FloCon 2022

18th Annual Open Forum for Large-Scale Data Analytics **Using Data to Defend**

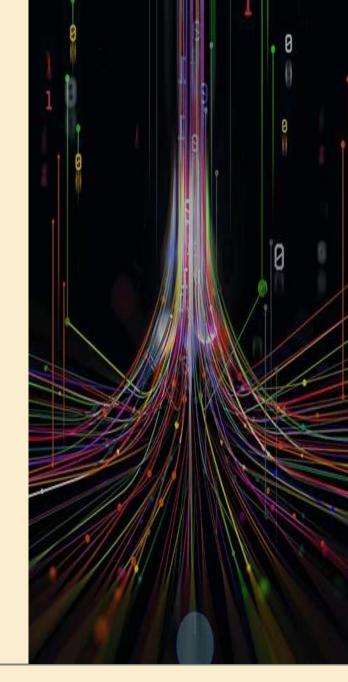
What Do We Mean by a Science of Security?

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Carnegie Mellon University Software Engineering Institute FloCon presentations: Cybersecurity Data Science (CSDS)

- 1. 2019 CSDS Field interview insights
- 2. 2020 CSDS Best Practices research results
- **3. 2021** CSDS Trends *emerging methods*
- **4. 2022** Security Science *future directions*



Cybersecurity Data Science

I. <u>Foundations</u>

- What is CSDS?
- Status as a profession?
- II. <u>Interviews</u> 50 CSDS practitioners
- III. Best practices
- Data management
- Scientific methods
- Cross-domain collaboration

Scott Mongeau Andrzej Hajdasinski

Cybersecurity Data Science

Best Practices in an Emerging Profession

Foreword by Timothy Shimeall

Springer

Cybersecurity Data Science: Springer 2021



Data Science offers both solutions and challenges

Carnegie Mellon University Software Engineering Institute

Engineering Analysis Theory

Data Science offers both solutions and challenges

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SOLLOGIER

TRYING TO GONVINCE PEOPLE THAT BOOKS ARE INTERSTING

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Sciency... things



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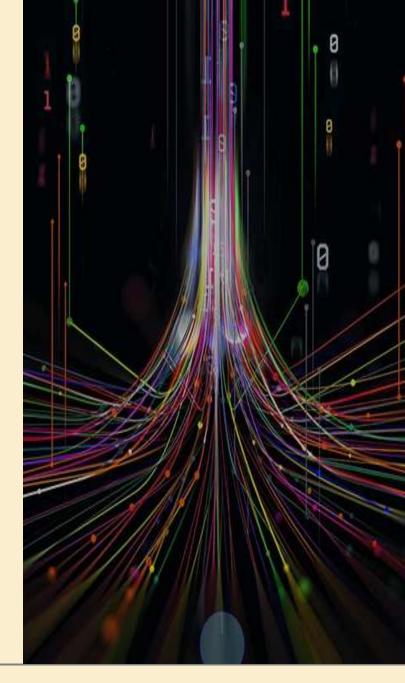
Talking about science...

without offending anybody.

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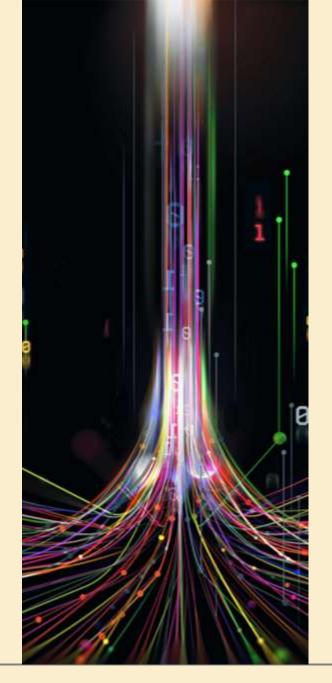
Agenda

- 1. Science of Security?
- 2. Engineering versus Science
- 3. Science as a Process
- 4. Operational Security Science
- 5. Where do we go from here?



What Do We Mean by a Science of Security?

