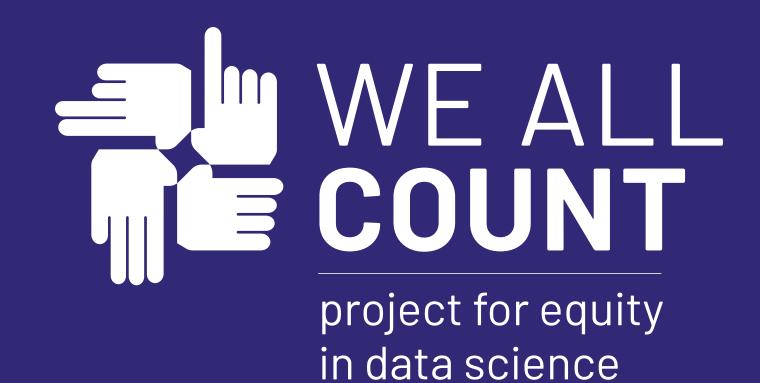
Foundations of Data Equity



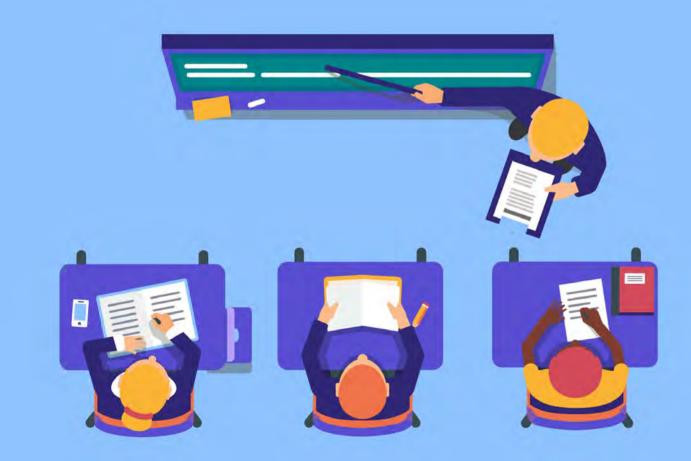
The worst equity problem facing data science is that people are making predjudiced choices but they don't know it.

We want a "silver bullet" against bias, prejudice, human error, and injustice.

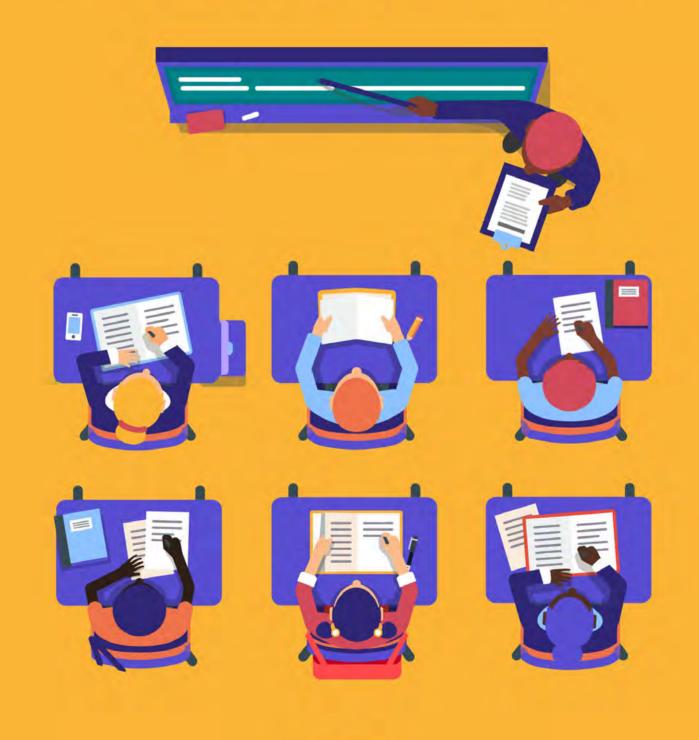


What is the average number of students across these three classrooms?

CLASSROOM A

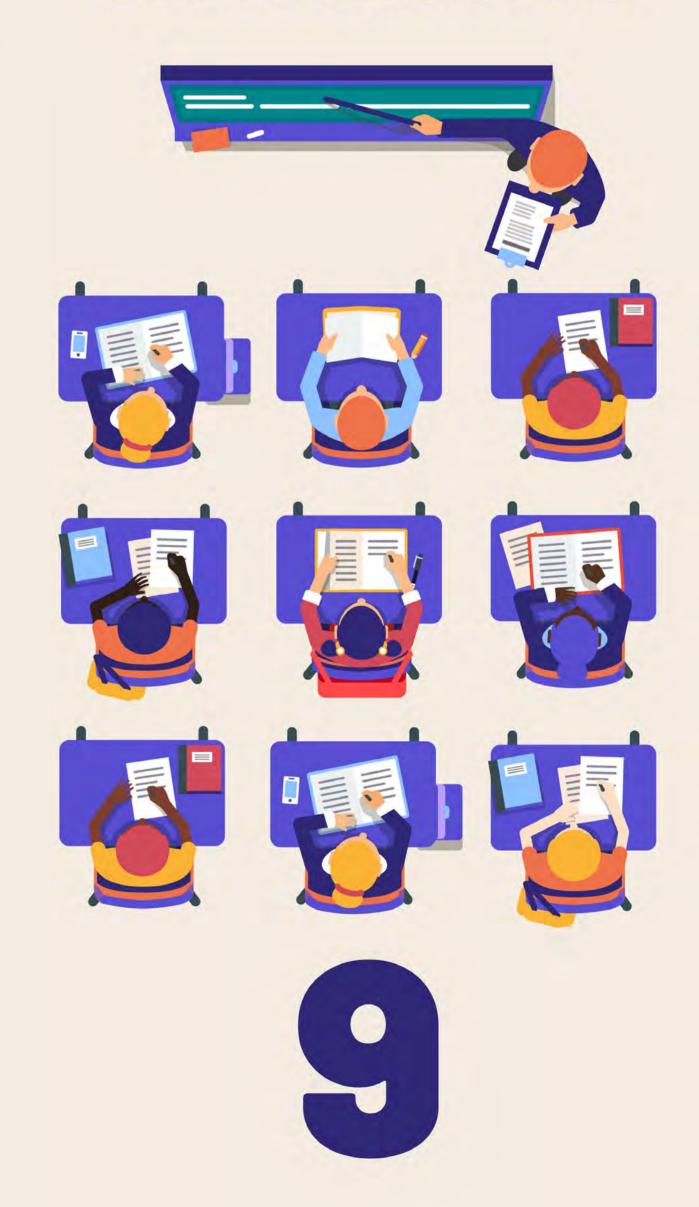


CLASSROOM B



6

CLASSROOM C



If you got 6, you're right.

If you got 7, you're right.

You just did the same math, but from different perspectives.

From the teachers' perspective:

$$3+6+9=18$$
 $18 \div 3 = 6$

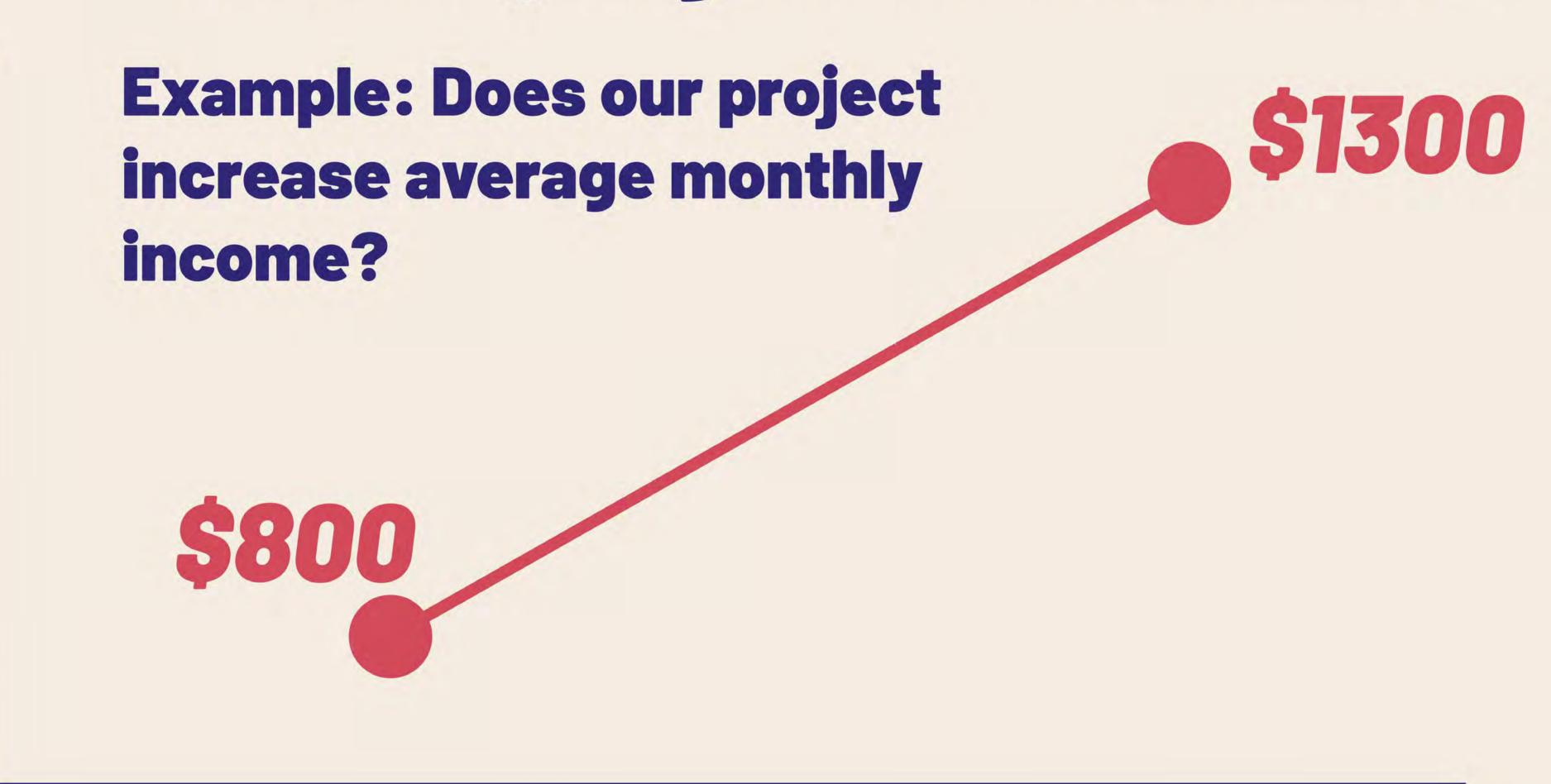
From the students' perspective:

$$3+3+3+6+6+6+6+6+6+9+$$

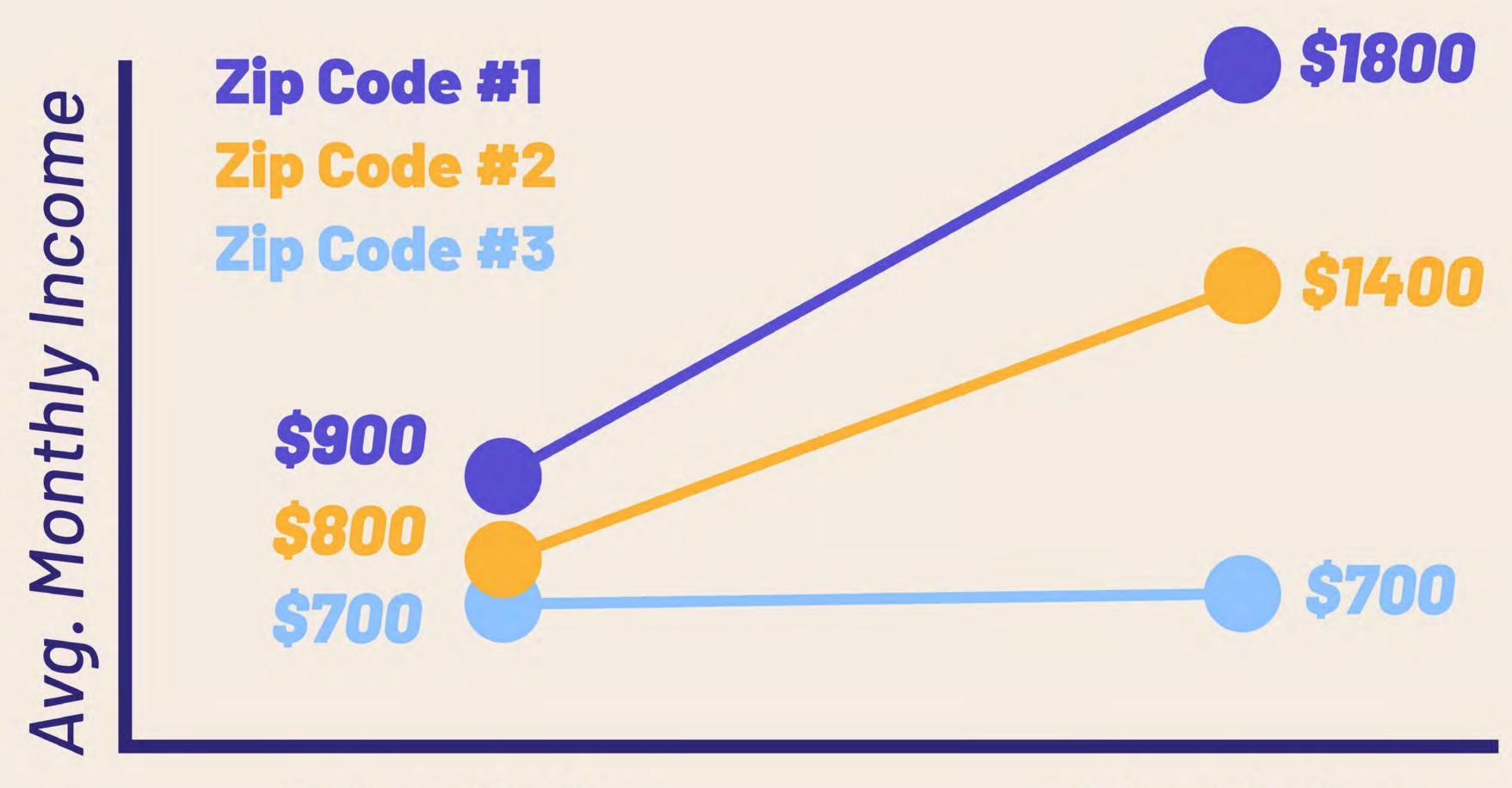
 $9+9+9+9+9+9=126$

RARARA

ls our project a success?



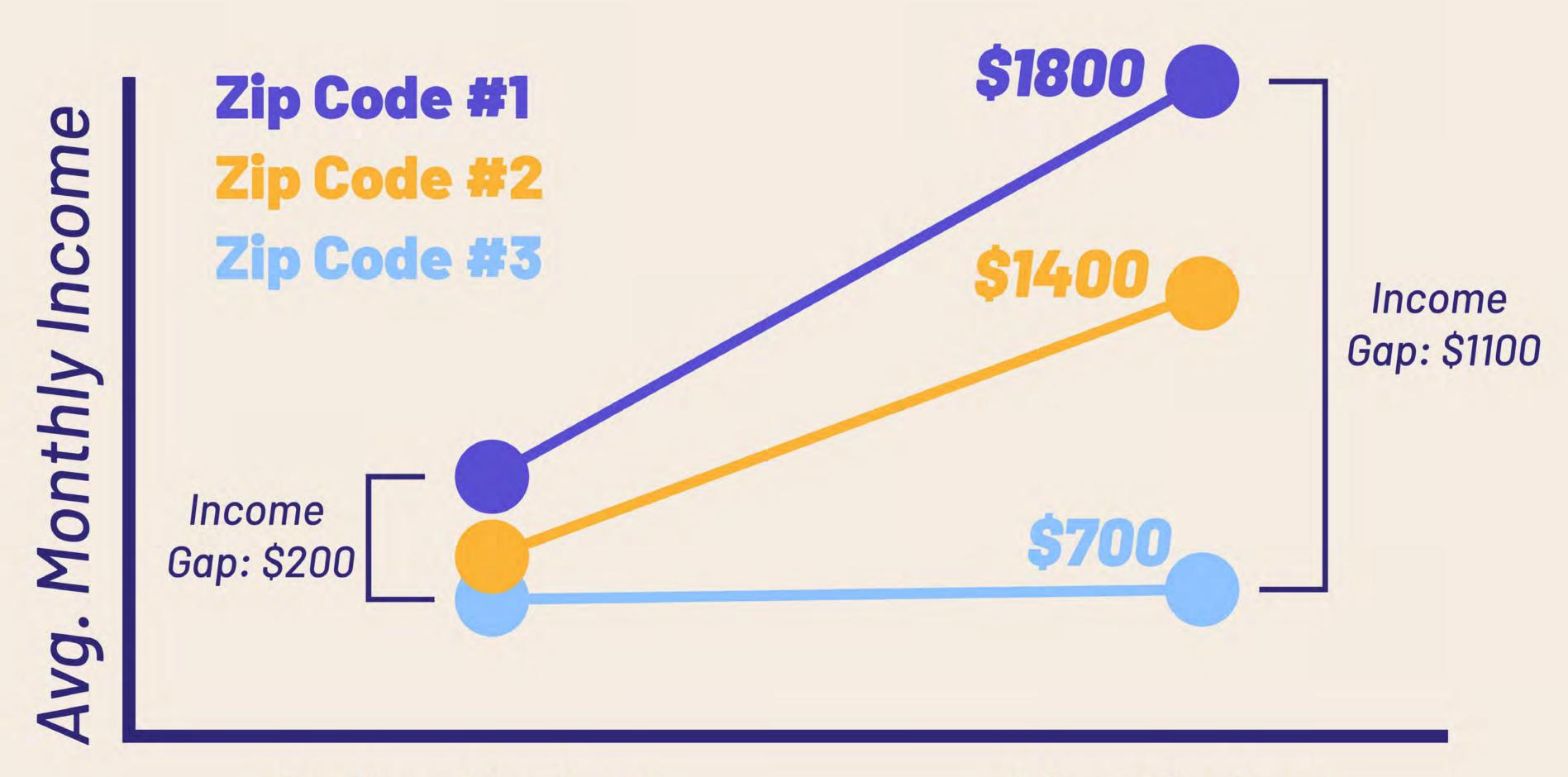
ls our project a success?



Before Project

After Project

Is our project a success?



Before Project

After Project

This isn't just a math problem. It's a human problem.



\$700



We want to think of data science like this:

CHOOSE: What question should we ask?



Objective Data



Objective Results



Objective Analysis



Objective Decision







There is no 'right' choice.

Equity is a process not a binary state.



"We were aware that we were asking them to give up some work time to come to the trainings and we didn't want them to be penalized for that, so we controlled for it in the model.

We were patting ourselves on the back for being so equity-minded."

