

# Comparing Causal Search Analyses with Traditional Statistical Methods

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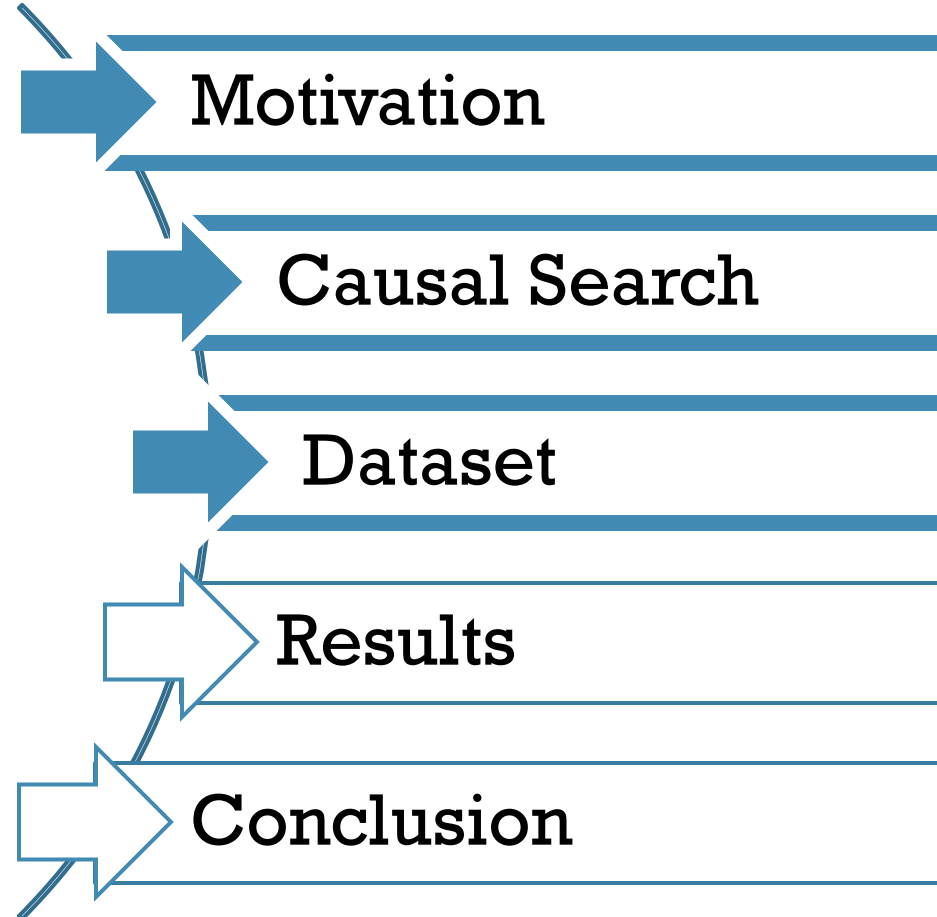
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**\* This presentation and paper is a follow-up to the paper “Preliminary Causal Discovery Results with Software Effort Estimation Data” presented at Innovations in Software Engineering Conference (ISEC) 2018**