1. Science of Security



MobileSaaSCloudBYODCyber Security TeamMicroservicesVMs

Whitelisted 3rd party vendors

SOC

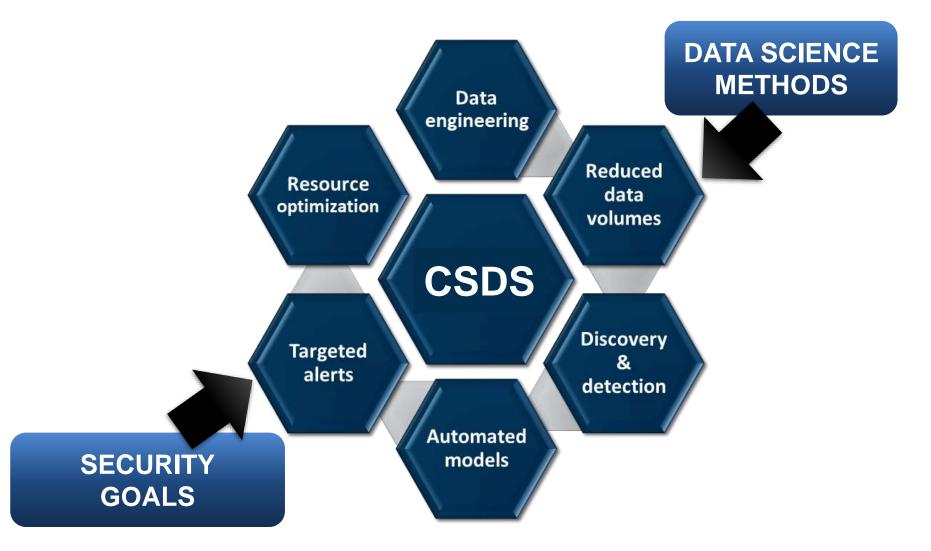
The "internet"

Firewall

Insider threat

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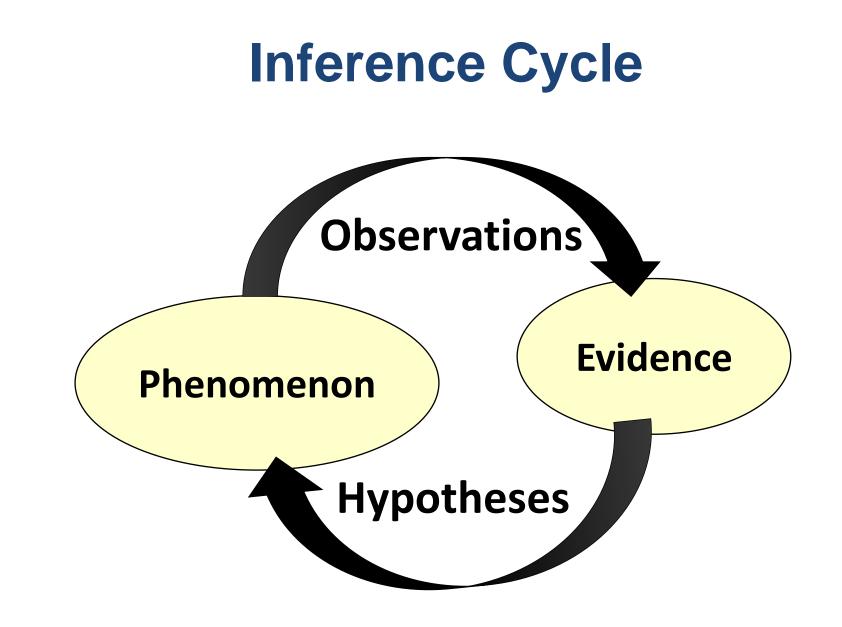
Framing CSDS



Limits of engineering-based prescriptions



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Science in security practice