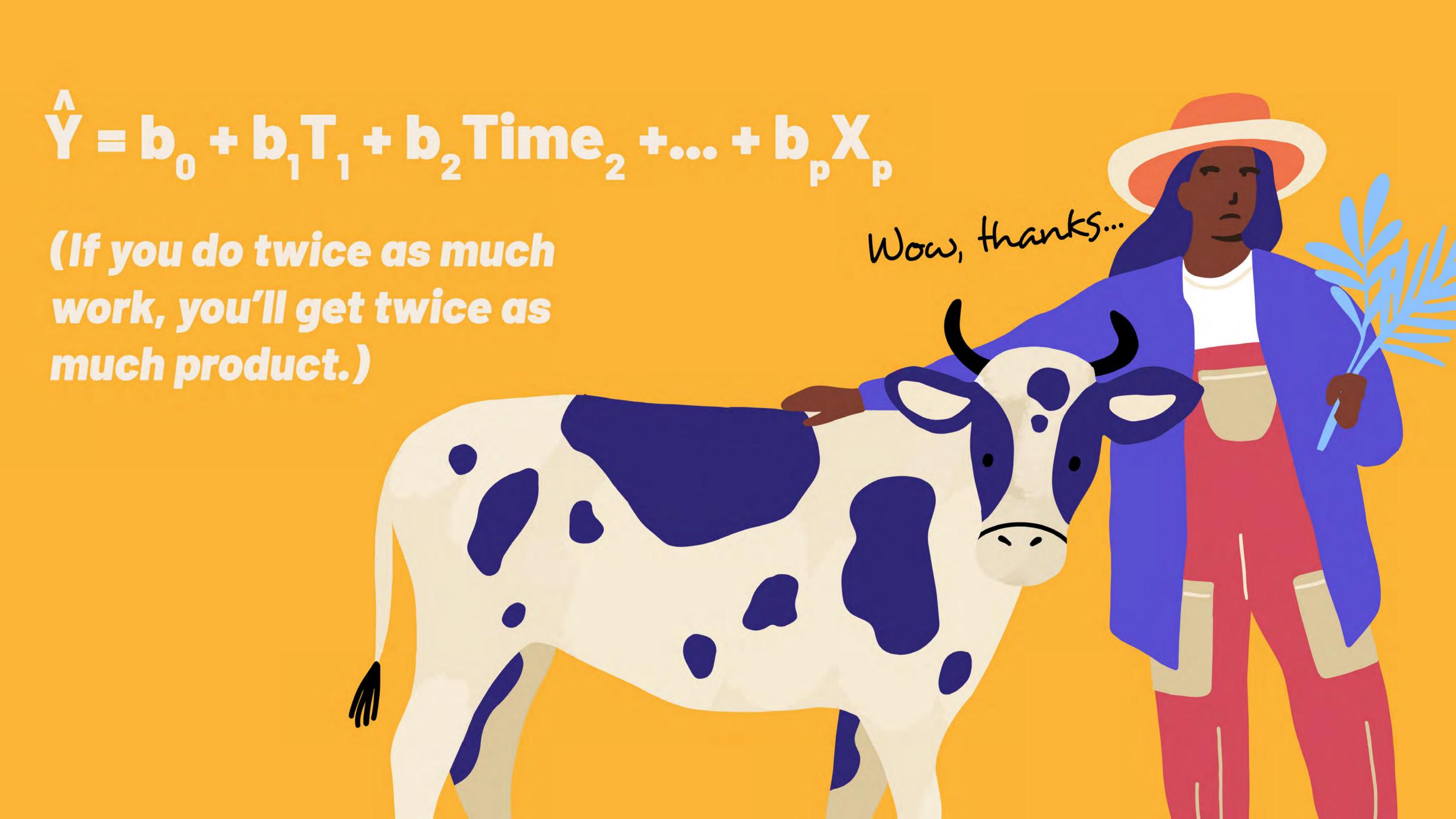


"I can't wait to get out of this program..."

"You think this was a success?!"

"I've barely slept in a month!"

Uh oh...



The first model wasn't "wrong".

But the new model was a better choice because it reflected the equity that we wanted it to.

We got two totally different answers even though we were using:

The same data.

The same research question.

The same methodology.

The same analysts.

Even controlling for the same variables.

Every choice reflects a worldview.

Even the smallest choices can have huge impacts.



AMERICAN INDIAN

Their results are not statistically significant.

Our results are not examinatively significant.

Our results contain uncertainty.

Our results contain uncertainty because...



Choosing Data Equity...

The Data Equity Framework











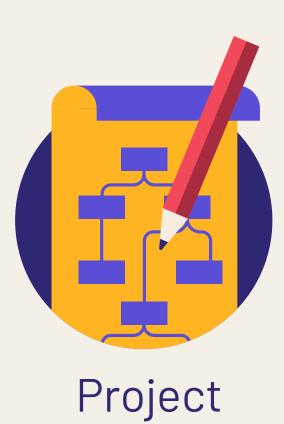




The Data Equity Framework addresses equity issues systematically in each step of a data project. Some form of these steps is universal to all types of data projects.







Design

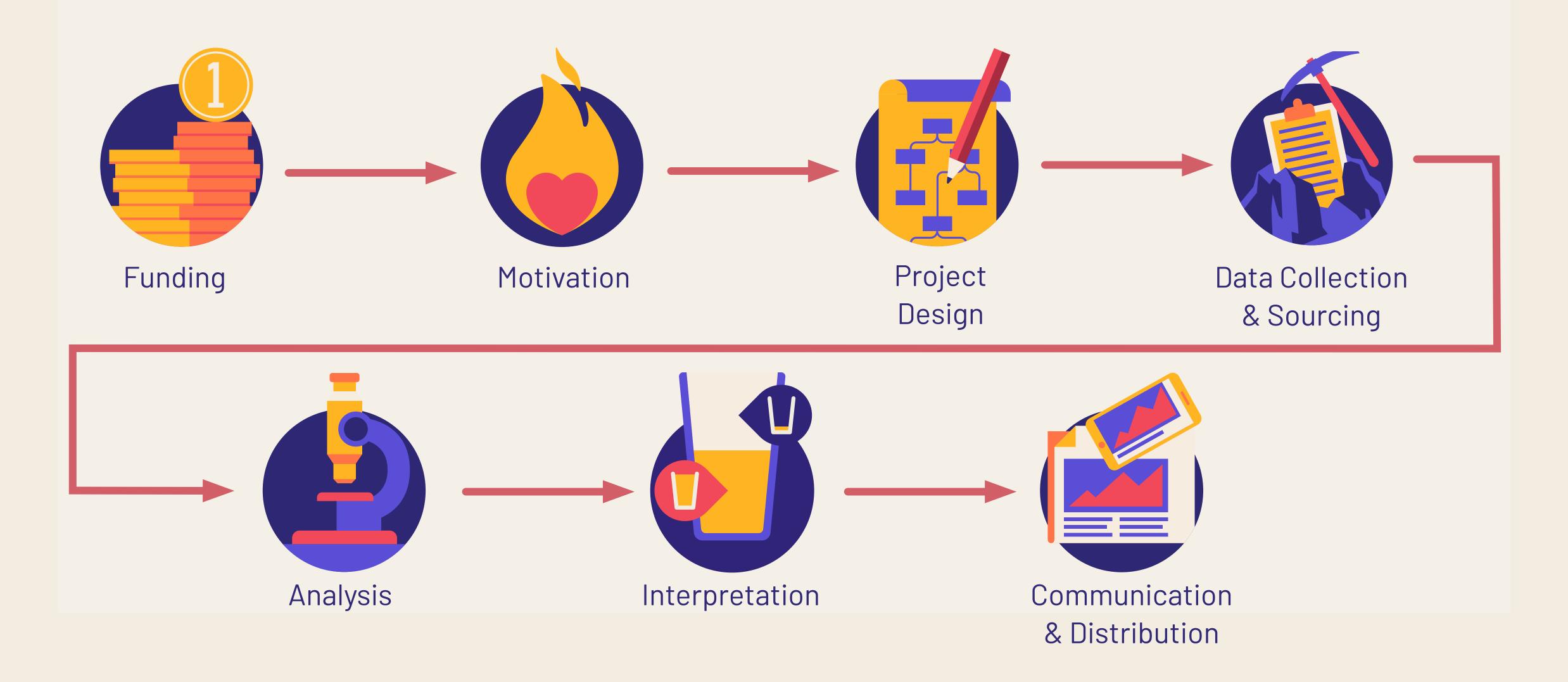








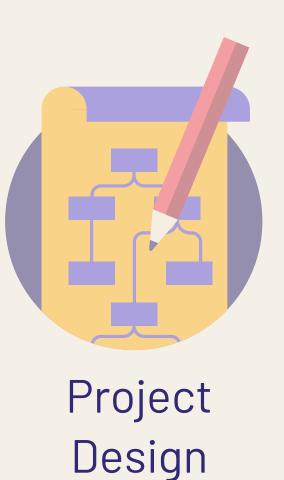
The order of steps reflects the typical data project process, but you can address them in any order.



Even if you can only address some of the steps in your project, it is still worth doing.













