Is Statistics giving you fits? How to think about data and statistical methods for driver behavior and safety in a changing transportation world

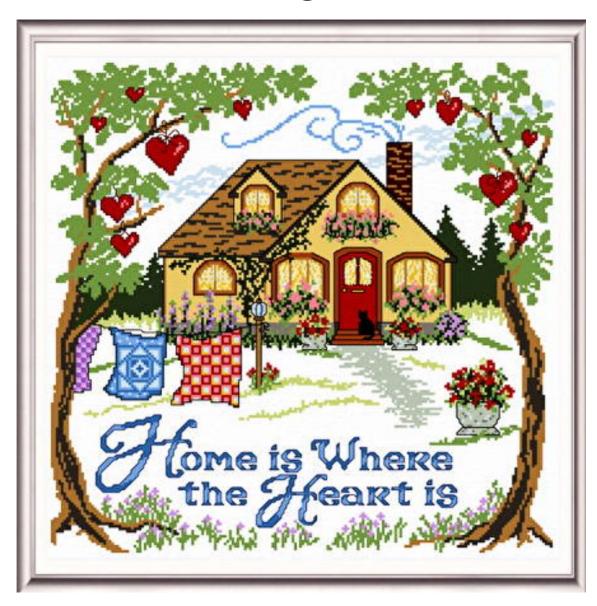
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Some Background First



Statistics is the Unfun Parent of Data Science

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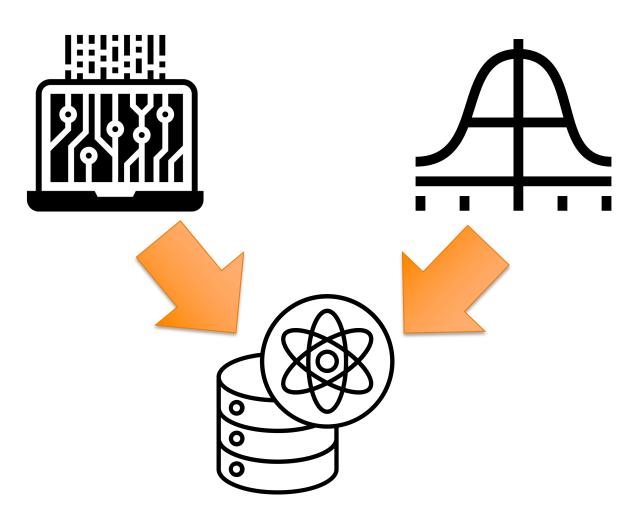


"Let me fill you in on a little secret...When I met your Dad, I was fun too. But I had to give all that up, because you can't have two fun parents. That's a carnival."

- Claire Dunphy

Statistics is the Unfun Parent of Data Science

Fun parent



Unfun parent

If Data Science = Computer Science + Statistics...

Good things that happen to other people are not bad things that happen to you (and vice versa)