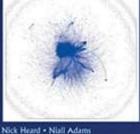




SCIENCE OF SECURITY

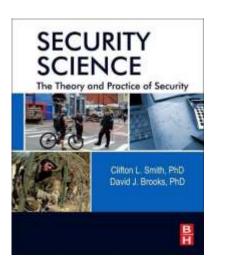


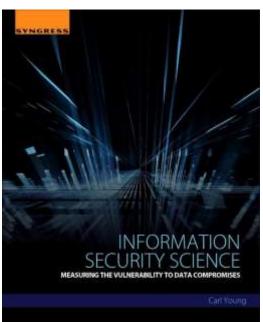




Patrick Robin-Delanchy • Melissa Turcotte

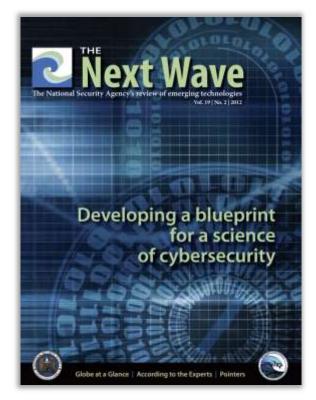
World Scientific





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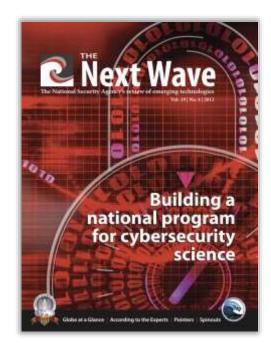
Science in security practice

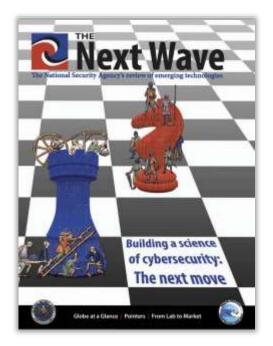


https://www.nsa.gov/Research/The-Next-Wave/

Vol. 19, No. 2 2012

- Amla et al Toward a secure and trustworthy cyberspace
- Landwehr Cybersecurity: from engineering to science
- Longstaff <u>Barriers to achieving a science of cybersecurity</u>
- Maxion <u>Making experiments dependable</u>
- Meushaw, Landwehr NSA initiatives in cybersecurity science
- Pavlovic <u>On bugs and elephants: mining for science of security</u>
- Schneider Blueprint for a science of cybersecurity
- Shostack The evolution of information security

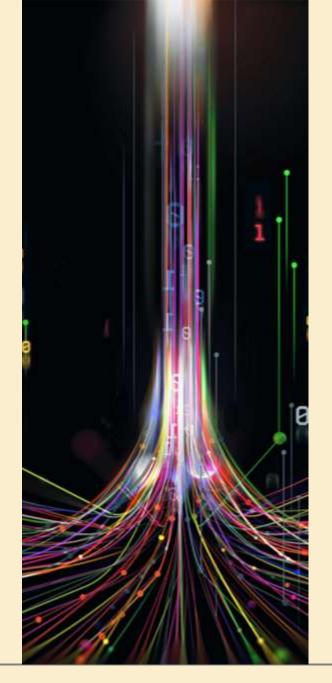




AND PRIVACY

What Do We Mean by a Science of Security?

2. Engineering versus Science



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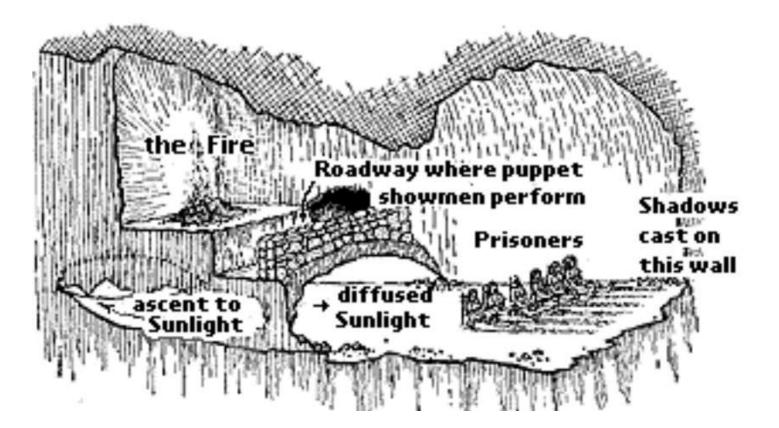
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Plato's Allegory of the Cave



Attribution https://www.philosophyzer.com/the-allegory-of-the-cave-by-plato-summary-and-meaning/

Plato's Allegory of the Cave



Simarro, Francisco Montero and Víctor López-Jaquero. "<u>From Plato's Dualism to user Interface Development</u>." WEBIST (2007).

