can use to gather information that could be used to decide how fair, transparent and explainable an algorithmic system is.
- Helps policymakers see what aspects of algorithmic systems need mitigation and can lead to algorithmic risks to consumers
- Helps policymakers hold those developing algorithms accountable for any decisions they made while building the system



THE PURPOSE IN PPM: AN APPLICATION OF ALGORITHM IN RENTING

- Business Problem: Would this person meet their rent obligation or not?
- Input: application information, employment, bank statements etc. of previous or current tenants
- Black box: a classification algorithm like Random Forest
- Output: tenant score that rates an applicant and is used to decide if an individual is given a lease or not



THE PURPOSE IN PPM: WHAT QUESTIONS DOES IT ANSWER?

Business Understanding:

- The Purpose section asks why the model is being built and what are the business objectives.
- Auditors are guided to seek information to make informed decisions about risks the business problem may pose to consumers, institutions, and society at large.
- Does the system pose too great a harm that it should not be built or put into production?
- What exactly is the business problem being solved?



Francesco Pochetti
@Fra_Pochetti

All ML projects which turned into a disaster in my career have a single common point:

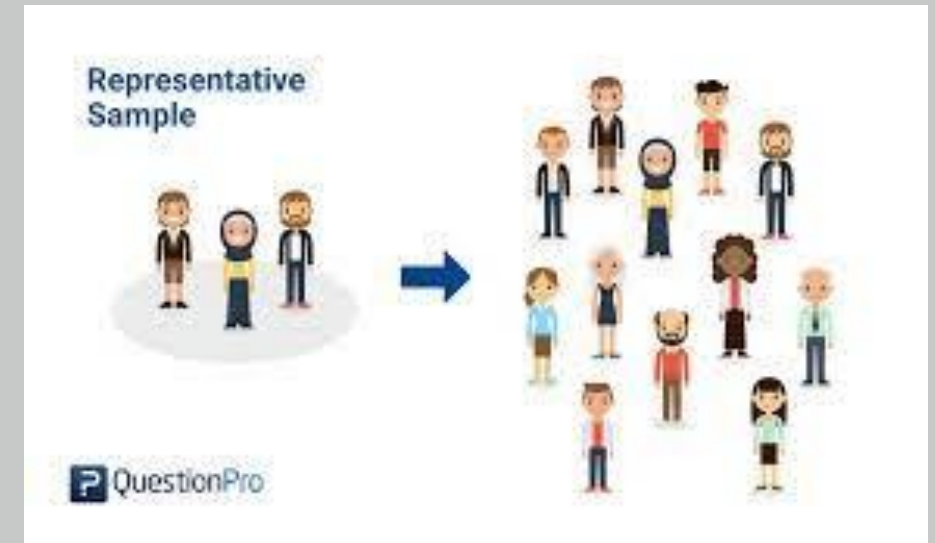


I didn't understand the business context first, got over-excited about the tech, and jumped into coding too early.

THE PURPOSE IN PPM: WHAT QUESTIONS DOES IT ANSWER?

Data Understanding :

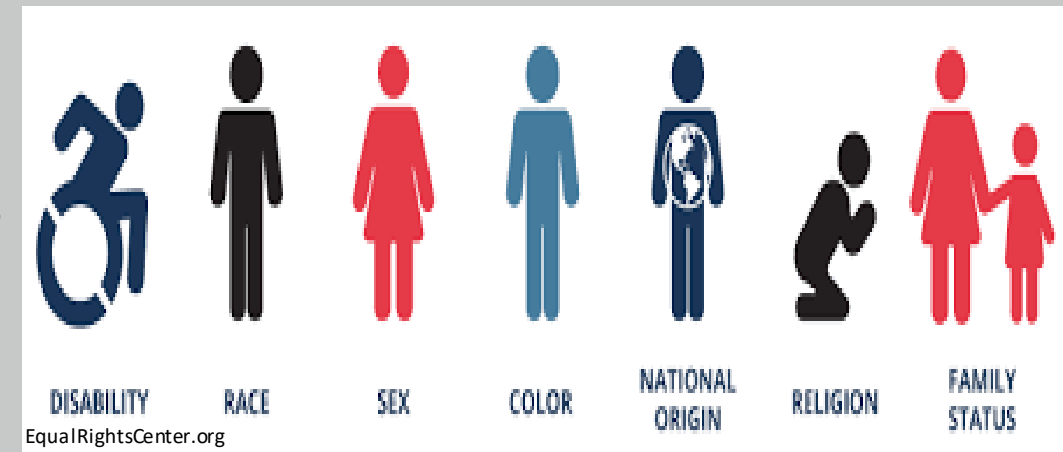
- Is there data to represent the business problem?
- Were any techniques used to mitigate risks associated with data paucity or data quality.



THE PURPOSE IN PPM: WHAT QUESTIONS DOES IT ANSWER?

Data Understanding :

- Is there proxy testing for protected classes?
- What metrics are being used to test model accuracy and fairness?
- Are model features or variables representative across protected class data?