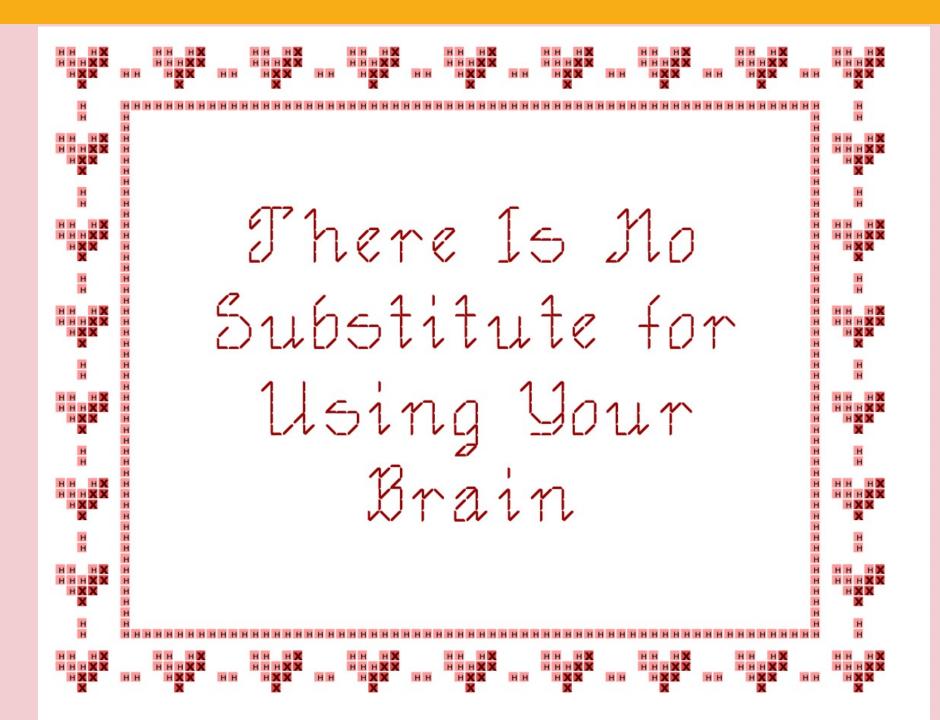
You need to have a plan and a goal

Long-delay feedback systems tend to oscillate

Separate the hazards

Unfun Parent

Fun Parent



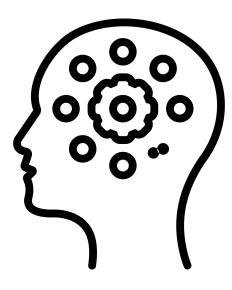
Know your analytical goals



Description

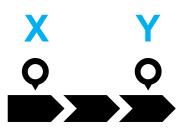


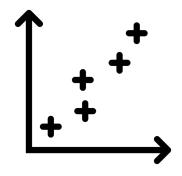
Prediction



Inference

Causal inference requires:



