



Community Science

Promoting Child Well-Being by Using Machine Learning Algorithms

Gaithersburg, MD 7/21/2017



THE STATE OF THE CHILD WELFARE SYSTEM

- 3.6 million child abuse & neglect referrals per year, 6.6 million children
- 5-7 child deaths per day
- A child abuse and/or neglect report every 10 seconds
- \$124 billion in economic costs, annually
- NO significant change over the last 10 years
- One of the worst records among all industrialized countries



***TOTALS AND AVERAGES DON'T HELP
THOSE ON THE FRONTLINES***



QUICK OVERVIEW OF TERMS

- Big data
- Social science
- Data science
- Analytics
- Machine learning



BIG DATA



WHAT ARE “BIG” DATA?

Big



Large



Small

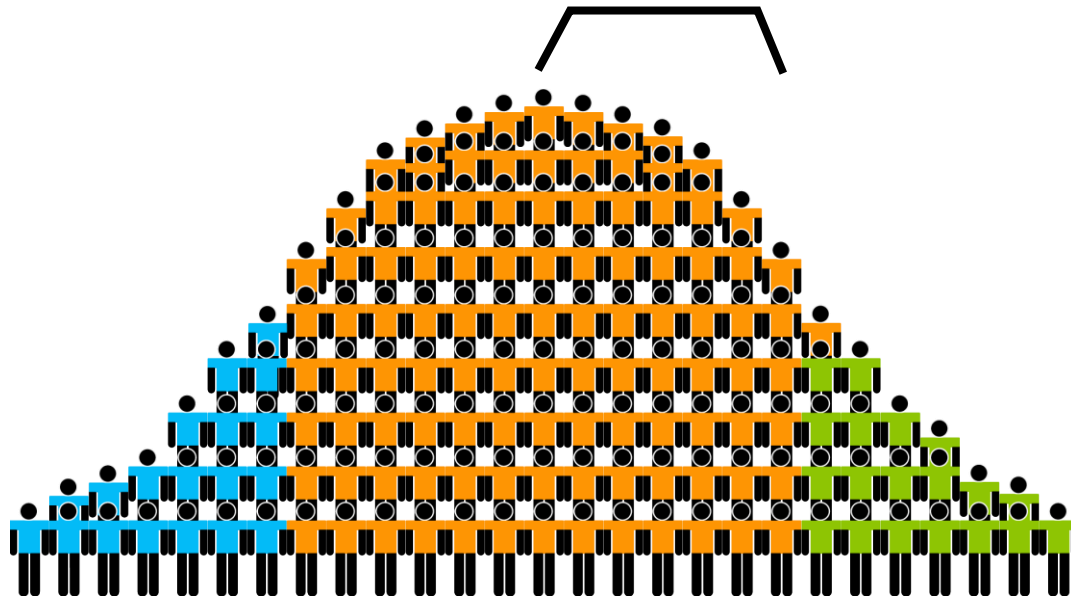


SOCIAL SCIENCE & DATA SCIENCE



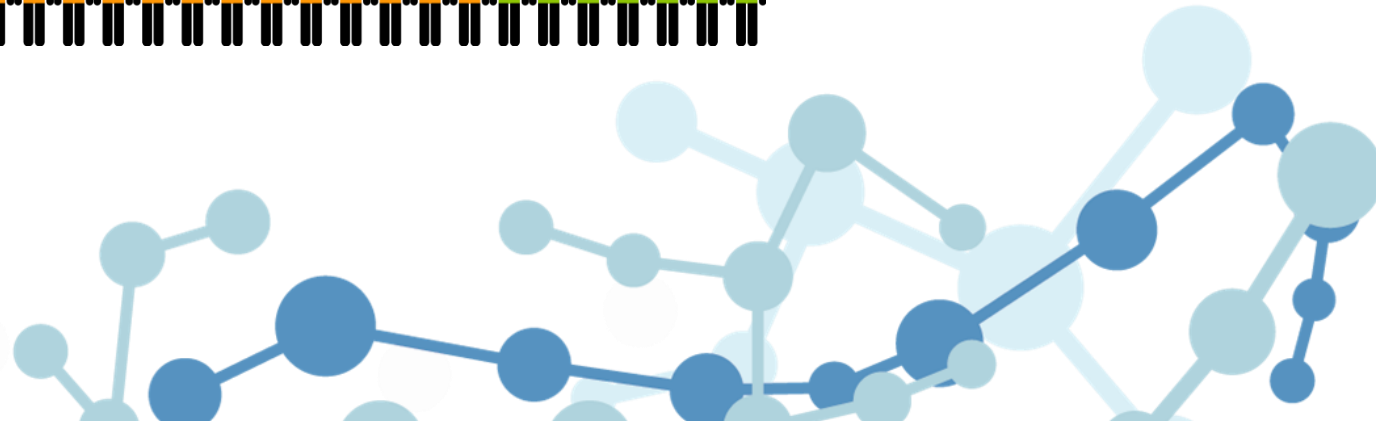
SOCIAL SCIENCE

The Population
Average Moved
($p < .05$)