NFHA PPM Auditing Framework: Process

John Merrill, Chief Technology Officer of FairPlay

March 22, 2022



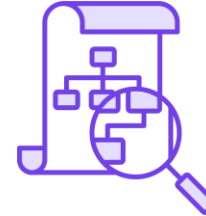
+ The Process Stage of the PPM Framework Includes Five Elements



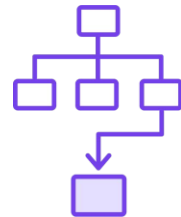
STAFF PROFILE



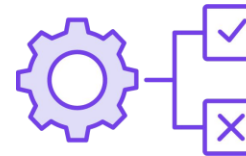
DATA ASSESSMENT



MODEL ASSESSMENT



**OUTCOME
ASSESSMENT**



**MODEL USE AND
LIMITATIONS**



+ Staff Profile



+ **Diverse Team**

Do the members represent a broad class of possible stakeholders?

+ **Functional Team**

Can the members work together?

+ **Communication**

Can the team explain its results to others outside of the team?

+ Data Assessment



+ **Collection**

How was data gathered?

+ **Appropriate**

Is the data well-suited to the model's objective?

+ **Representative**

Does the data reflect the characteristics and qualities of the subject being modeled?

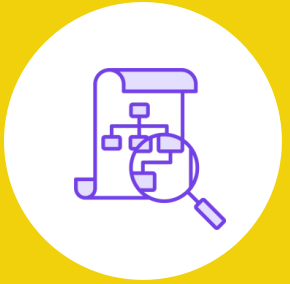
+ **Accurate**

Are the values in the data correct and consistent?

+ **Well-behaved**

Are the data targets and values well-behaved? Are any computed features reasonable?

+ Model Assessment



+ **Model Type**

Is the model type appropriate to the data and problem?

+ **Parameters**

Which variables are included in the model?
How is each one treated?

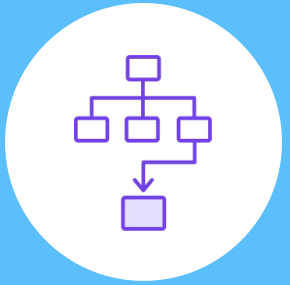
+ **Hyper-Parameters**

How did you set the parameters? For example, did you control how fast the algorithm learned? When did you cut off its learning?

+ **Fairness Constraints & Less Discriminatory Alternatives**

What steps did you take to ensure model was fair? If a model exhibits disparities for a group, did you search for fairer alternatives? If so, how?

+ Outcomes Assessment: Did the model meet its objectives?



+ **Accurate**

Are the model's predictions correct?

+ **Fair**

Do the model outcomes favor one group over another?

+ **Explainable**

Are the model's decisions known and sensible?

+ **Stable & Robust**

Do the model's predictions vary widely when things change?

+ Model Use and Limitations



+ **Limitations and Assumptions**

What does the model assume about the data?

+ **Other Applications**

To what other problems can the model outputs be or not be applied?

+ **Disaster Response**

Are there procedures and processes in place to handle disasters?



CONTACT

john@fairplay.ai



The Monitoring Element of the PPM Framework

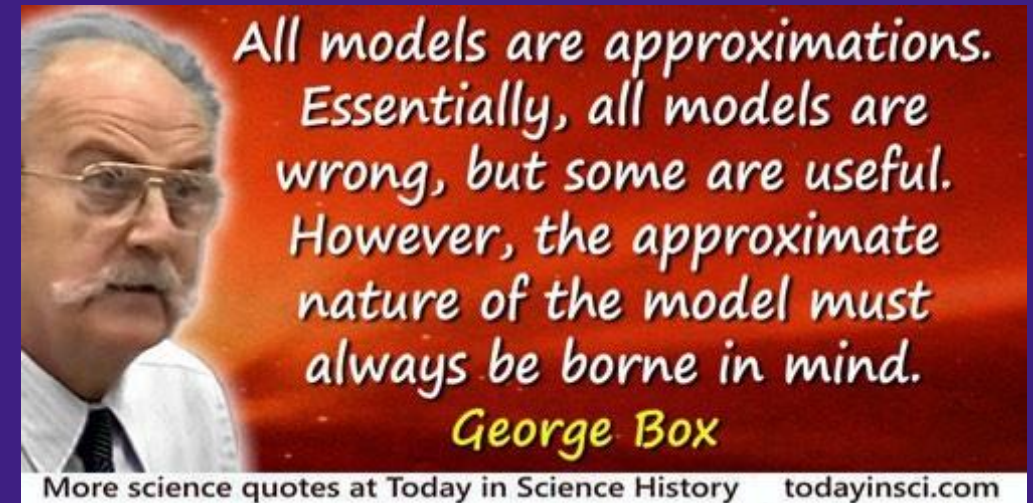
Michael Akinwumi, Chief Tech Equity Officer, NFHA

Outline

- Model Use and Model Limitations
- An algorithmic solution needs monitoring
- Q & A (All)

Model Use and Limitations

- Developer needs to be transparent about blind spots in their model. Every model has a limitation.
- What business problem can the model be applied to?



An algorithmic solution needs monitoring

- “All models are wrong, but some are useful.”
- Document what metrics developers plan to use for monitoring
 - The usefulness of the model and
 - Potential harms it may cause to consumers while in production
- Integrity of the model should be protected, and consumer privacy should be protected in the event of a hack





Question & Answer