Medical practice: clinical vs. research

Clinical practice

- Intuition based on experience
- Heuristics
- Processes and procedures
- Intervention on conditions
- Extrapolate from symptoms
- Applications of technology
- Rough consensus



Medical research

- Statistics
- Cumulative theory
- Protocols and rigor
- Address underlying causes
- Substantiate supposed causes

Clinical versus Actuarial Judgement

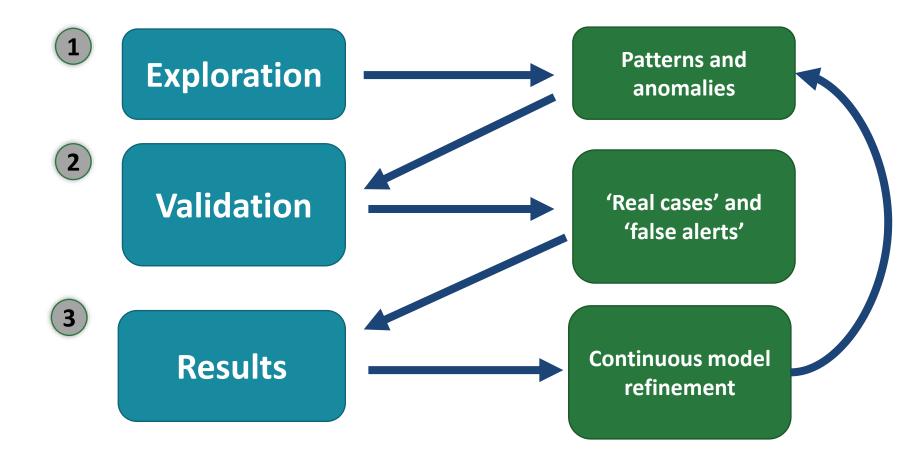
- Clinical: ad hoc field experience
- Actuarial: empirical statistical observations and testing
- When data is unavailable, operational judgment defers to clinical instincts



Dawes, R., Faust, D., & Meehl, P. (1989) <u>Clinical versus actuarial judgement</u>. Science 243:6

Research comparing these two approaches shows the actuarial method to be superior. Factors underlying the greater accuracy of actuarial methods, sources of resistance to the scientific findings, and the benefits of increased reliance on actuarial approaches are discussed.

Continuous Improvement Process



What Do We Mean by a Science of Security?

3. Science as a Process



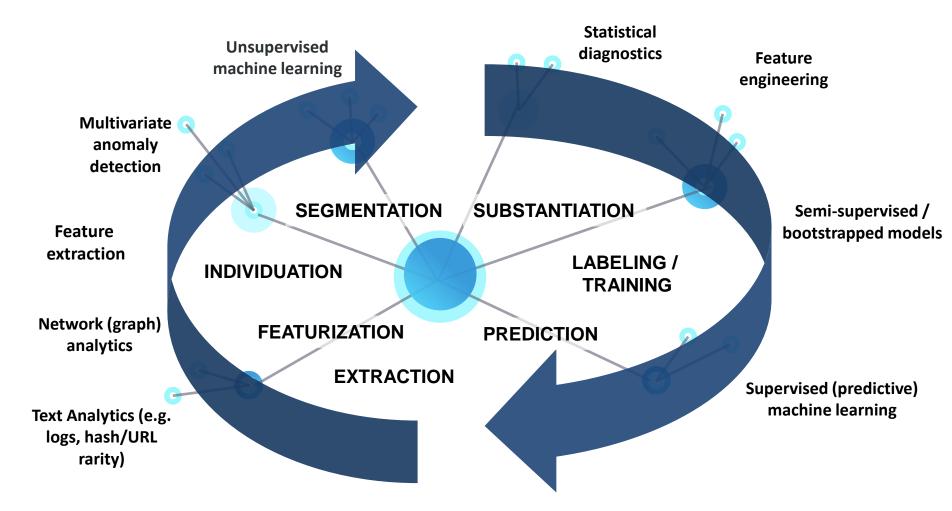
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CSDS Process: Linking Methods