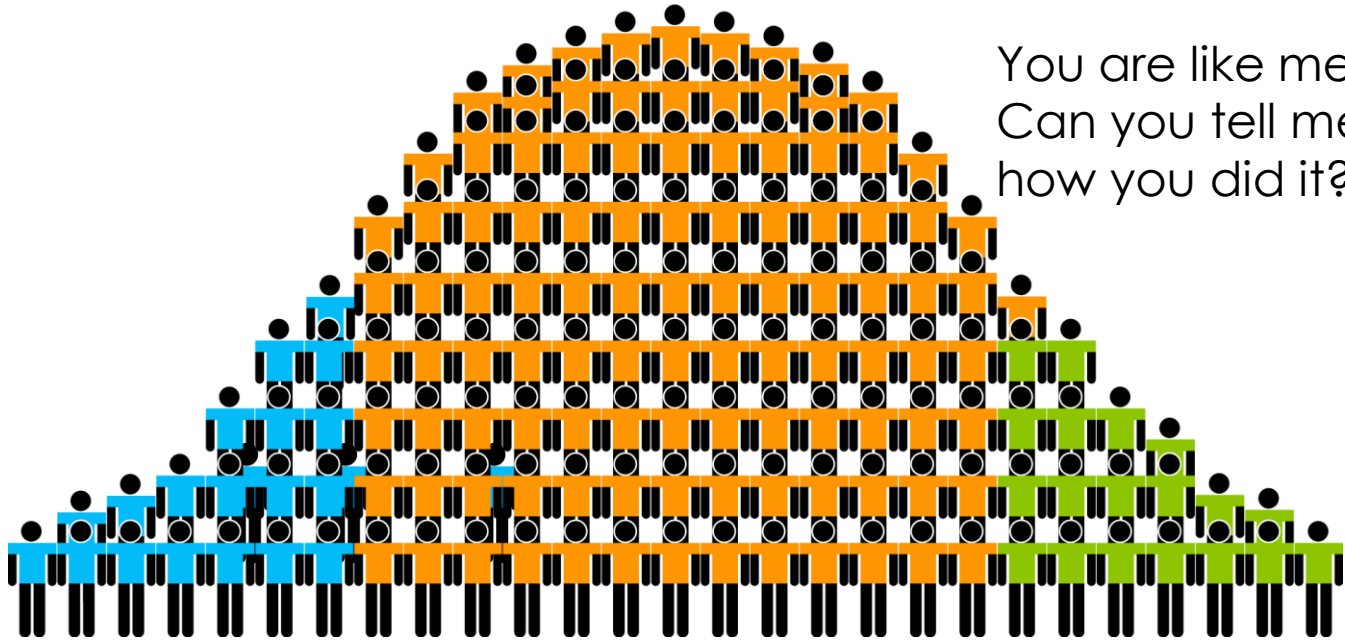


Evidence-Based Decision-Making

If we repeat the intervention as it was implemented during the experiment, the population average will improve 95 times out of 100.



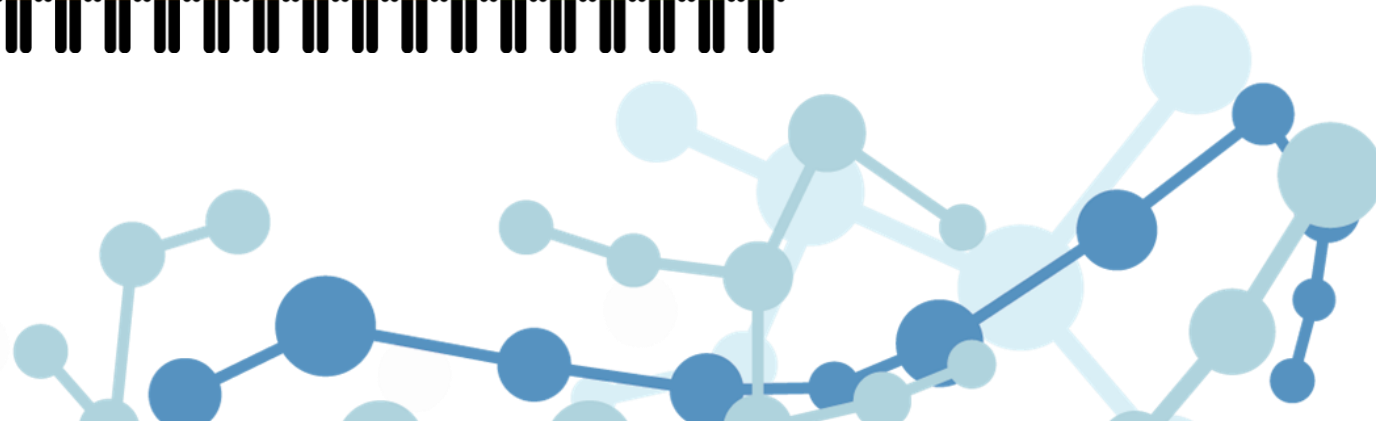
DATA SCIENCE



You are like me!
Can you tell me
how you did it?