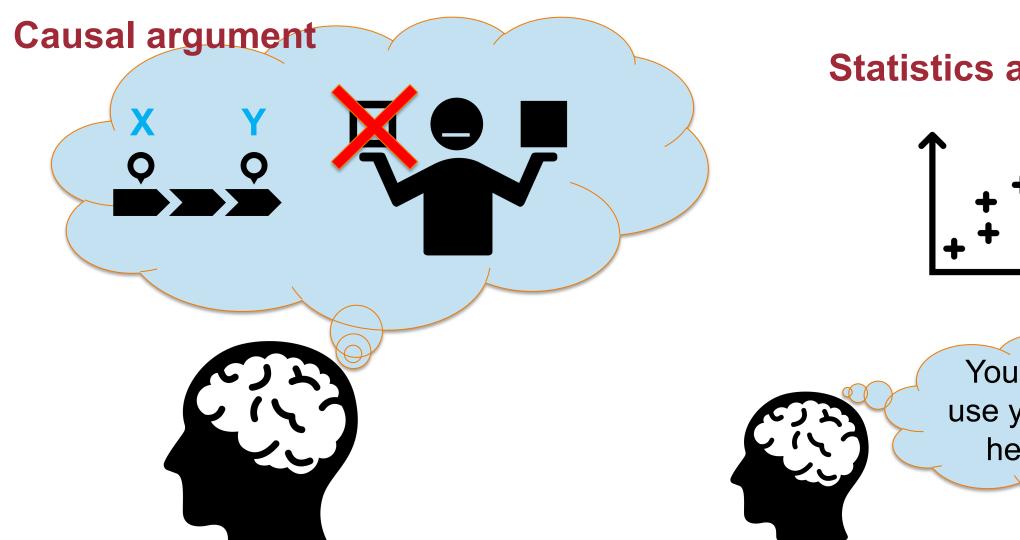




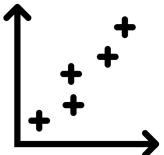
Temporal precedence

Covariation

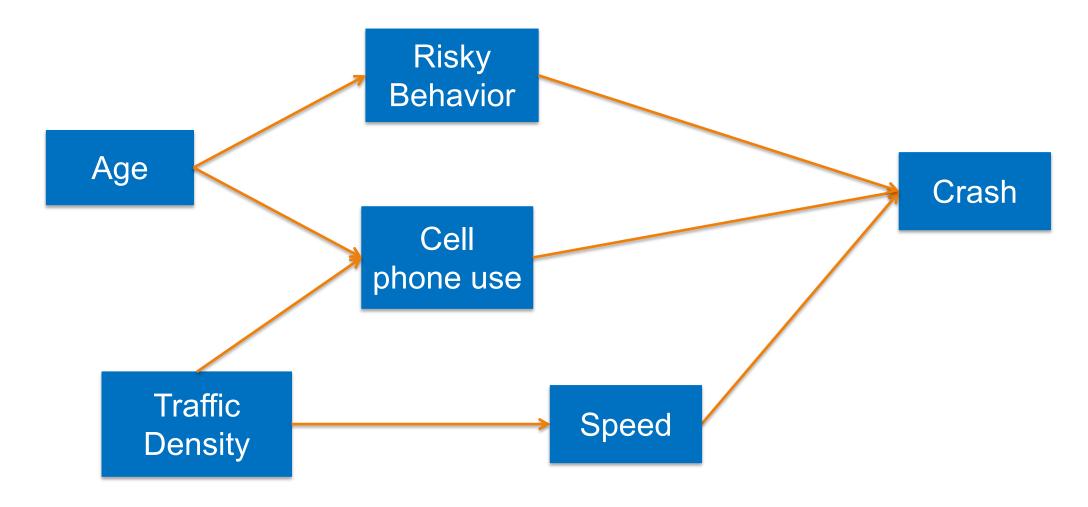
No Alternative Explanations



Statistics and data



You have to use your brain here too!



Directed Acyclic Graph => DAG

Requirements for estimation of causal effects:



1. Consistency: the cause is sufficiently well-defined



2. Non-interference: observations are independent



3. Exchangeability: "cause present" and "cause absent" groups are the same in all other ways that matter



4. Positivity: anyone in the dataset could have been in the "cause present" or "cause absent" group

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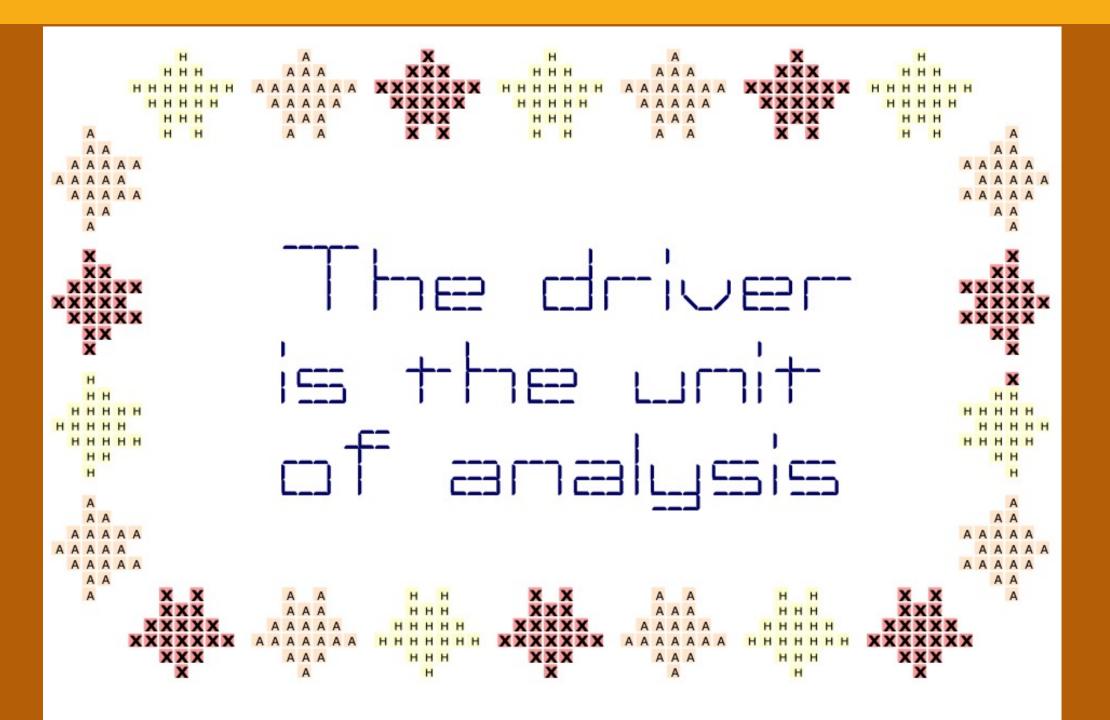
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4. Positivity: anyone in the dataset could have been in the "cause present" or "cause absent" group



Example: Do L3 ADS drivers respond to takeover requests more slowly when they are on their cell phone?





My dataset has 10,000 trips!!!