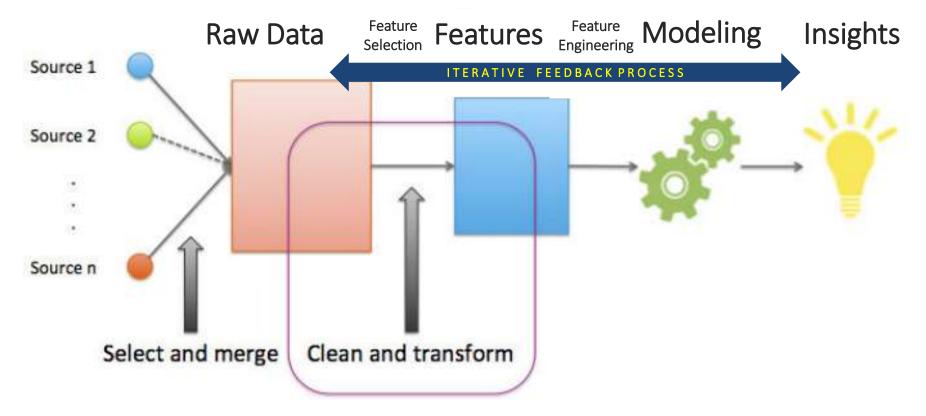


We are already doing science, albeit in bits and pieces...

Scientific processes

Data Data management CSDS CSDS CSDS CSDS CSDS Main Constant Cross-domain collaboration

Data Management: Analysis + Features



SOURCE: Alice Zheng, Amanda Casari. 2016. Feature Engineering for Machine Learning Models. O'Reilly Media.

CSDS: High-Level Functional Process



What Do We Mean by a Science of Security?

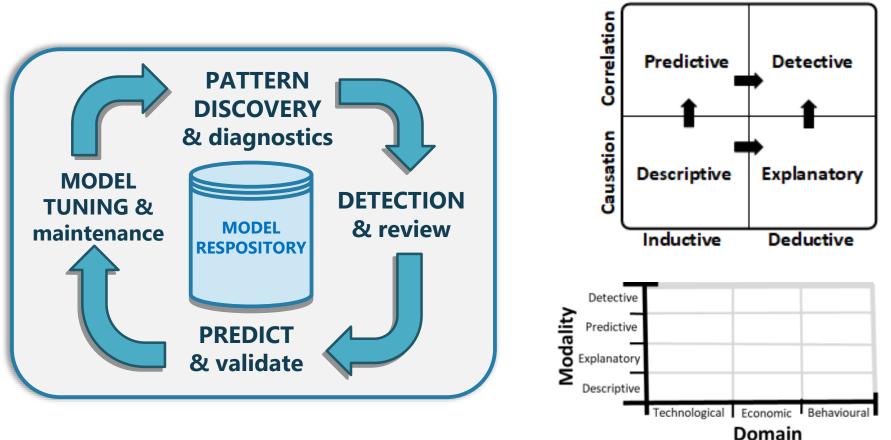
4. Operational Security Science



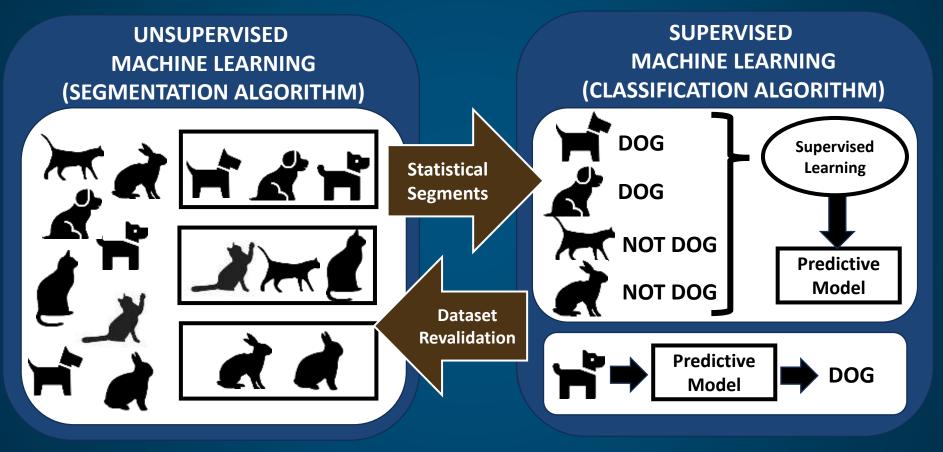
CSDS as a Process: Discovery and Detection



Cyber Analytics: Staged Discovery Process



Unsupervised Category Extraction => Targets



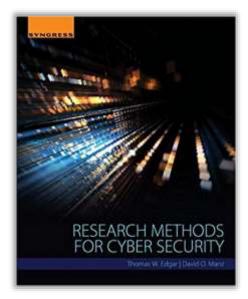
Research Methods for Cybersecurity

• Experimental

i.e. hypothetical-deductive and quasi-experimental

• Applied

- i.e. applied experiments and observational studies
- Mathematical
 - i.e. theoretical and simulation-based
- Observational
 - > i.e. exploratory, descriptive, machine learning-based



Manz, D. and Edgar, T. (2017) Research Methods for Cyber Security

Labels: What constitutes 'evidence'?