Probability-Based Decision-Making

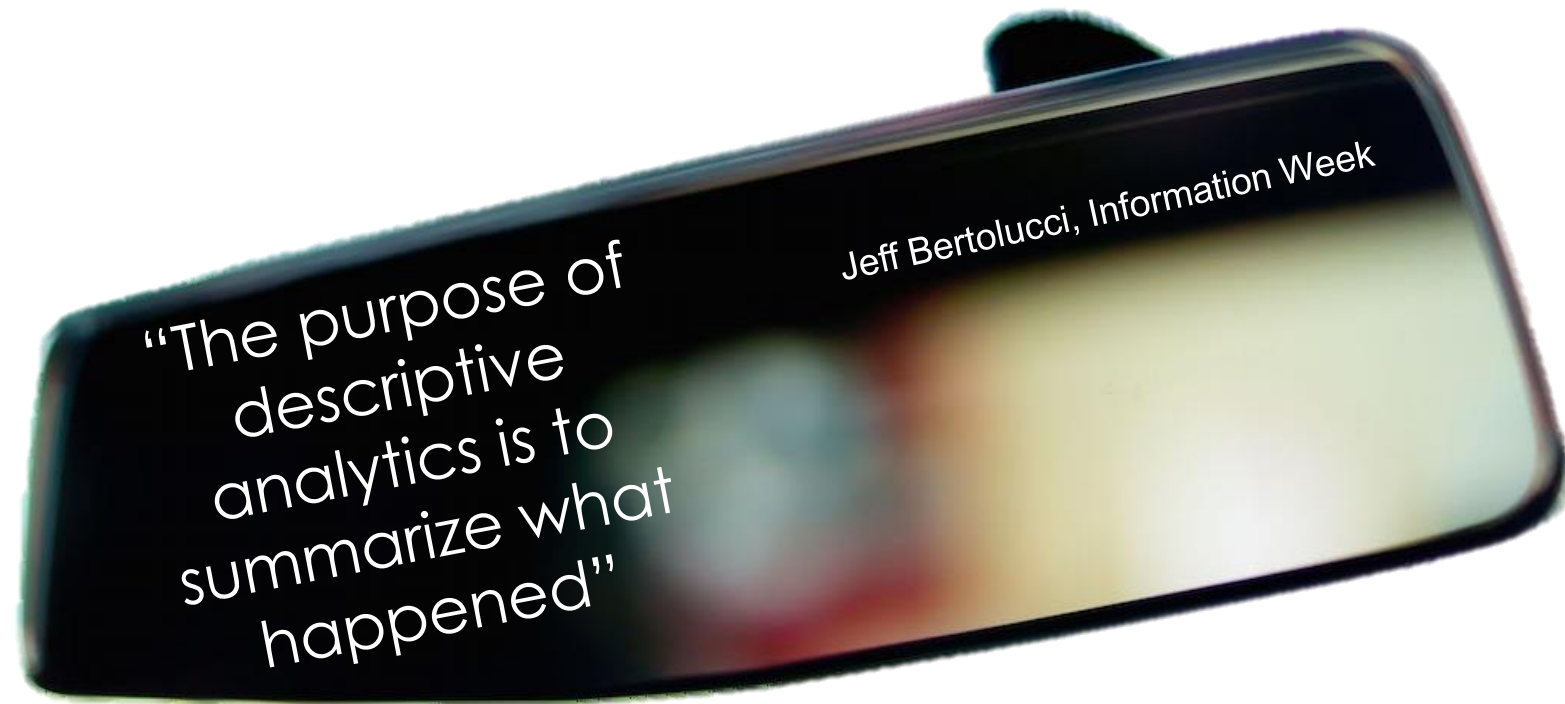
If a specific type of intervention experience worked for 90% of the cases that were very similar to the current case, then there's a 90% chance it will work.



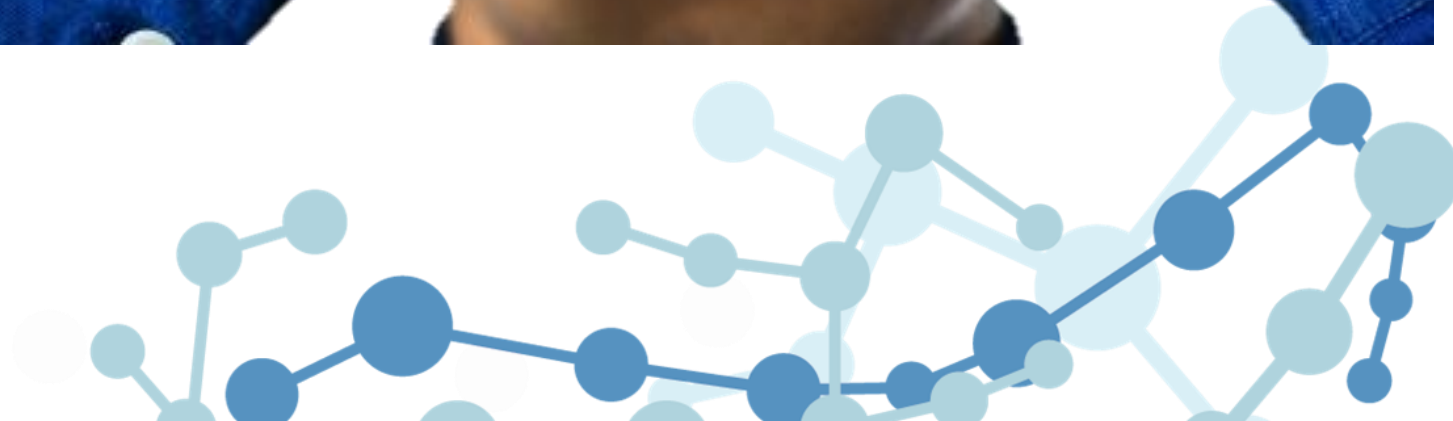
ANALYTICS



DESCRIPTIVE ANALYTICS



PREDICTIVE ANALYTICS



PRESCRIPTIVE ANALYTICS

“...prescriptive analytics tells you what to do about it.”