# Outline

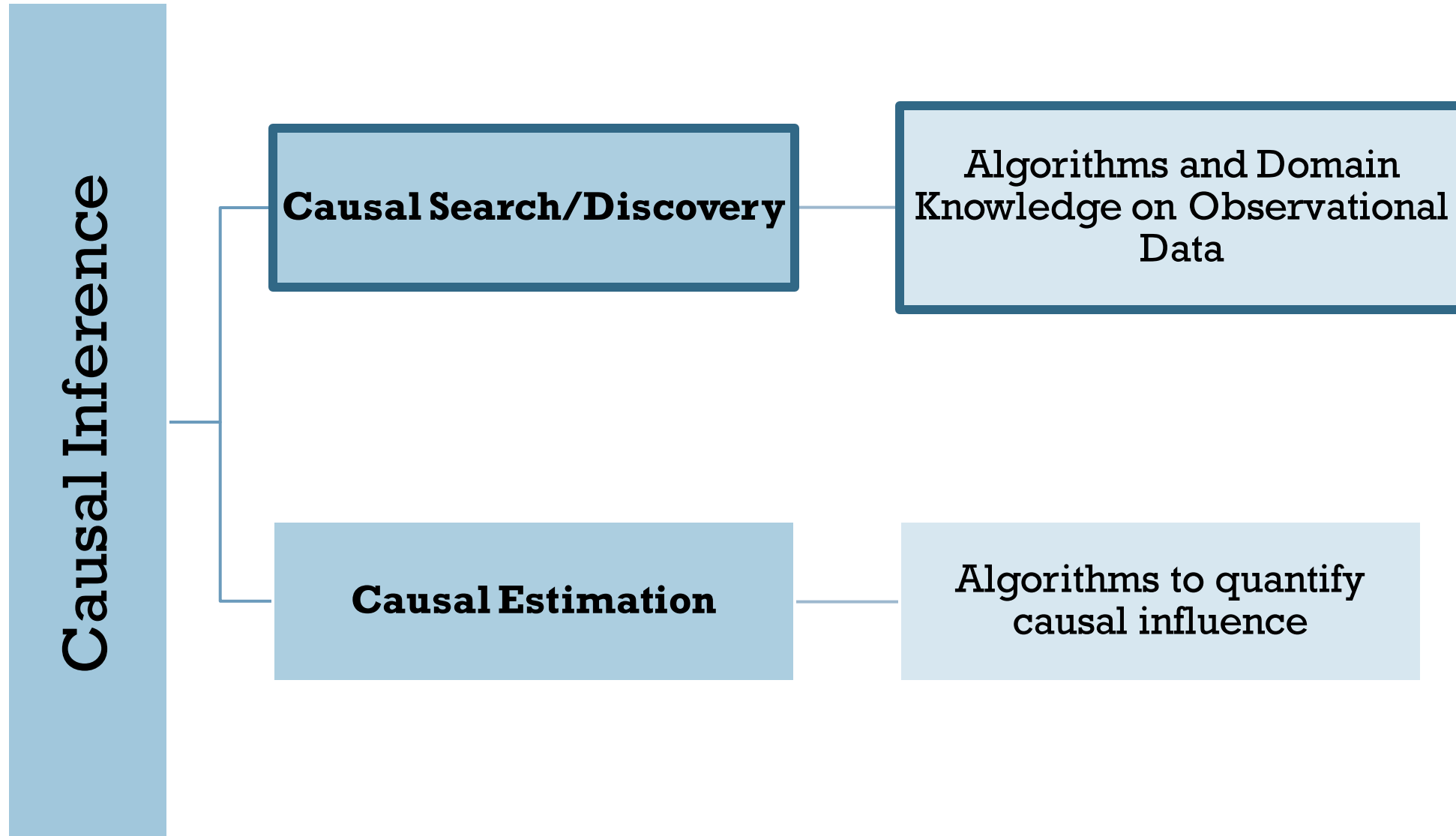


## COCOMO<sup>®</sup> II

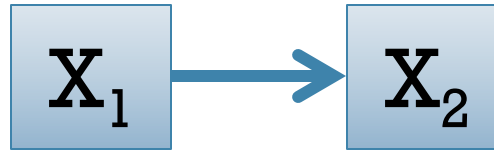
### Parametric Cost Models

$$Effort = 2.94 \times Size^E \times \prod_{i=1}^{17} EM_i$$

- ◆ Input: size, product and personnel attributes
- ◆ Effort in Person-Months (PM)
- ◆ Size in KSLOC (1000 SLOC)
- ◆ Domain Experts
- ◆ Data calibration
- ◆ No causal analysis