	Inductive	Deductive
Syntn	 Simulations Laboratory 	- Thought experiments
oyntnesized	- Red Teaming	- Expert opinion
ןכ	sourced	intelligence
Collected	- Field evidence - Probing & testing - 3 rd party	- Rules & signatures - Research & threat

EXAMPLES OF SECURITY EVIDENCE

1. Field evidence (e.g. observed incidents) 2. Sourcing own data from field testing (e.g. local experiments) 3. Honeypots 4. IDSs (Intrusion Detection Systems) 5. Simulation findings 6. Laboratory testing (e.g. malware in a staged environment) 7. Stepwise discovery (iterative interventions) 8. Pen testing (attempts to penetrate the network) 9. Red teaming (staged attacks to achieve particular goals) 10. Incidents (records associated with confirmed incidents) 11. Reinforcement learning (self-improving ML to achieve a goal) 12. Research examples (datasets recording attacks from research) 13. Expert review (opinion and guidance from experts) 14. Intelligence feed (indications from a 3rd party service) 15. Thought experiments (e.g. boundary conditions, counterfactuals)

CSDS Methods & Approaches

METHODS

Deep learning – computer vision & acoustics

Network graph analytics

Natural language, semantic & knowledge engineering

Forecasting and time series analysis

NOVEL METHODS

- Pattern recognition
- Classification
- Feature extrapolation
- Diagnostics
- Multivariate inferential statistics

DISCIPLINES

Fraud analytics, forensics & criminology

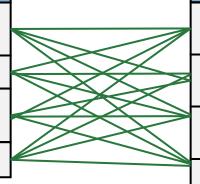
Medical, epidemiological, ecological, bioinformatics

Social and behavioral sciences, including (micro-) economics & game theory

Risk management

NOVEL CONTEXTS

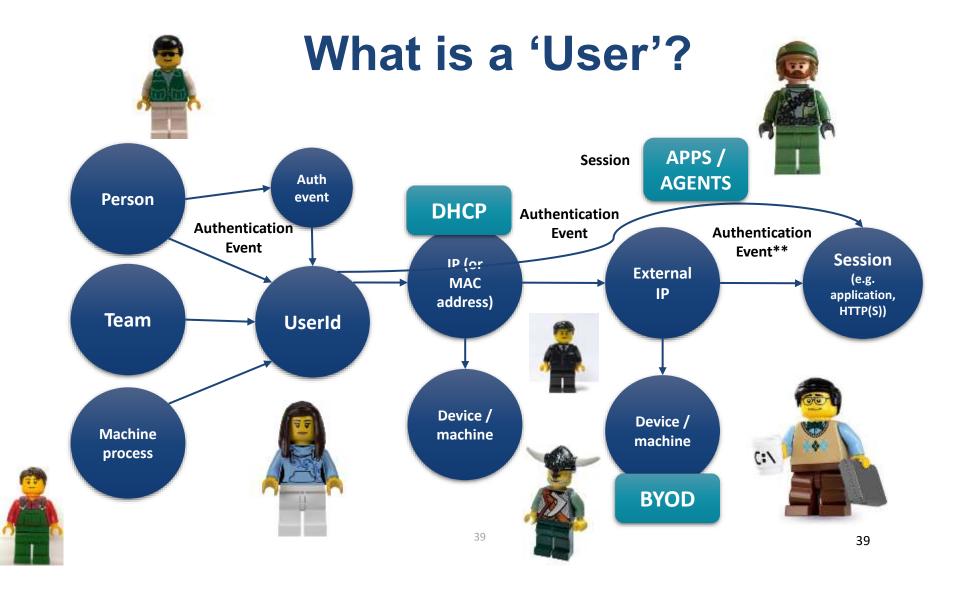
- Complex multi-domain models
- Technological
- Economic
- Behavioral
- Techno-economic-behavioral context
- Deterministic + latent variables