Jeff Bertolucci, Information Week



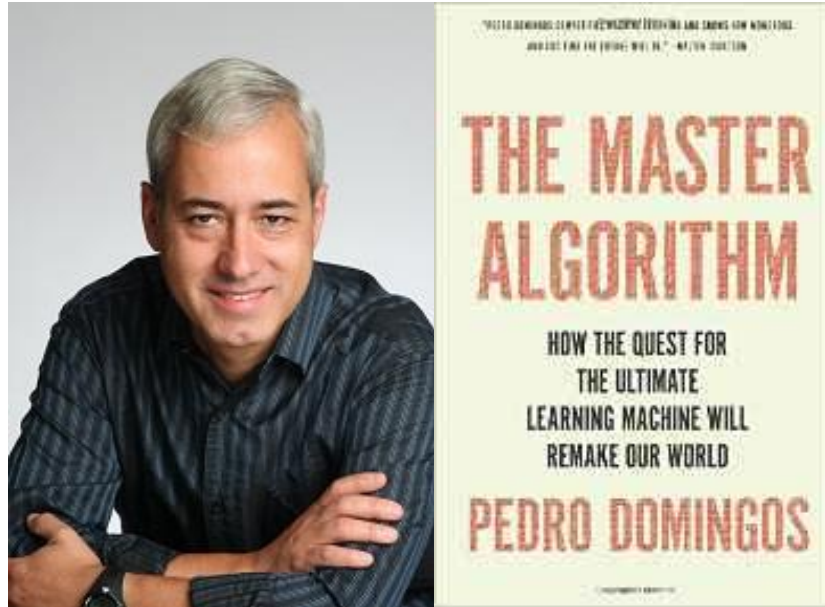
**Find the
Positive
Deviants!**



MACHINE LEARNING



MACHINE LEARNING



Machine learning is “like the scientific method on steroids, making observations, forming hypotheses, testing hypotheses, and refining hypotheses, millions of times faster than any scientists could do.”

Pedro Domingos
Professor
Dept. of Computer Science & Engineering
University of Washington

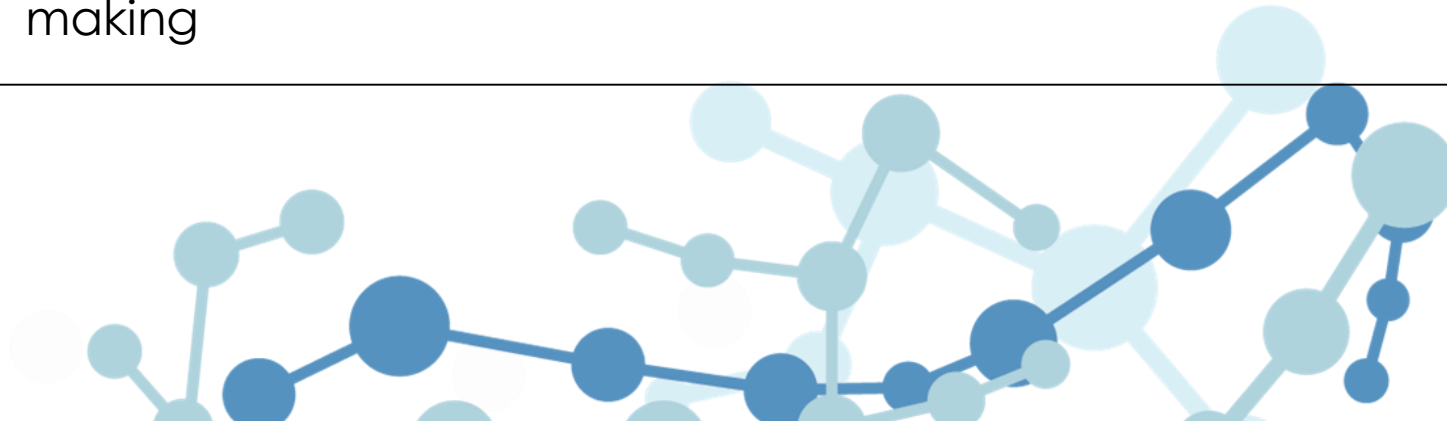


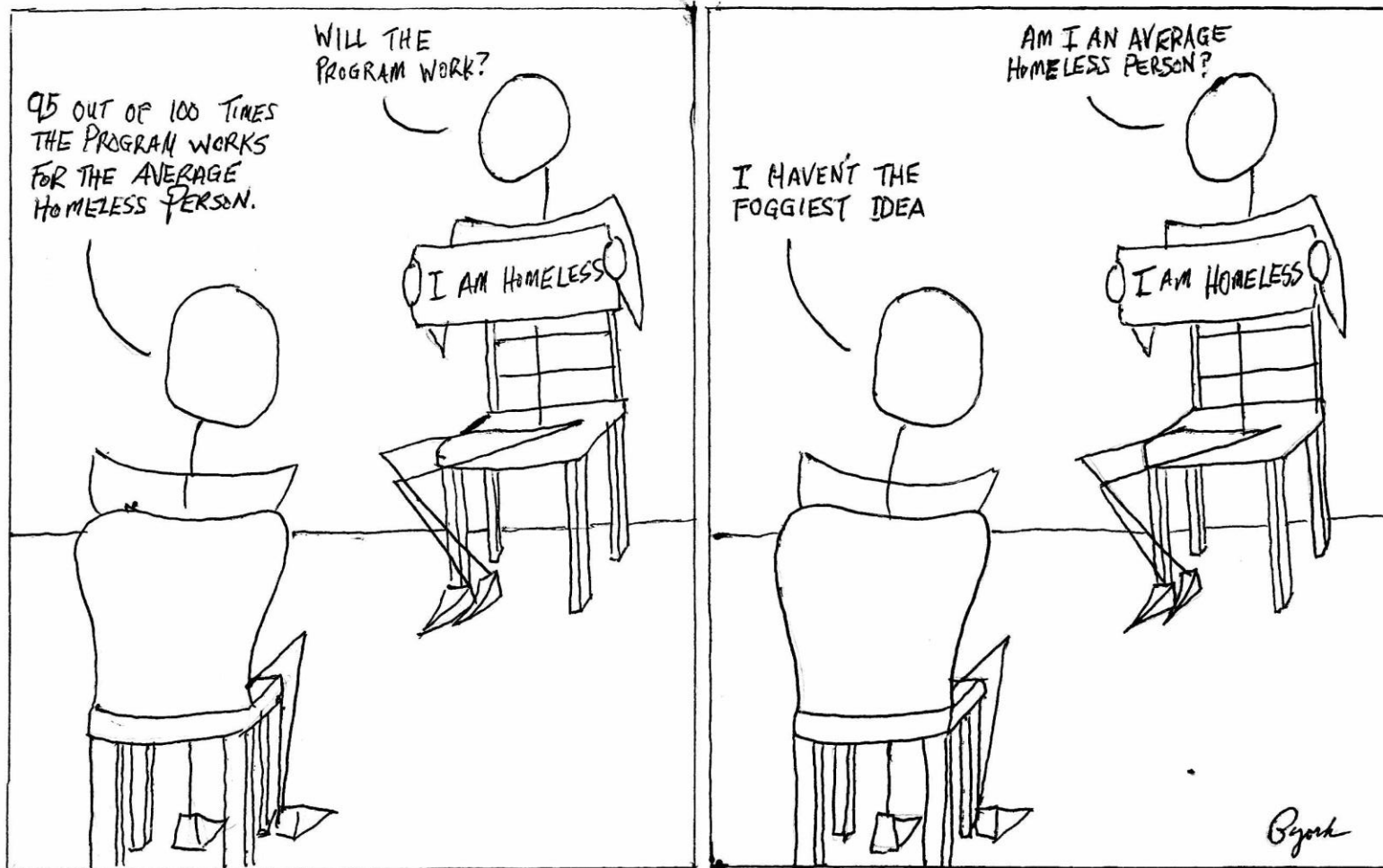
HOW DO WE LEARN ABOUT WHAT WORKS?