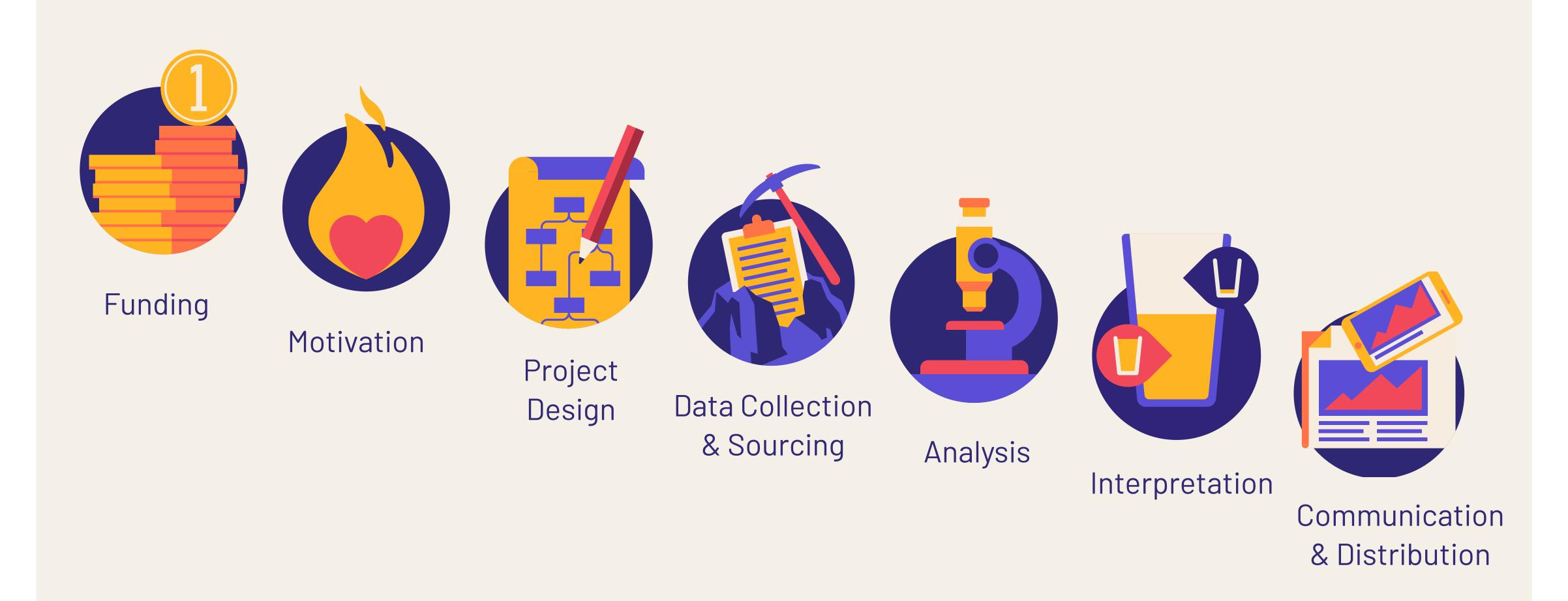


What you do in each step of the Data Equity Framework will have equity impacts and interactions in other steps and your project as a whole.

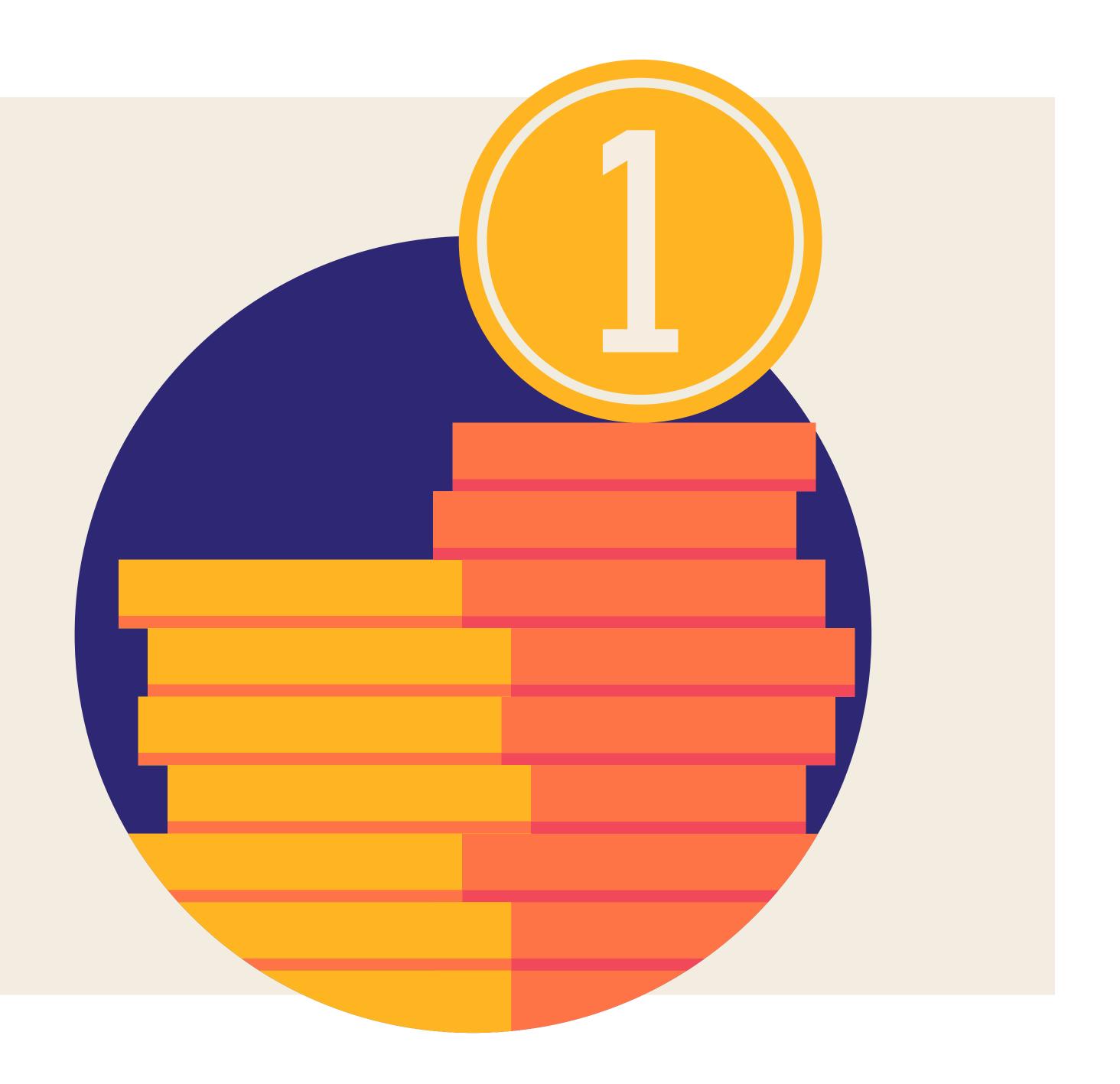
The Data Equity Framework works holistically and individually.



Our plan for this workshop:



Funding



Motivation

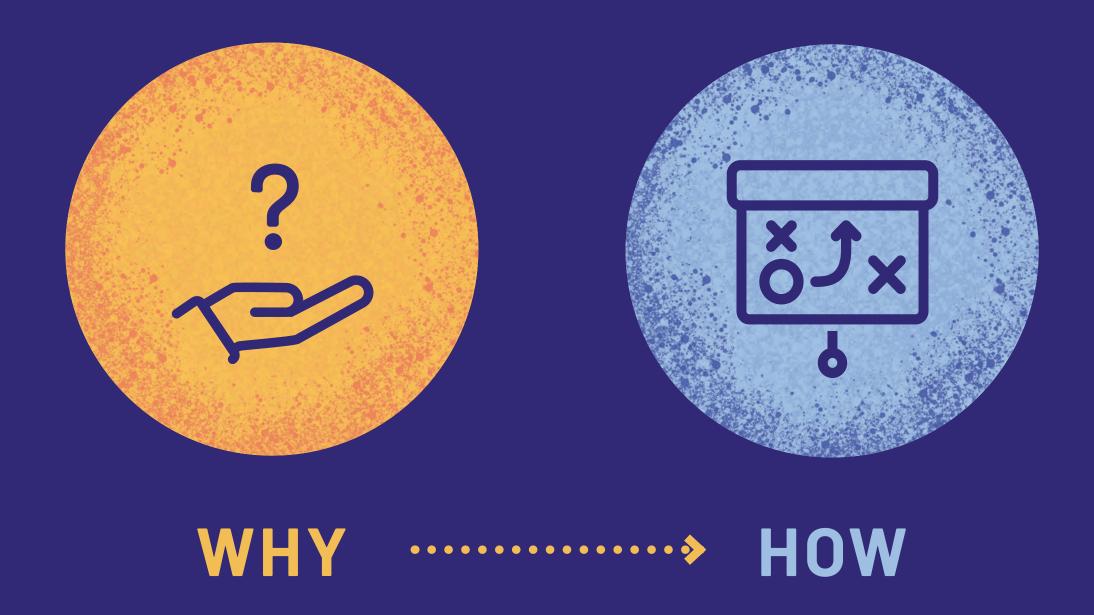




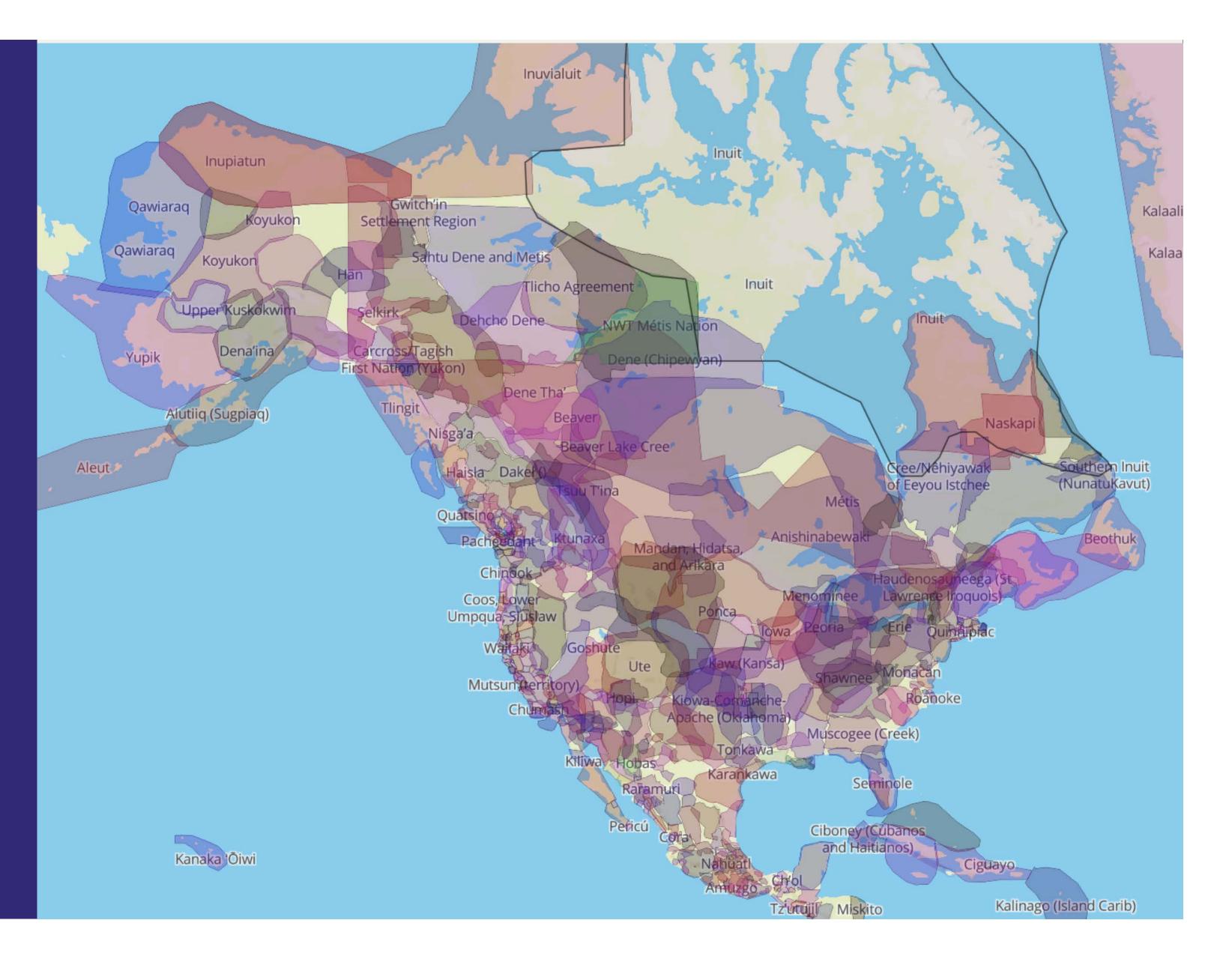
Step 3: Project Design

Project Design is the phase where the WHY becomes the HOW.

Critical step in data equity.



'Geographical Sample Area' based on whose geography?

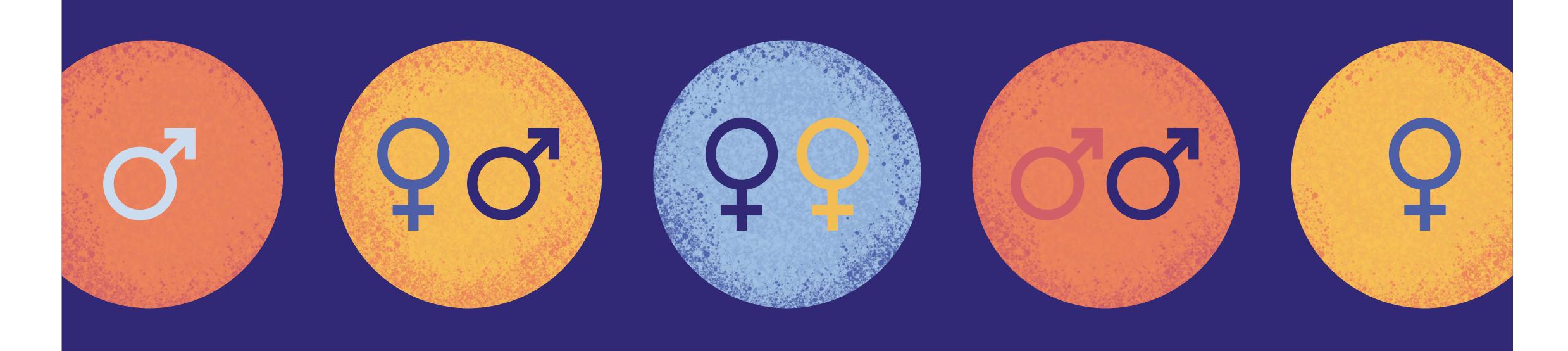


RCTs: The Gold Standard ... of What?



Phase 4: Data Collection & Sourcing

Measuring social constructs (Demographics) Who is constructing the categories? Whose definitions are we using?



Measuring social constructs (Demographics)

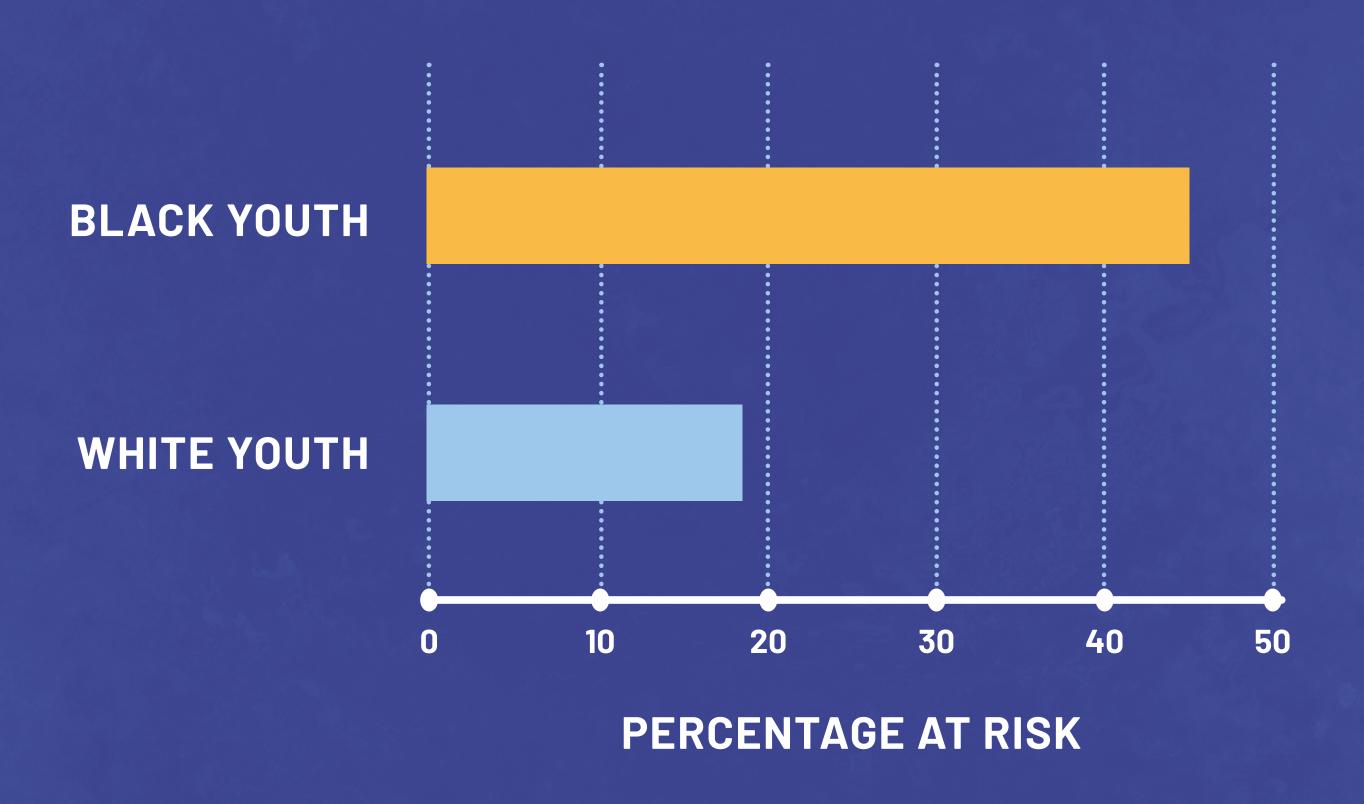
And how are you going to use this?



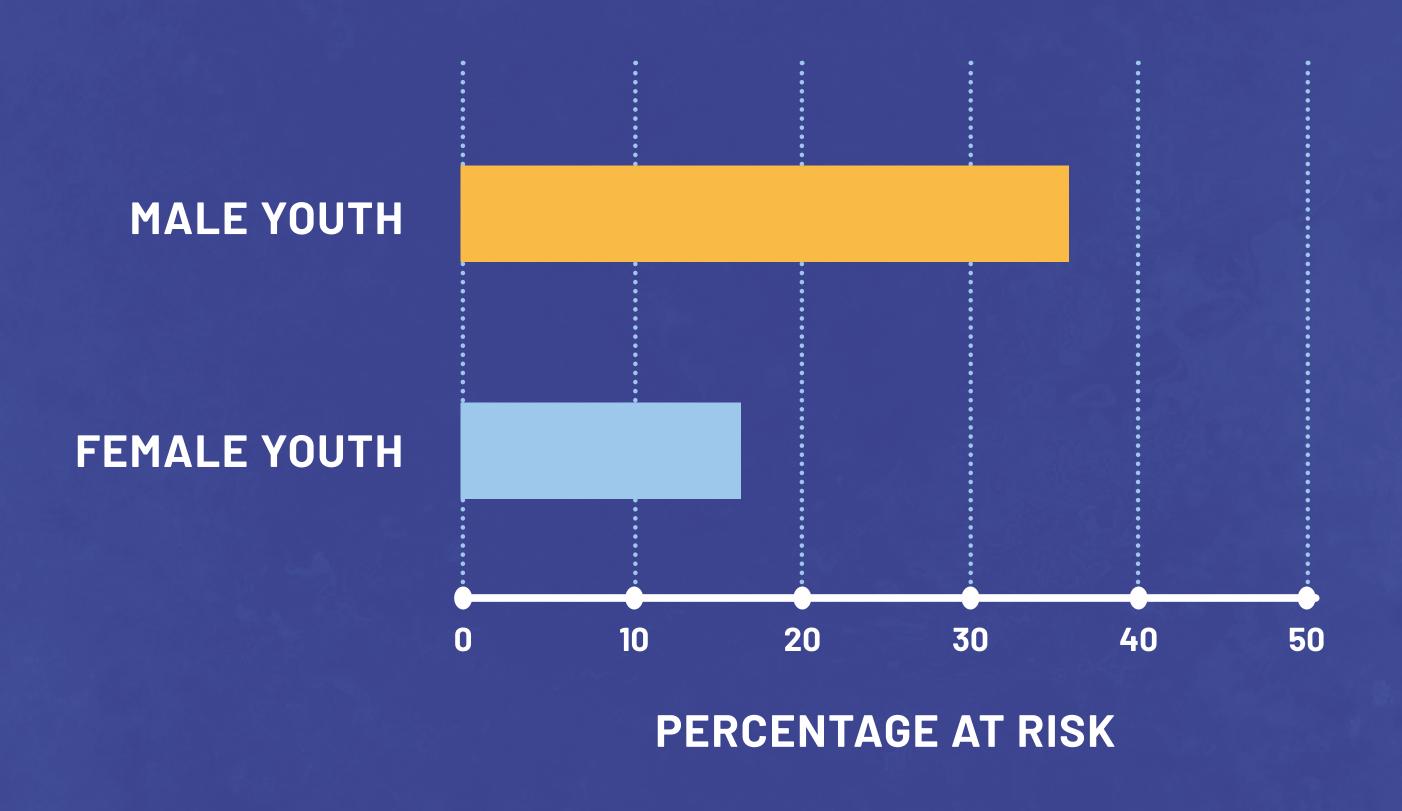


Step 5: Analysis

In our community, which children are most at risk for mental health issues?