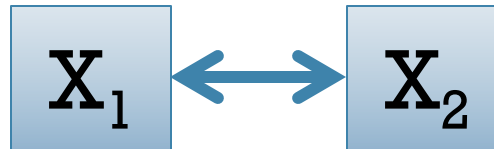
Causal Search  
Algorithm  
Results



Change in  $X_1$  causes  
change in  $X_2$



Insufficient information  
to select orientation



May be common  
confounder of both  
variables, missing from  
dataset

# Causal Search Algorithms

## PC SEARCH

- **Constraint-based algorithm**
- First developed by Peter Spirtes and Clark Glymour
- Highly accurate and scalable

## PC STABLE

- **Constraint-based algorithm**
- Variant of PC, addressing problem that output depends on order of variables in dataset

## FGES (FAST GREEDY EQUIVALENT SEARCH)

- **Score-based algorithm**
- Reduces bi-directed edges in output

## FASK (FAST ADJACENCY SKEWNESS)

- **Score-based algorithm**
- Makes use of asymmetry in univariate distributions to determine edge orientations

## Dataset: Unified Code Count (UCC)

### PROJECT DESCRIPTION

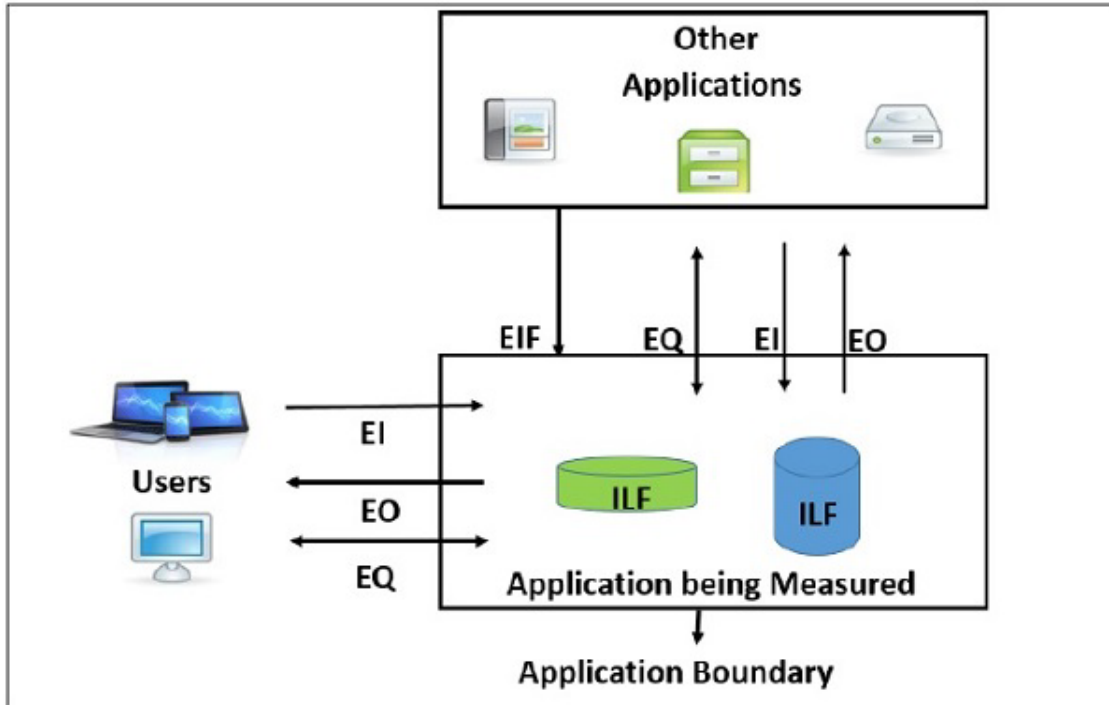
- ◆ Code metrics tool (logical SLOC, cyclomatic complexity)
- ◆ Maintained at USC
- ◆ UCC is released to users across world, primarily used in U.S. Aerospace industry
- ◆ Recommended for SLOC-based size input to Software Resources Data Report (SRDR)
- ◆ Implemented in C++
- ◆ Modularized architecture
- ◆ Tasks started and finished between 2010 to 2014