EXPERIMENTS

<u>Purpose:</u>	Understand causation for a population
<u>Method:</u>	Conduct small to large scale experiments, collecting new data on subjects, ideally using the scientific method
<u>Metric of Success:</u>	Significant [average] difference between experimental and control group
<u>Pro:</u>	Determines what actually causes improvement for a population (tries to remove bias)
<u>Con:</u>	Can't guide decision-making on case-by-case basis
<u>Implication:</u>	Good for policymaking and funding decisions, not for individual decision-making



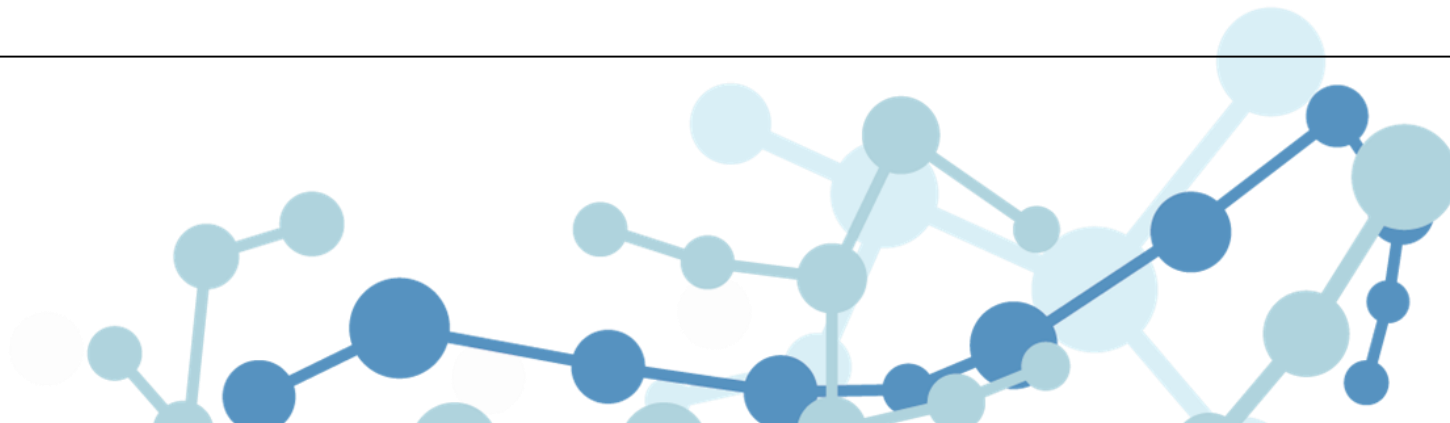


WHEN EVIDENCE-BASED RESEARCH MEETS THE REAL WORLD



ANALYTICS

<u>Purpose:</u>	Predict an outcome for an individual
<u>Method:</u>	Conduct algorithmic modeling (statistical, machine learning/AI) on existing datasets to learn patterns of associations
<u>Metric of Success:</u>	Predictive accuracy across every member of the population
<u>Pro:</u>	Accurate decision-making for individual cases
<u>Con:</u>	Not causation (includes bias in the models)
<u>Implication:</u>	Good for case prediction, not for unbiased prescription or evaluation