In our community, which children are most at risk for mental health issues?



In our community, which children are most at risk for mental health issues?



We saw that the Black children were at risk, that male children were at risk, and children not living in poverty were at risk.

WHITE MALES IN POVERTY
BLACK MALES IN POVERTY
WHITE FEMALES IN POVERTY
WHITE FEMALES IN POVERTY
BLACK FEMALES IN POVERTY
BLACK FEMALES IN POVERTY
BLACK FEMALES IN POVERTY

BLACK FEMALES IN POVERTY

D 10 20 30 40 50 60

PERCENTAGE AT RISK

About Distribution of Key Predictor Variables

BLACK WHITE
MALE FEMALE
NOT POOR POOR

It seems like you're letting the data talk.
But really you're telling the data what to say.



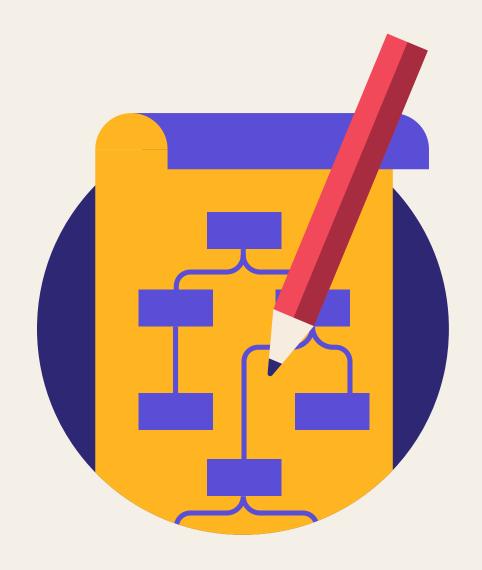


Example:

Project for a school district in Texas.

Research question: "What are the trends in Black, Indigenous and Hispanic male dropout rates compared to White male students?

Which of our two interventions are working best?"



Example:

Project for a school district in Texas.

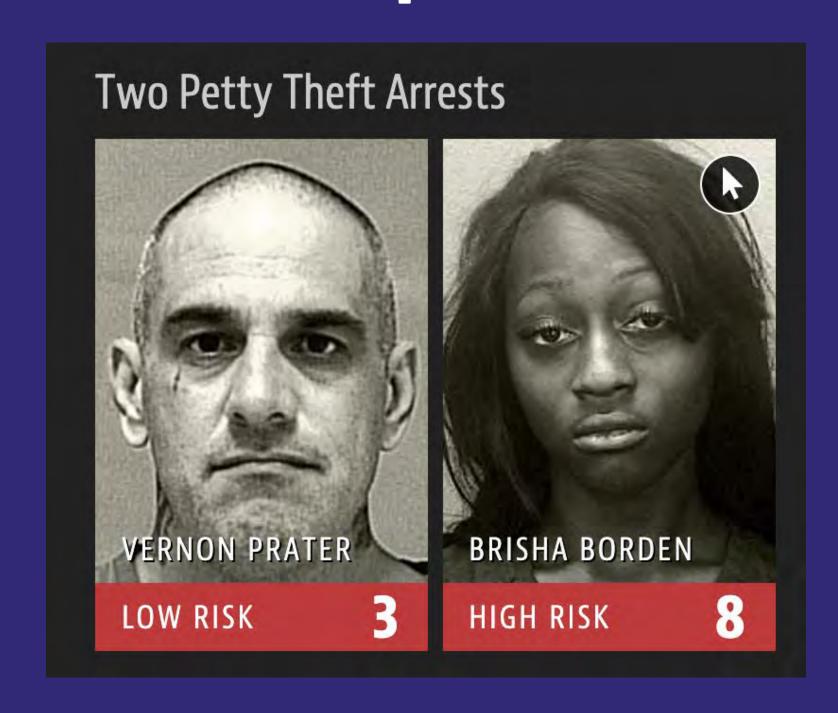
Changed to:

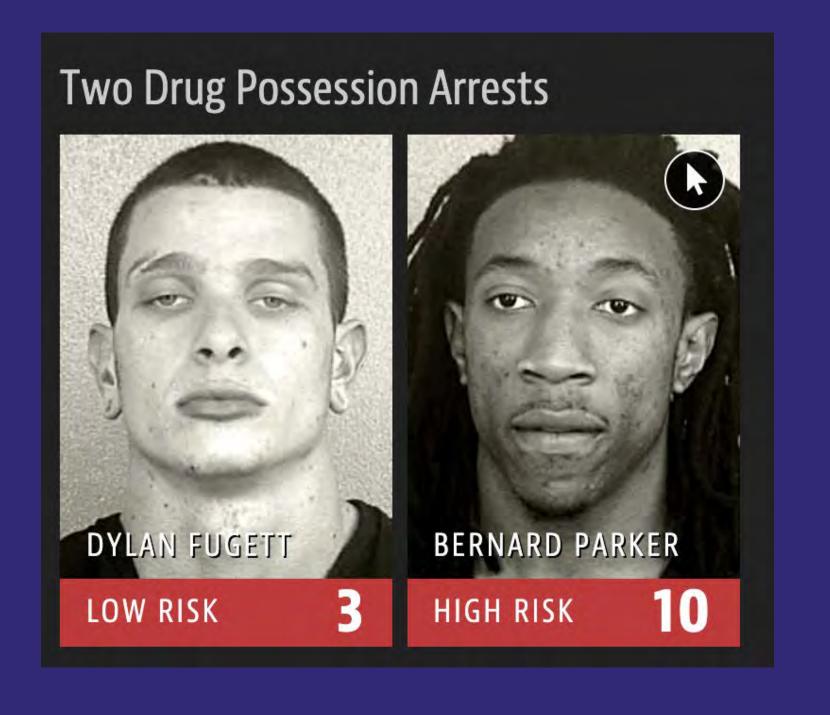
"How much are our interventions effective in removing the barriers we put up to Black, Indigenous and Hispanic boys remaining in school in our district?"

Interpretation



COMPAS gives a score that predicts how likely it is that this person will reoffend.





COMPAS gives a score that predicts how likely it is that this person will reoffend.

COMPAS gives a score that predicts how likely it is that this person might be in contact with the police again, be arrested by those police and not have the money for immediate bail release.

Results Stage

Interpretation Stage



This is the end of the Analysis Step (5).

Results are represented as numbers.

This is where we apply meaning by seeing what our methodology can tell us and how we can interpret the analysis based on our perspective.



This is the end of the Interpretation Step (6).

The numbers have meaning and a narrative to explain them.

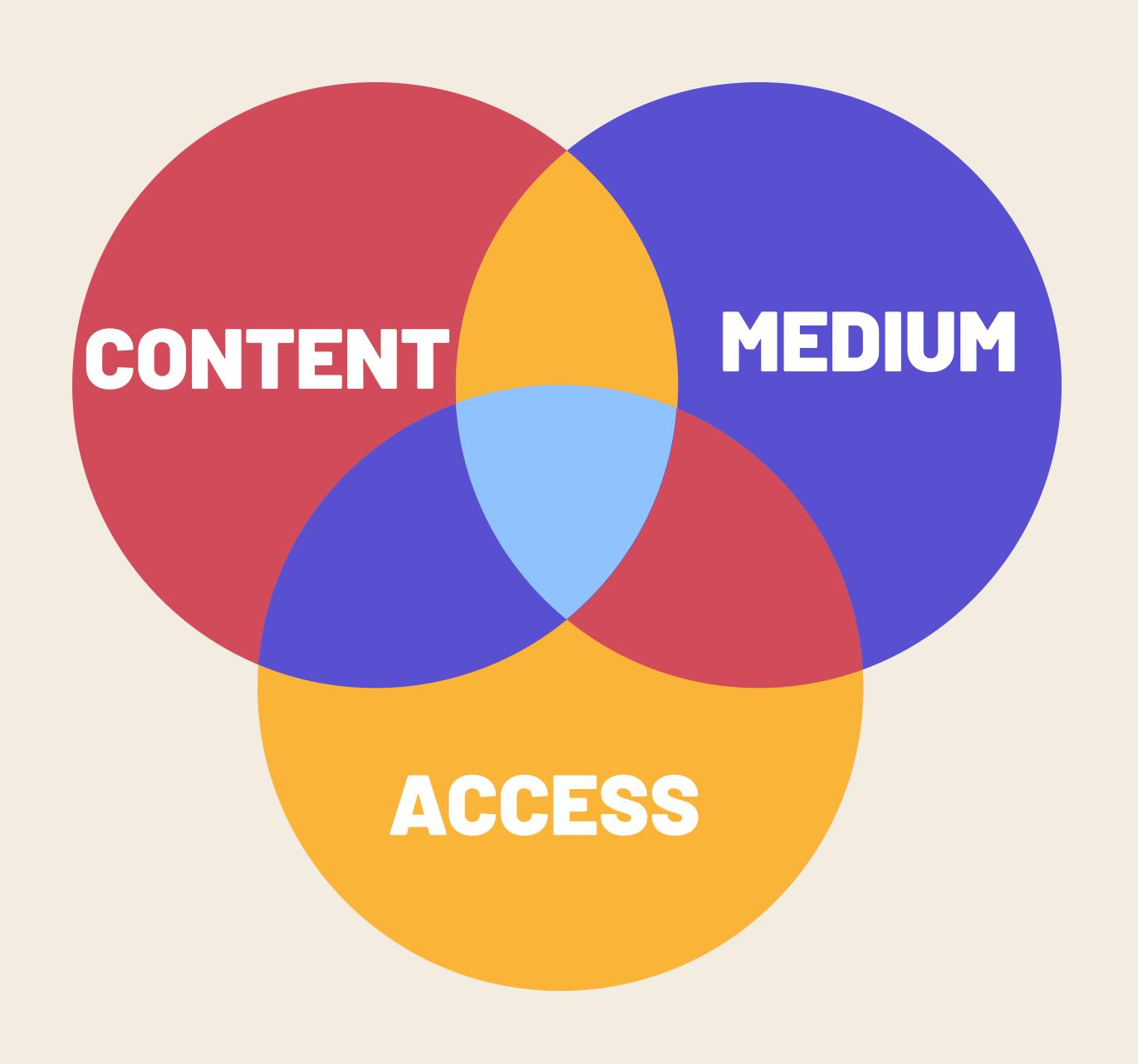
"Our study shows that being black puts you at the highest risk for adult illiteracy."