Dataset A



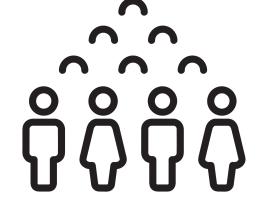
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10,000

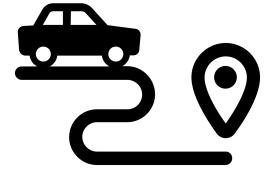


Dataset B



10,000

@



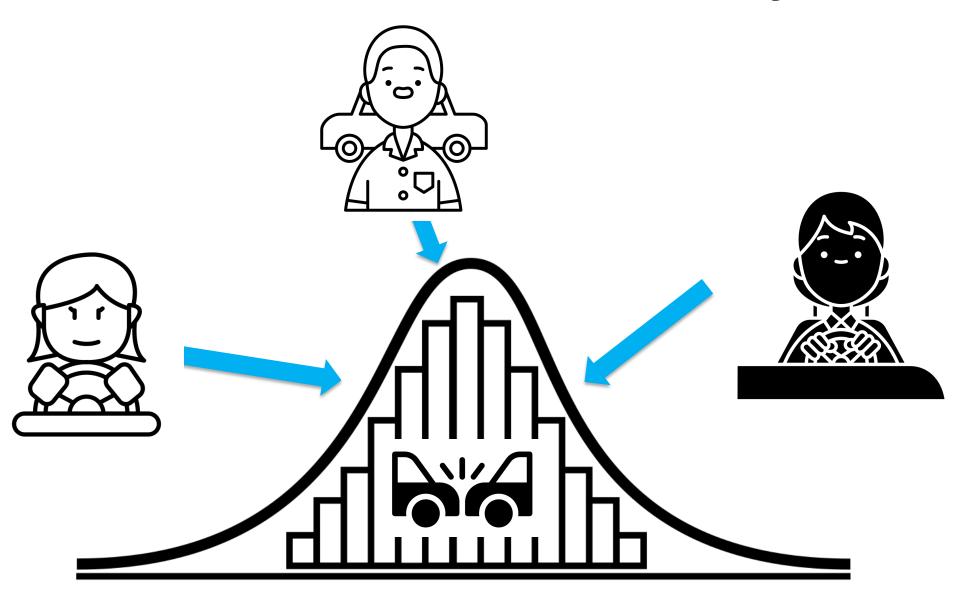
Why do we care about Big Data?

Big sample size

Why do we care about sample size?

- Generalizability
- Statistical power

My dataset has 10,000 trips!!!



Statistical power depends on:

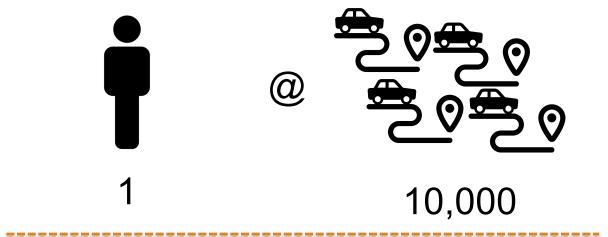
- 1. Sample size
- 2. Variability in the outcome
- 3. The size of the difference of interest
- 4. The selected significance, or Type I error level (typically 0.05)
- 5. The statistical test being used

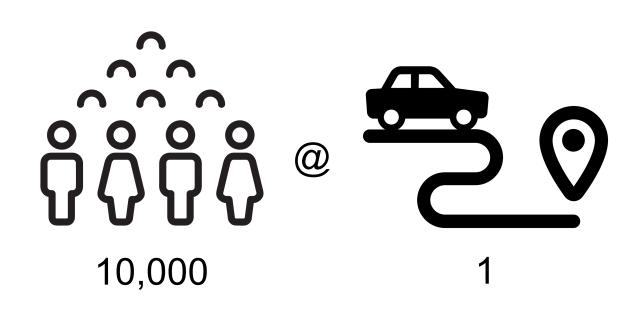
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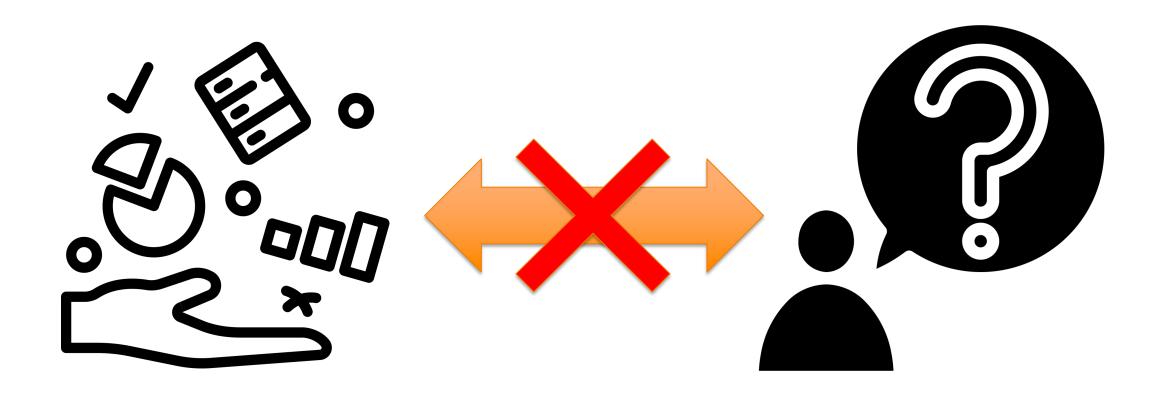
Statistical power:

- More trips/person = reduces variability in the estimates of takeover time
- But sample size = # drivers
 Generalizability:
- More observed variability sources = more generalizability



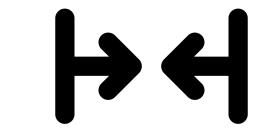




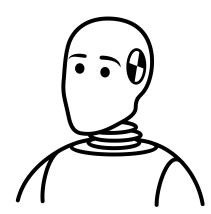


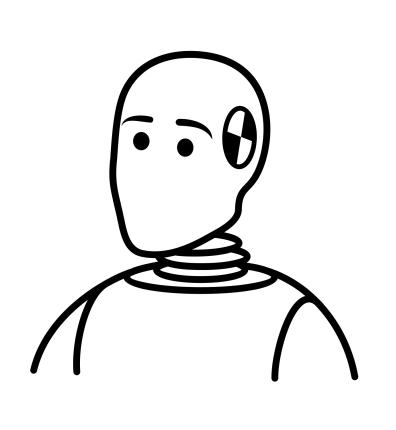
Adjust the research approach

1. Make inferences about the right things

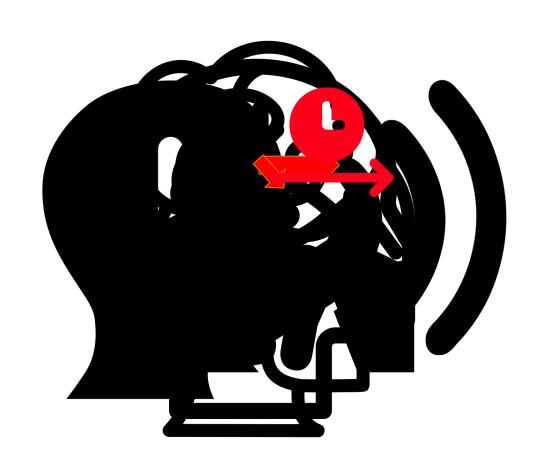


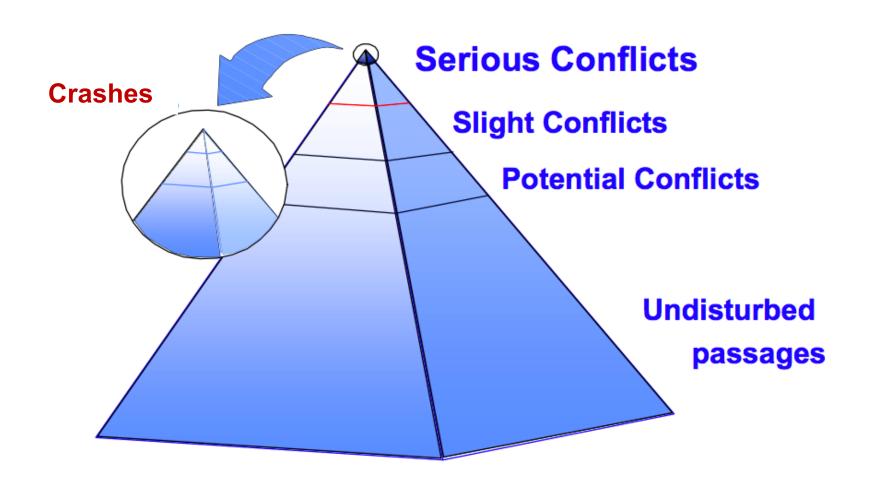
2. Use surrogates











For safety, common surrogates are *driver-controlled* kinematics

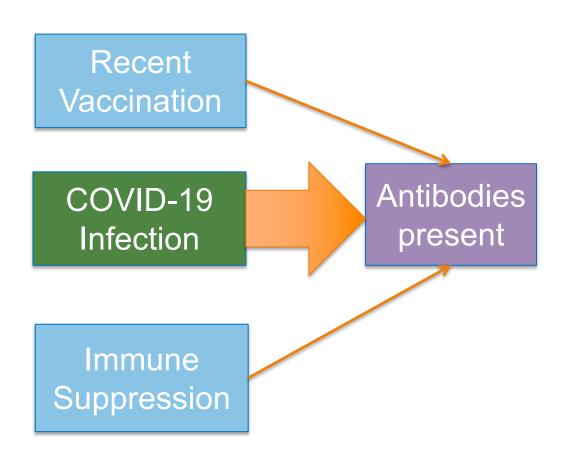


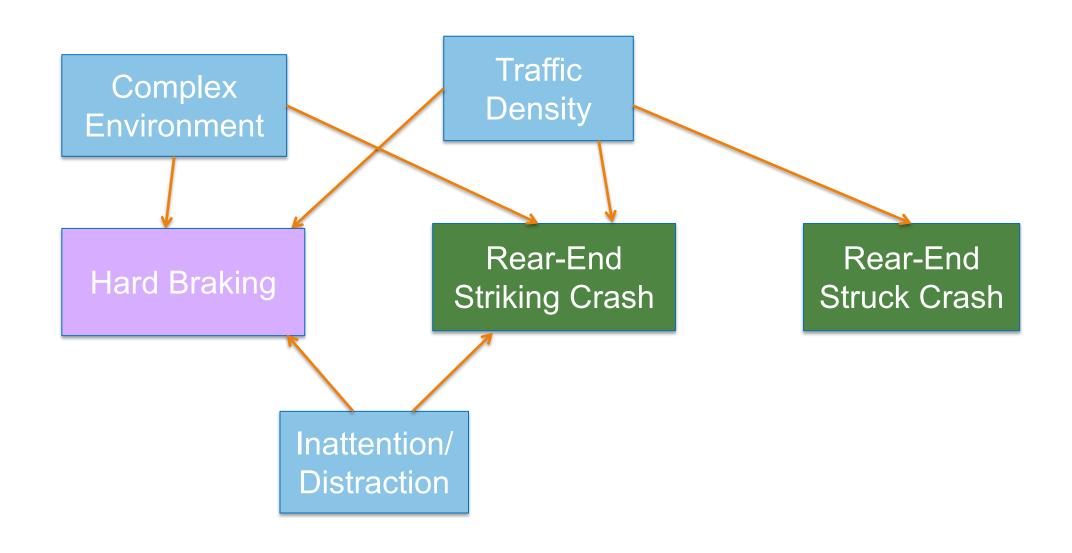