THE BIG IDEA

BRIDGING SOCIAL SCIENCE AND DATA SCIENCE, USING MACHINE LEARNING



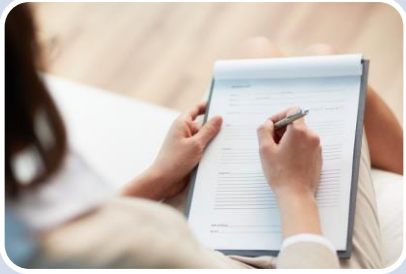
SIMPLE INSIGHTS FOR ACTION

SIX STEPS



STEP 1: ACCESS & PREPARE ADMINISTRATIVE DATA

MACHINE LEARNING CAN ADDRESS THE 'GARBAGE IN, GARBAGE OUT' DATA QUALITY PROBLEM



Intake
History
Background
Situation

Interventions
Dosages
Methods
Experiences

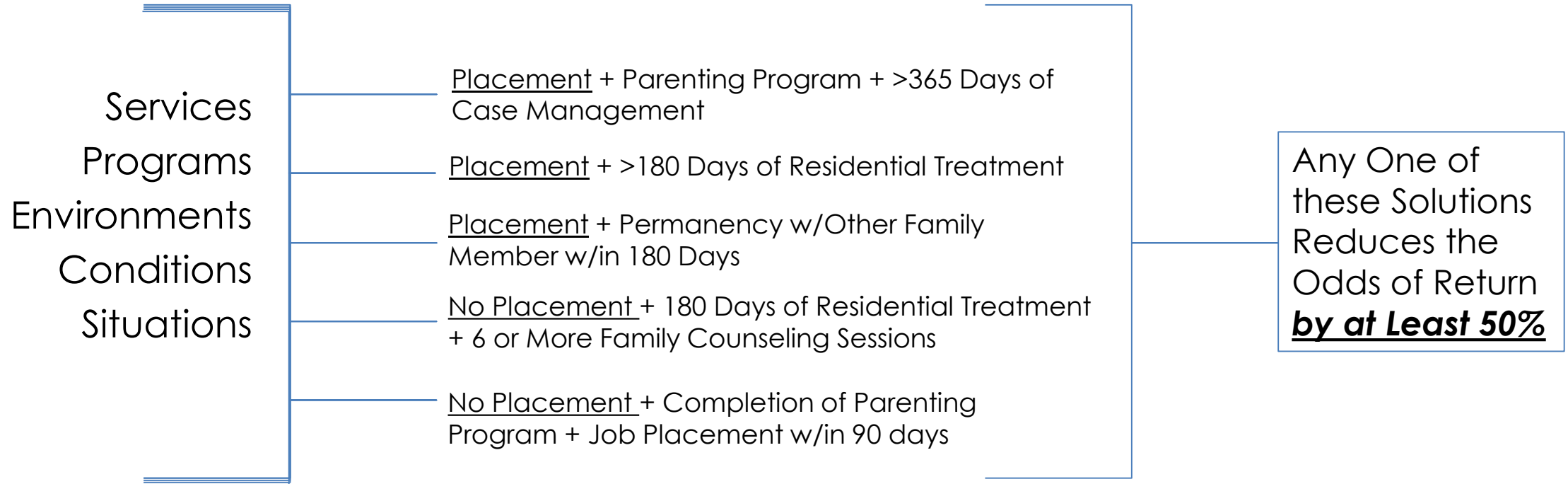
Outcomes
Results
Milestones
Goals

1. More than a few hundred cases
2. Clean up missing data issues
3. Don't sweat statistical assumptions



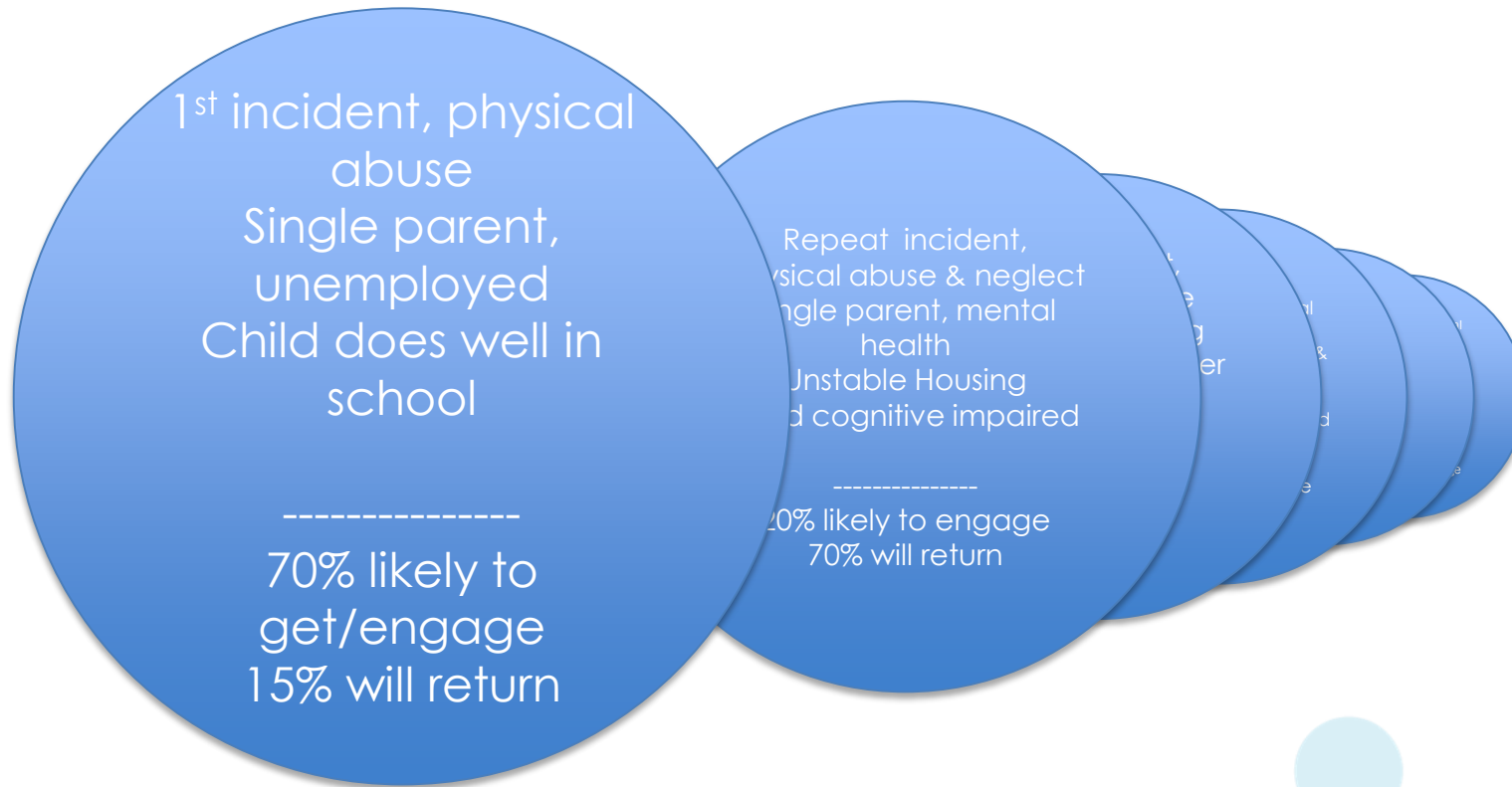
STEP 2: DISCOVER WHAT WORKS

MACHINE LEARNING ALGORITHMS FIND EVERY PATHWAY TO SUCCESS



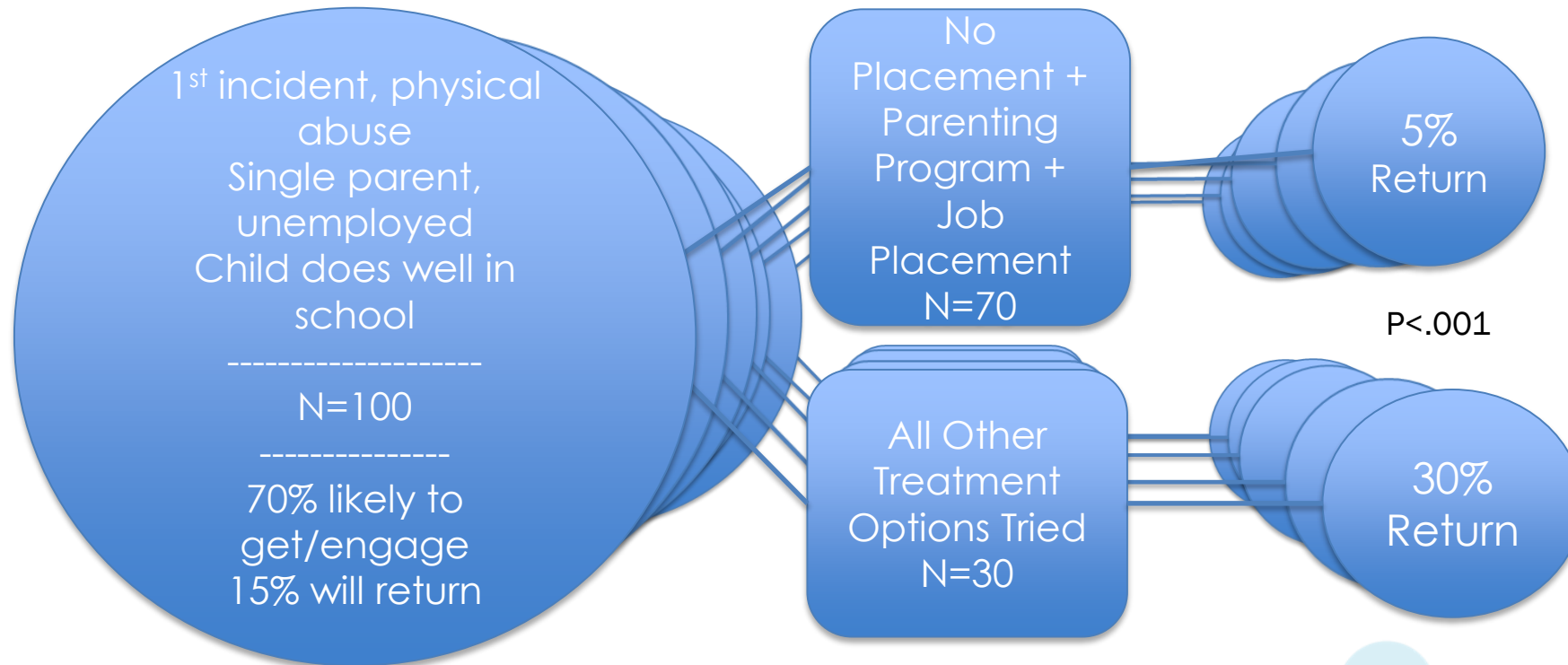
STEP 3: FIND MATCHED COMPARISON GROUPS

BACKGROUND & HISTORY THAT PREDICT LIKELIHOOD TO ENGAGE IN AND/OR RECEIVE WHAT WORKS



STEP 4: DETERMINE WHAT WORKS FOR EACH GROUP

MACHINE LEARNING FINDS NATURALLY OCCURRING EXPERIMENTS WITHIN THE DATA



STEP 5: EVALUATE SUCCESS