## Dataset Attributes

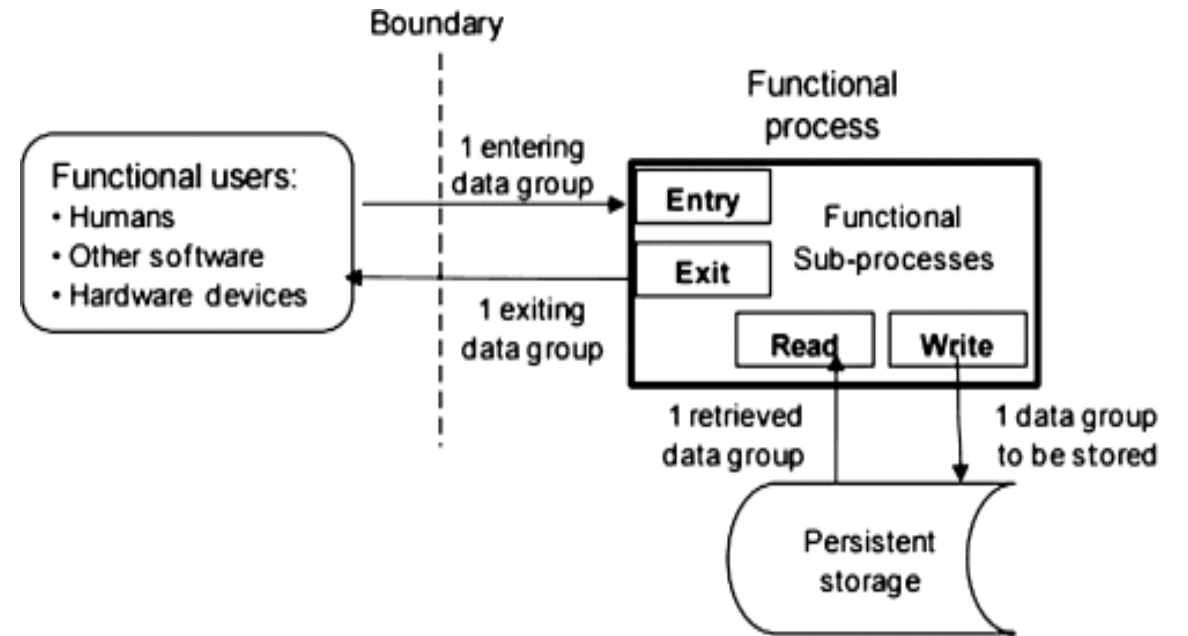
1. Equivalent SLOC (ESLOC)
2. IFPUG Function Points (FPs)
3. IFPUG Software Non-functional Assessment Process (SNAP)
4. COSMIC Function Points (CFPs)
5. Total Effort
6. Applications Experience (APEX)
7. Platform Experience (PLEX)
8. Use of Software Tools (TOOL)
9. Personnel Continuity (PCON)
10. Product Complexity (CPLX)
11. Analyst Capability (ACAP)
12. Programmer Capability (PCAP)
13. Documentation Match to Lifecycle Needs (DOCU)