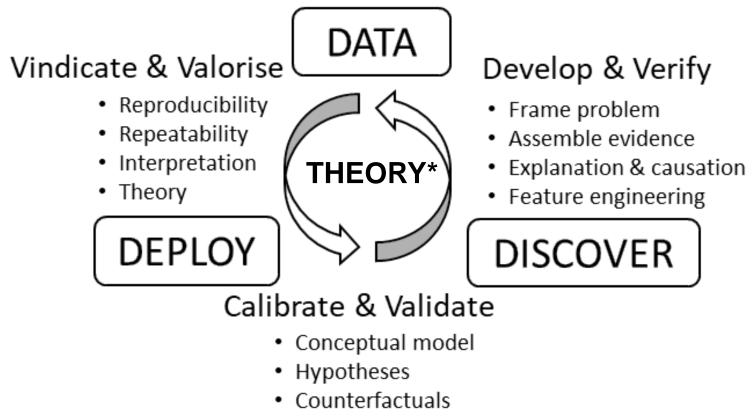
What Do We Mean by a Science of Security?

5. Where do we go from here?





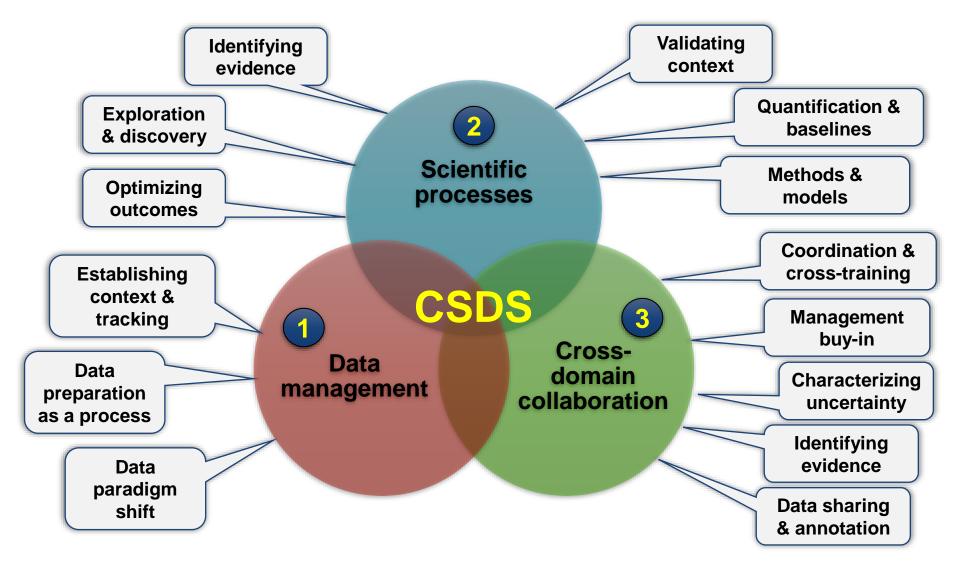
CSDS as a Unified Process



Falsification

* As cumulative, validated, socialized knowledge

CSDS: A Roadmap for Rigor



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END

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APPENDIX

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Scientific methods in cybersecurity

Cybersecurity Data Science (CSDS) Corpus

April 12, 2020

Advocacy, Best practices, Management, Methods, Research, Theory

OVERVIEW

For those interested in the rapidly emerging field of cybersecurity data science (CSDS), below



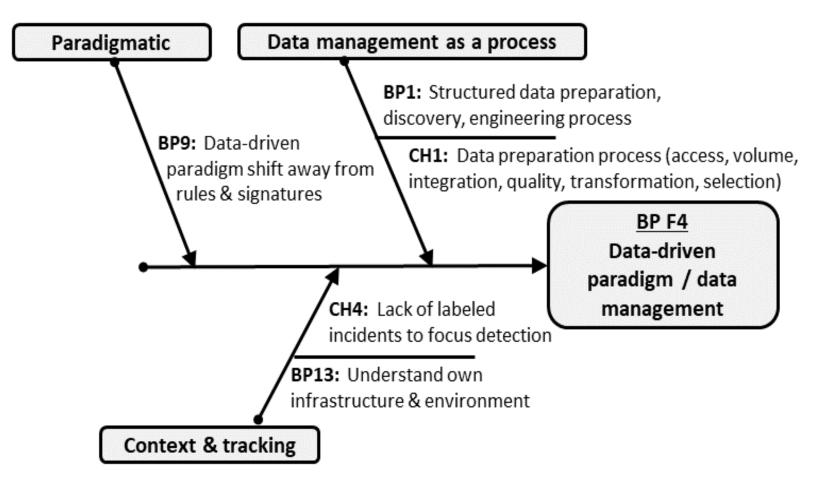
is a corpus of 33 book-length works. The list covers publications going back to 2001, although two-thirds of the works (22 out of 33) were published in the last five