MODELS EVALUATE HOW MANY CASES GOT WHAT WAS NEEDED, NOT POPULATION AVERAGES



Got What Worked & Didn't Return

$$70 + 90 + 85 + 75 + 55 = 375$$

75% Didn't Return

Due to Receiving the Right Solutions



STEP 6: USE THE MODELS FOR DECISION MAKING

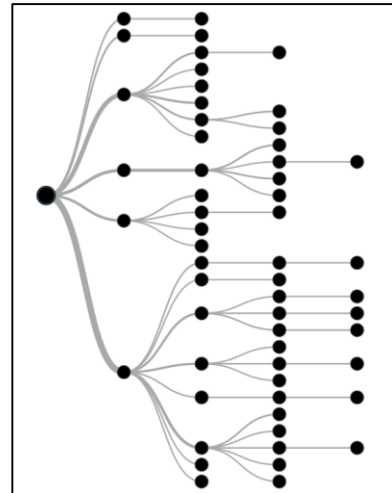
BUILD APPLICATIONS & LEARNING COMMUNITIES



**NO NEW DATA
SYSTEM REQUIRED!**



SCORING ENGINE



**DECISION MAKING
MODELS**



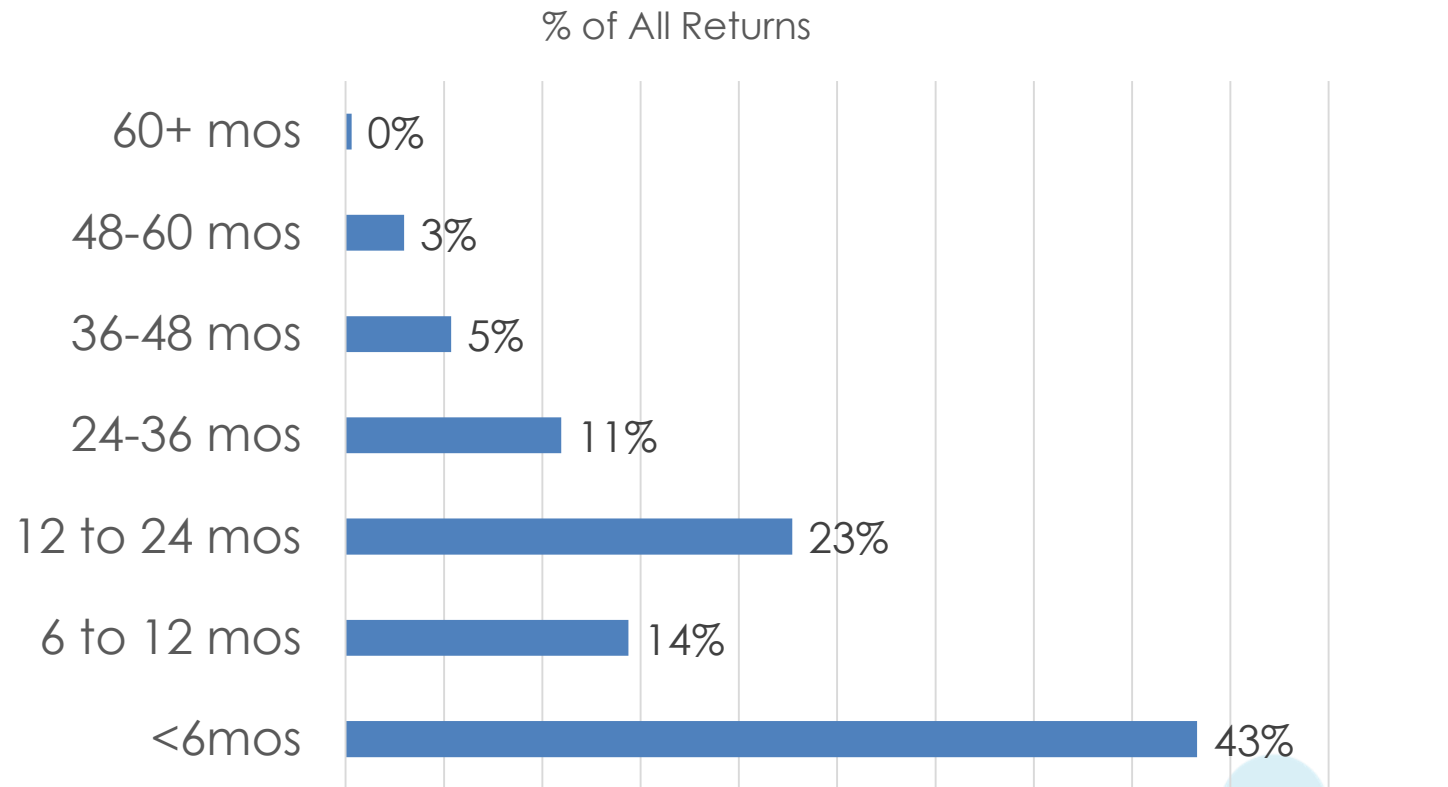
COMMUNITY OF LEARNING



BROWARD SHERIFF'S OFFICE



BROWARD SHERIFF'S OFFICE: 24% OF ALL CASES RETURN



BROWARD SHERIFF'S OFFICE

- Accurately predicts next incident in 83 cases out of 100
- A prescription model that can accurately recommend which cases should go into either intensive out-of-home services or in-home community-based programs, and if these recommendations were applied, the result would be a 30% reduction in return cases
- A rigorous impact evaluation model, using machine-guided matching, that proves that in 60% of the cases, intensive out-of-home case management services from a specific provider significantly reduces their likelihood for a subsequent incident.
- A rigorous impact evaluation model that also proves that 40% of the cases are misplaced in these intensive services, and as a direct result, are 175% more likely to return with a subsequent incident.
- Specific investigators make a difference.



CASEY FAMILY PROGRAMS



FIRST PLACE FOR YOUTH



SYSTEM IMPLICATIONS



IMPLICATIONS- A CAUTIONARY TALE

- Data analytics do not have to be the next "shiny new object;"
- Changing child welfare practice means developing the capacity of organizations, systems and communities;
- Outside knowledge is needed;
- Algorithms inherit system values and bias;
- Don't forget social or contextual factors; and
- The purpose is to accelerate learning how to protect and improve the well being of children and youth.



QUESTIONS



CONTACT FOR FURTHER INFO

Peter York
Principal Associate
Community Science
pyork@communityscience.com

