



QUANTIFYING CYBER RISKS

MARIEKE KLAVER, PETER LANGENKAMP (TNO)

SANDER ZEIJLEMAKER (DISEM INSTITUTE)

› INTRODUCTION

› Dr. Marieke Klaver

- › (PhD in mathematics)
- › Researcher at TNO
- › Focus areas: risk management, cyber security, critical infrastructures

› Dr. ir. Peter Langenkamp

- › (PhD in Physics)
- › Cyber Security Researcher, dept. Cyber Security and Robustness @ TNO
- › Focus: Risk analysis, secure multi-party computation, self-sovereign identity

› Sander Zeijlemaker RA RE MSc CISA CISM SCF

- › Managing Director @ Disem Institute
- › PhD researcher Radboud University (system dynamics and security economics)
- › Focus areas: cyber security, strategic assurance & optimalization, and executive coaching

› CONTENTS

- 01. INTRODUCTION QUANTIFYING CYBER RISKS
- 02. METHODOLOGY APPLIED WITH CASE STUDIES
- 03. HANDLING DYNAMICS IN CYBER RISKS
- 04. CONCLUSIONS



INTRODUCTION QUANTIFYING CYBER RISKS

› **QUANTIFYING CYBER RISKS**

BACKGROUND

Research project by TNO and NCSC

› Objective

- › To develop methods and tools for quantifying cyber risks
- › Based on case studies in critical sectors

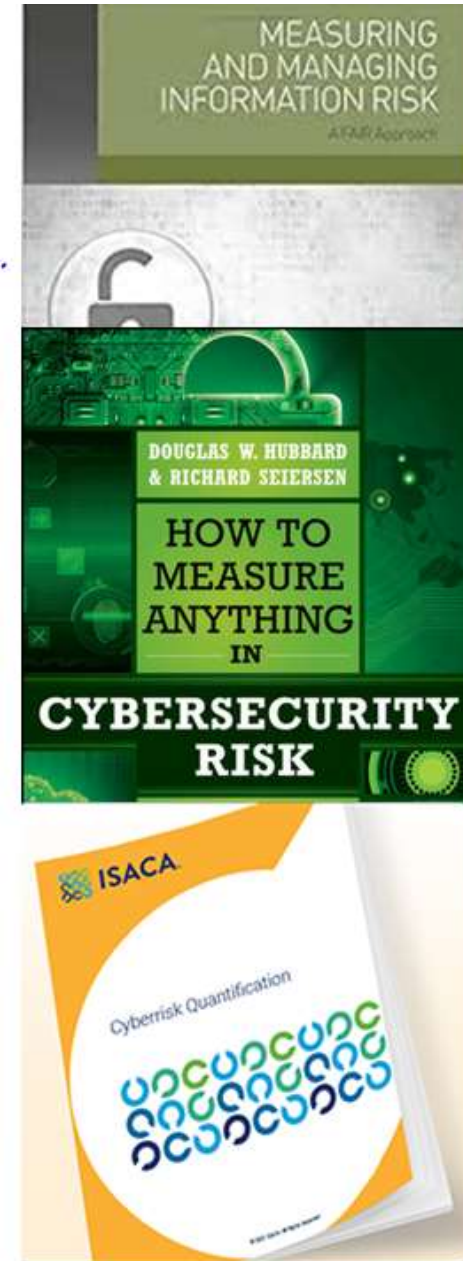
› Quantifying cyber risks helps to:

- › Provide a basis for cybersecurity investment decisions
- › Prioritize measures when resources are limited (distinctiveness)
- › Make cyber risks more comparable to other business risks
- › Facilitate information sharing on cyber risks
- › Assess how the effects of risks or measures propagate through the system (sensitivity analysis)
- › Potentially automate (parts of) analyses

› DEVELOPMENTS QUANTIFYING CYBER RISKS

- › Methods for quantifying cyber risk
 - › Underpinning qualitative risk heat maps
 - › Quantifying likelihood and impact
 - › Several methods exist, however as of yet there is no commonly accepted standard available

- › Supported by modelling approaches
 - › Bayesian Belief Network
 - › System Dynamics