years (2016 to 2020).

<u>CSDS Corpus</u>

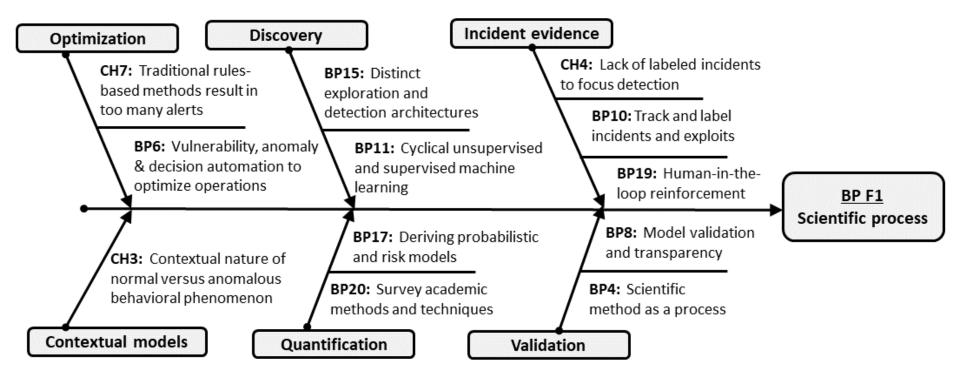
воок	AUTHORS
Data-Driven Security	Jacobs & Rudis, 2014
Fraud Analytics Using Descriptive, Predictive, and Social Network Techniques	Baesens et al., 2015
Essential Cybersecurity Science	Dykstra, 2016
Dynamic Networks and Cyber-Security	Adams & Heard (Eds.), 2016
How to Measure Anything in Cybersecurity Risk	Hubbard & Seiersen, 2016
Research Methods for Cyber Security	Edgar & Manz, 2017
Big Data Analytics in Cybersecurity	Savas & Deng (Eds.), 2017
Cybersecurity Analytics	Verma & Marchette, 2020

CSDS Best Practice I: Data Management



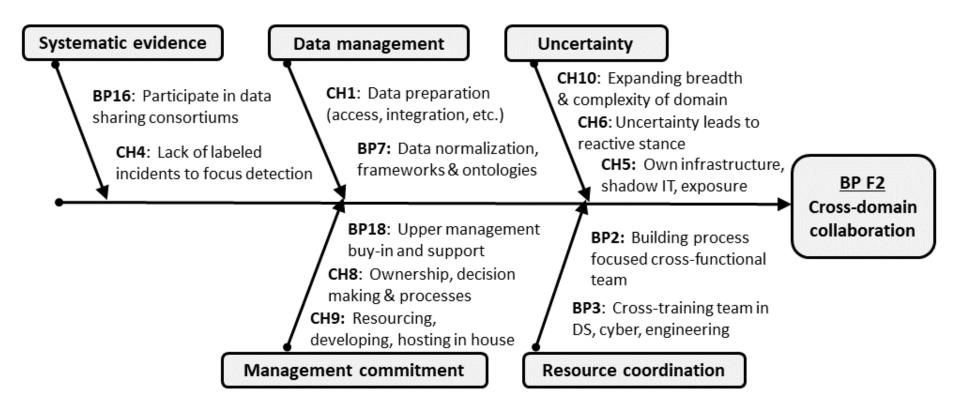
Mongeau 2021 'Cybersecurity Data Science' Springer

CSDS Best Practice II: Scientific Process



Mongeau 2021 'Cybersecurity Data Science' Springer

CSDS Best Practice III: Cross-Domain Collaboration



Mongeau 2021 'Cybersecurity Data Science' Springer

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