Communication & Distribution



EQUITABLE DISTRIBUTION

Each audience will have a 'sweet spot' for how to best distribute your information.

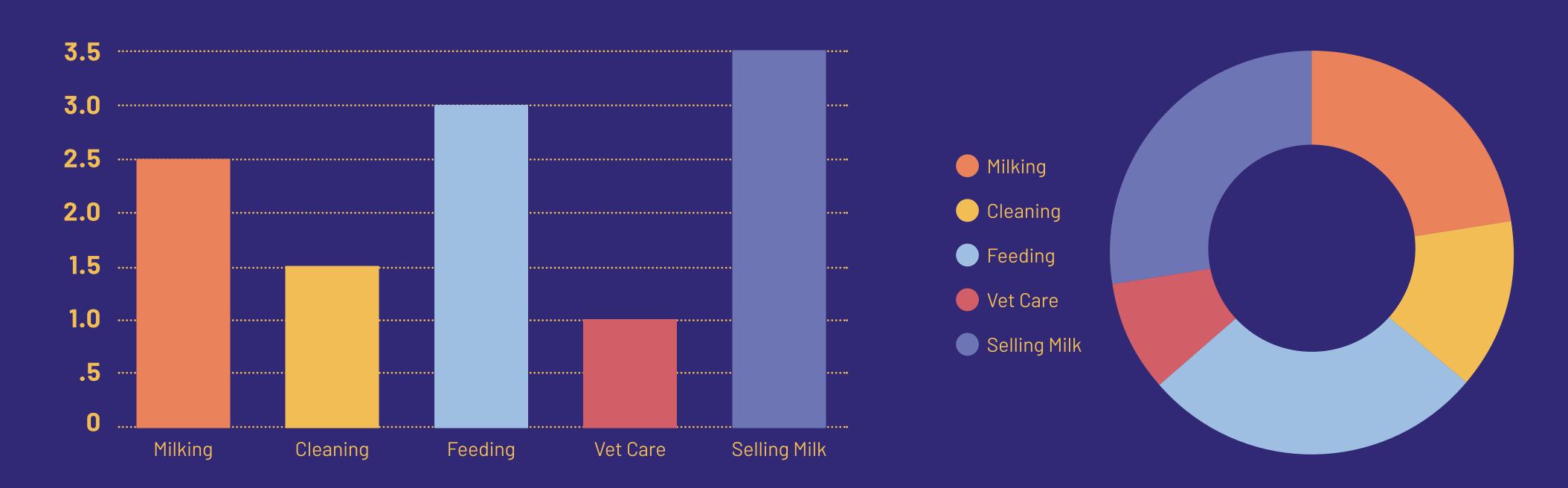


Data Visualization

Ditch your 'best' practices for an adaptive, user-oriented system.

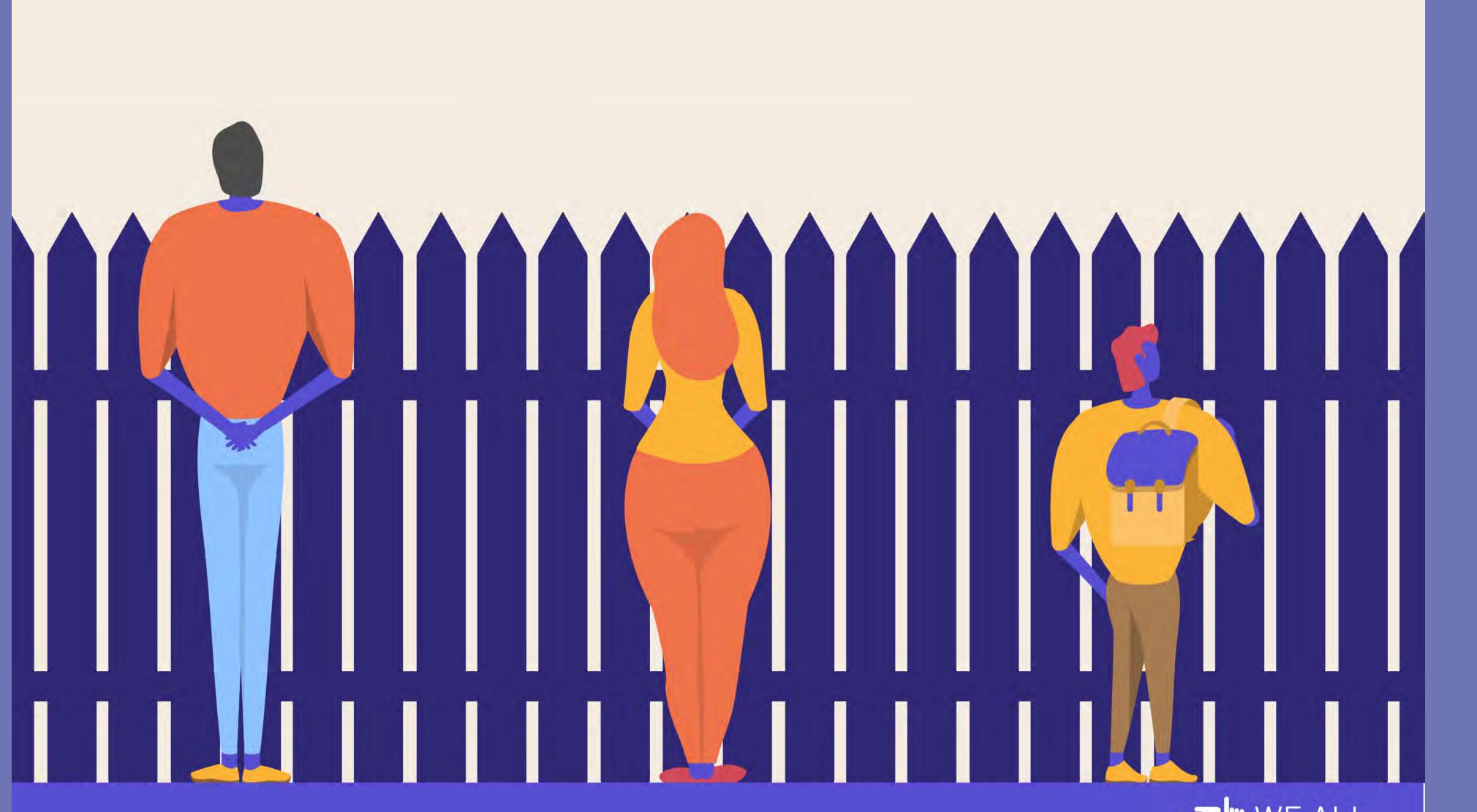


Data Viz "best practices" are not culturally universal.