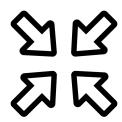




What to do?



1. Know your surrogate causal model



2. Narrow your focus



3. Clarify interpretation

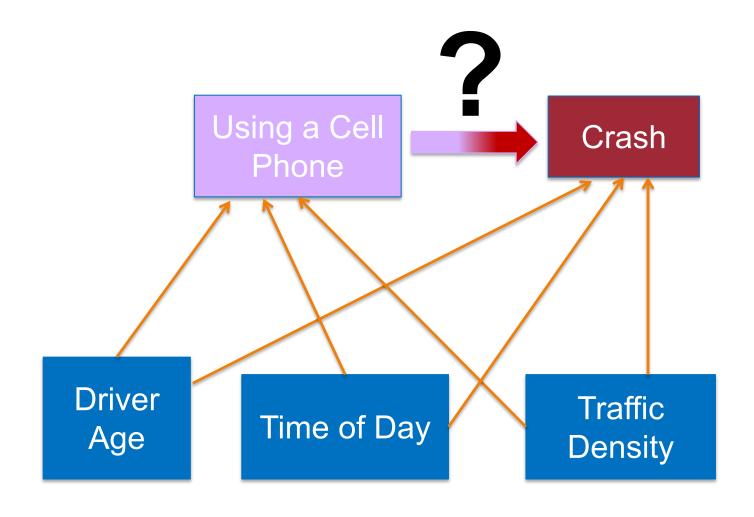
нн X X H H X X Context matters.

(predicted and predicted asset risk) crash risk)



Why do we care?

EXCHANGEABILITY



No adjustment: OR = 2.38

With adjustment: OR = 1.98

From: Flannagan, C., Bärgman, J., & Bálint, A. (2019). Replacement of distractions with other distractions: A propensity-based approach to estimating realistic crash odds ratios for driver engagement in secondary tasks. *Transportation research part F: traffic psychology and behaviour*, 63, 186-192.