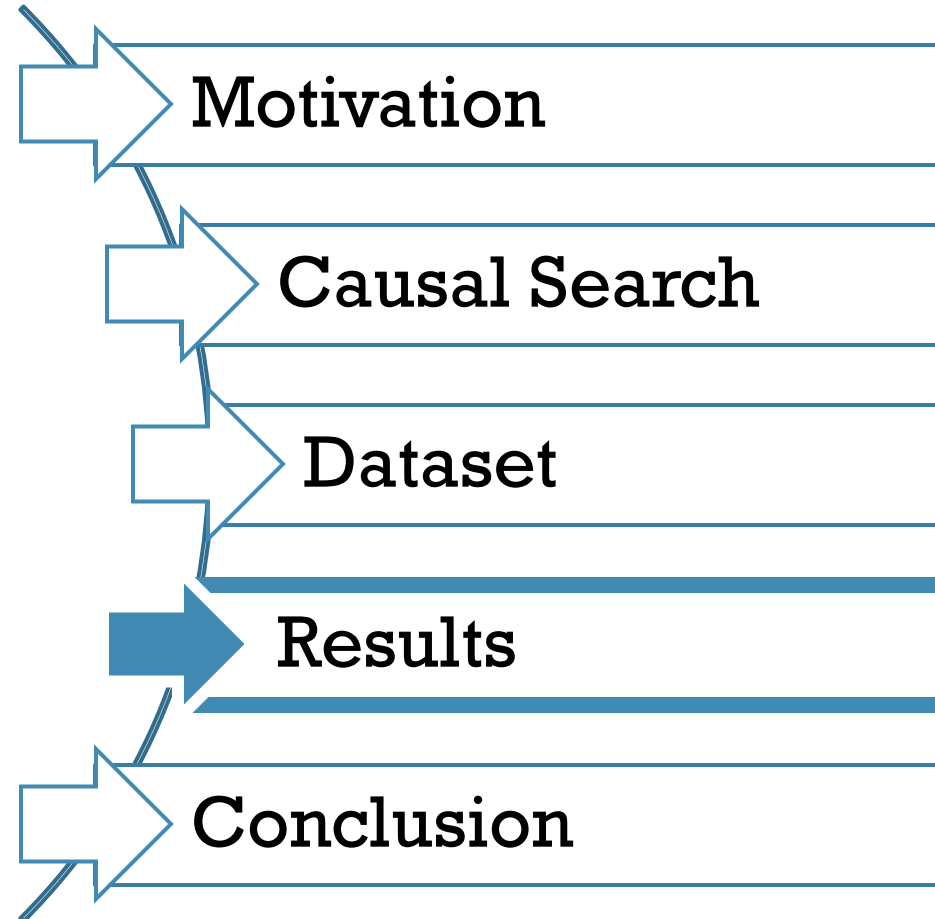
## IFPUG



## COSMIC



# Outline



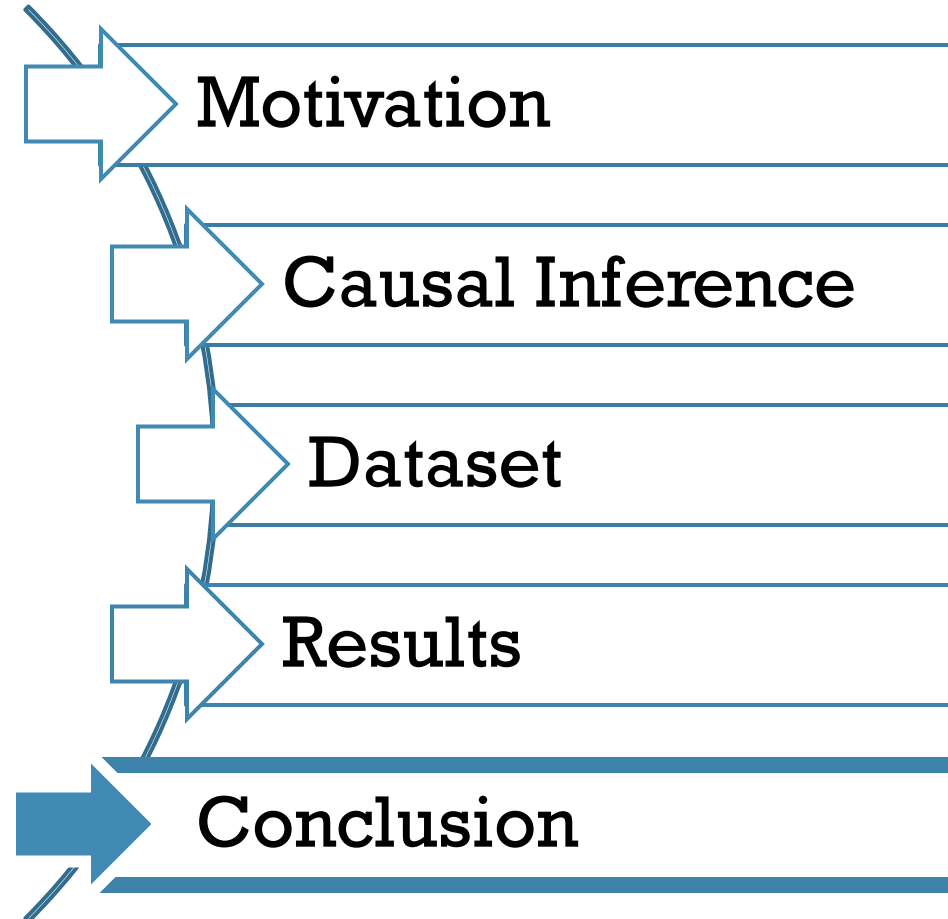
Algorithm	Direct Cause of Total Effort							
	ESLOC	FP	SNAP	CFP	CPLX	ACAP	PCAP	DOCU
PC All Size Metrics				YES			YES	
PC ESLOC							YES	
PC FP							YES	
PC SNAP			YES				YES	
PC CFP				YES			YES	
PC-Stable All Size Metrics				YES				
PC-Stable ESLOC					YES			
PC-Stable FP					YES			
PC-Stable SNAP					YES			
PC-Stable CFP				YES				
FGES All Size Metrics				YES			YES	
FGES ESLOC							YES	
FGES FP							YES	
FGES SNAP							YES	
FGES CFP				YES			YES	
FASK All Size Metrics		YES	YES	YES				YES
FASK ESLOC								YES
FASK FP		YES						YES
FASK SNAP			YES					YES
FASK CFP				YES				YES