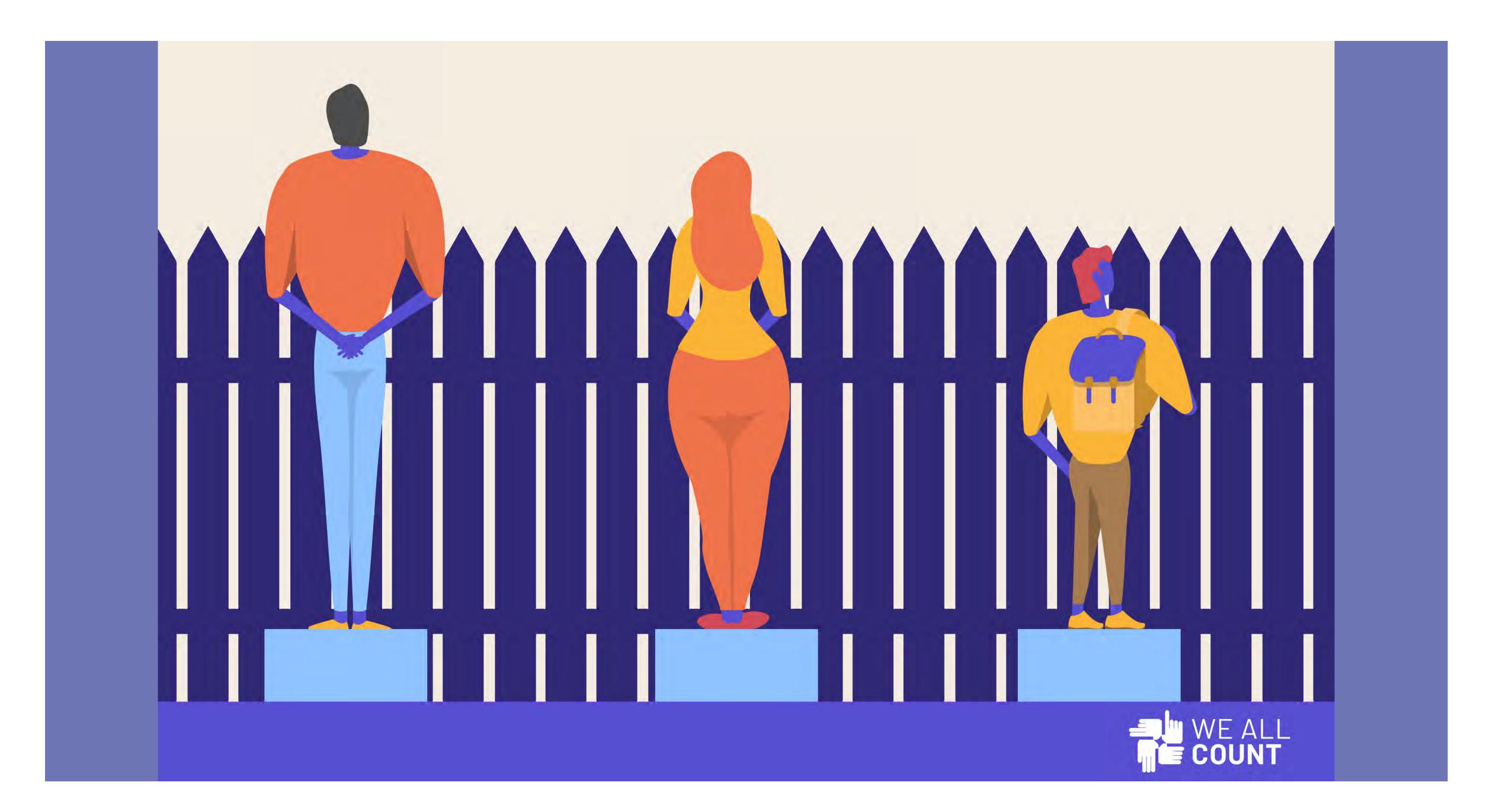


TIME SPENT ON DAIRY ACTIVITIES PER DAY

Why does this matter for equity?

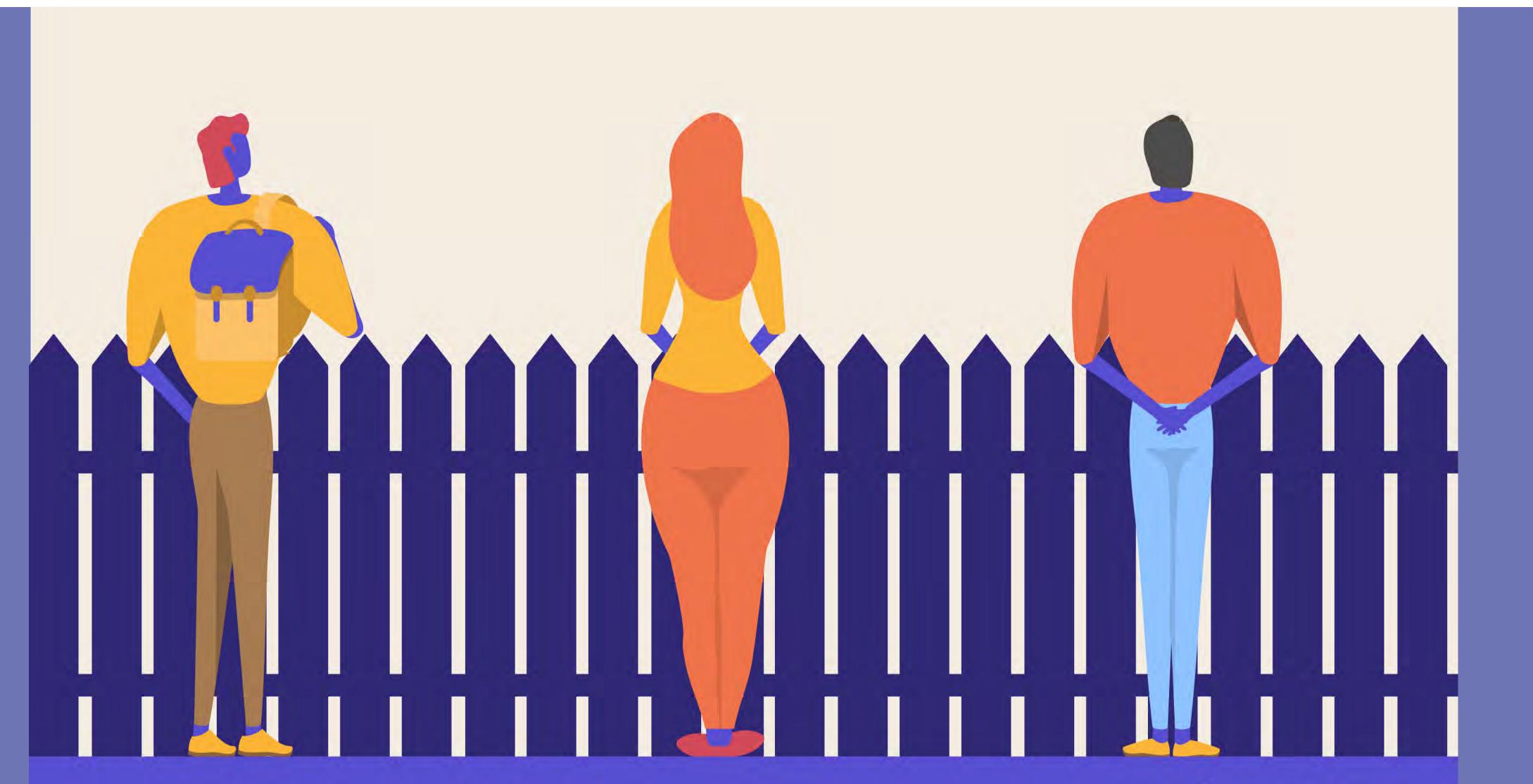










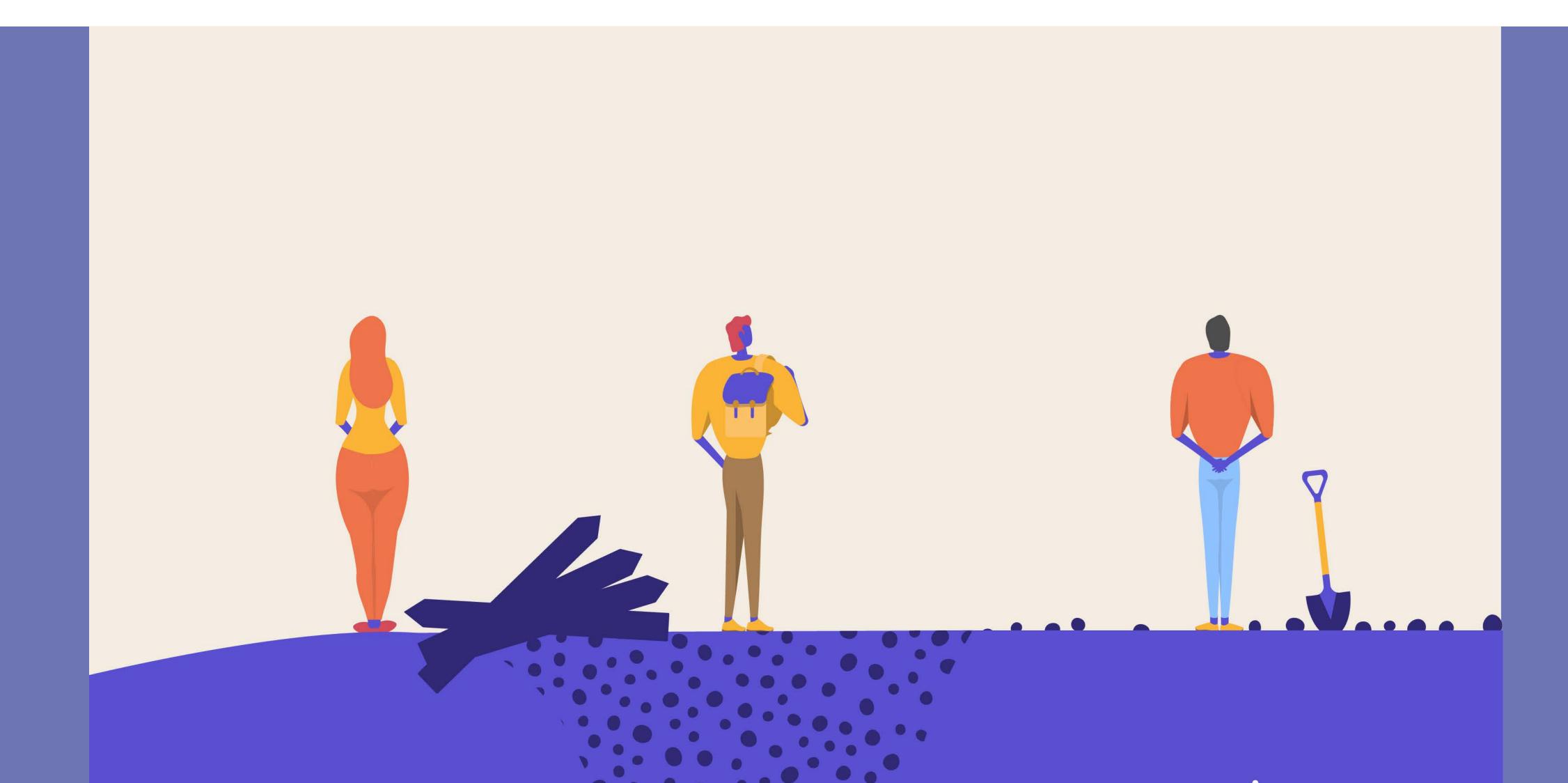














Our data reflects how we see the world, but that's a good thing.

It means we can choose equity.



project for equity in data science

Thank you!

weallcount.com/dop hello@weallcount.com Thank you.

weallcount.com

Heather Krause, PStat

support@weallcount.com

@datassist