**Note: Black cells are Not Applicable as factor not present in model**

# Multiple Regression

<b>Vars</b>	<b>R-Sq</b>	<b>R-Sq (adj)</b>	<b>R-Sq (pred)</b>	<b>CFPs</b>	<b>ESLOC</b>	<b>FPs</b>	<b>SNAP</b>	<b>ACAP</b>	<b>PCAP</b>	<b>DOCU</b>	<b>CPLX</b>
1	57.6	56.1	51.0						X		
1	54.6	53.0	47.9	X							
2	84.9	83.8	78.3	X					X		
2	78.5	76.9	67.4				X		X		
3	85.3	83.6	76.1	X					X		X
3	85.1	83.4	77.2	X		X			X		
4	85.4	83.0	73.0	X			X		X		X
4	85.3	83.0	68.6	X				X	X		X
5	85.4	82.3	61.9	X			X	X	X		X
5	85.4	82.3	67.0	X	X		X		X		X
6	85.4	81.6	55.2	X	X		X	X	X		X
6	85.4	81.6	59.9	X			X	X	X	X	X
7	85.4	80.8	53.0	X	X		X	X	X	X	X
7	85.4	80.8	52.0	X	X	X	X	X	X		X
8	85.4	79.8	49.5	X	X	X	X	X	X	X	X

# Outline



# Threats to Validity

- **Small Sample Size**
  - Conditional independence testing can only detect strong correlational effects
  - Resulting in sparsely-connected graphs
- **Dataset of convenience – UCC**
  - Maintained by Masters students
  - UCC is released to users across world, primarily used in U.S. Aerospace industry
  - Continuing study with various datasets

# Conclusions

## SUMMARY OF RESULTS

- **Consistent Results**
  - CFP – Total Effort
  - SNAP – Total Effort
  - PCAP – Total Effort

## FUTURE WORK

- Run similar analyses on larger and varied cost estimation datasets
- Especially focus on impacts to acquisition cost and schedule
- Datasets including various source code metrics

# Acquisition Research Implications

- Expensive, prohibitive experiments of acquisition factors could be obviated by use of causal methods
- Revisit effort factors, potentially reducing number of required dataset characteristics
- Acquisition researchers could integrate causal conclusions for holistic model
- Identify and prioritize research funding towards causal research outcomes worthy of investment in repeatability and reproducibility studies
- Causal research findings more confidently tested by acquisition program interventions with less risk of waste



Thank you for your  
time!