› **APPROACH FOR USE CASES**

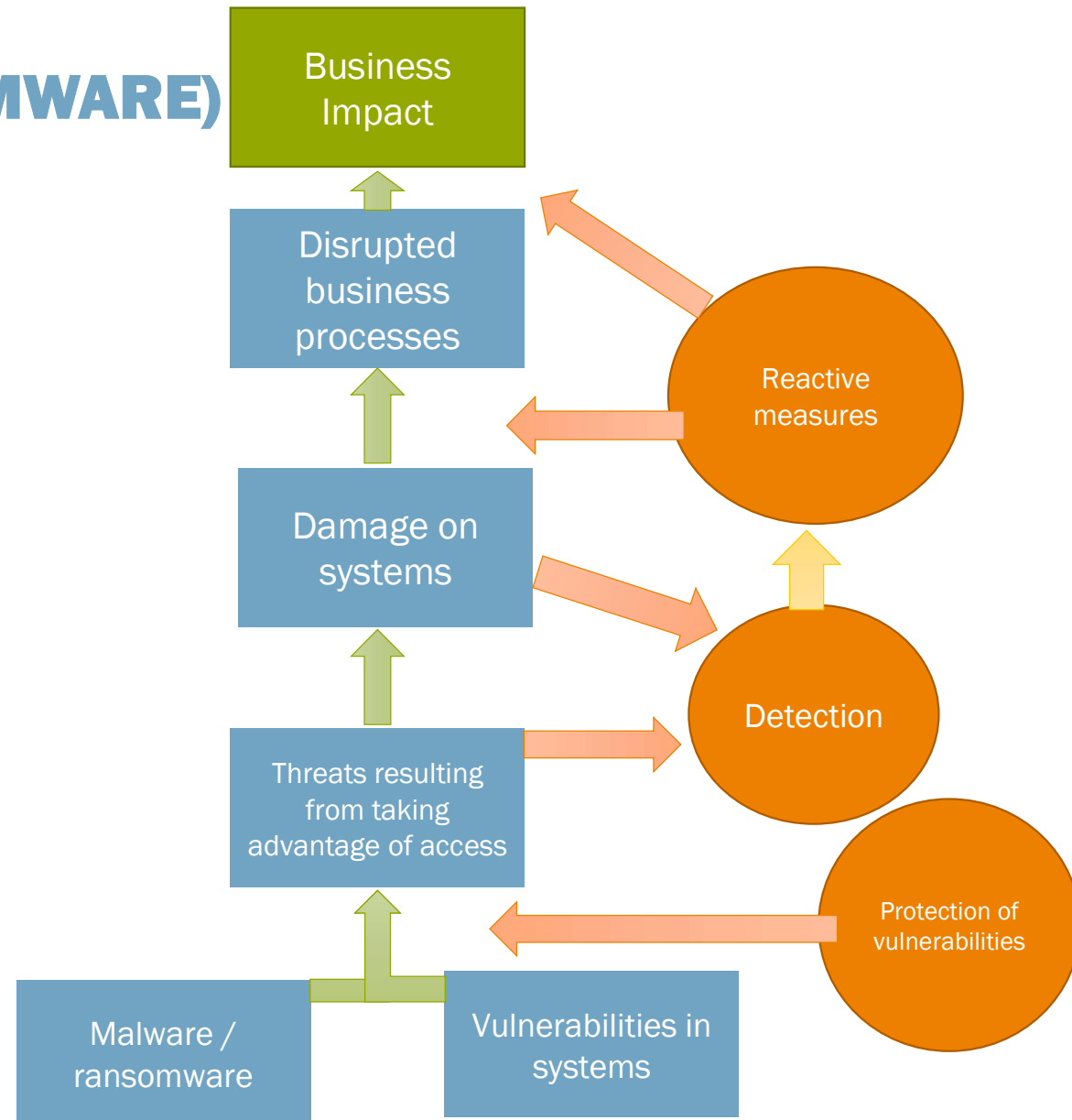
- › Developing a base model
 - › Based on a methodology developed for and with the financial sector
 - › Uses Bayesian Belief Network
- › Developing more detailed models in case studies
 - › Preparation
 - Select threats which will be applied to your use case (dreigingsscenario's)
 - Gather information on the cases' organisation (business processes, architecture, measures)
 - › Develop case specific model
 - › Workshops
 - assess and extend case model
 - risk assessments
 - › Present and discuss results with the case organisation

› BUILDING BLOCKS (RANSOMWARE)

Building block 1:
Data on the impact
of cyber incidents

Building block 3:
Measures and
systems

Building block 2:
Data on likelihood
and attack vectors



› BUILDING BLOCK 1: IMPACT ON BUSINESS

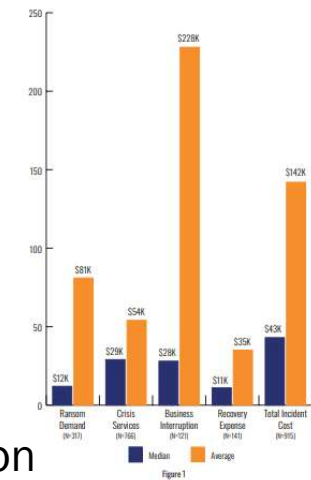
Different impact categories

- › Financial
 - › Crisis services
 - › Operational costs (down-time)
 - › Recovery services
 - › Ransom paid
 - › Costs of data breaches
 - › ...
- › Reputation
- › Safety
- › Environment
- › ...

Data sources:

- Incident reports
- Insurance data
- Security companies
- Open sources
- Sensor data
- Sector based information

NETDILIGENCE® CYBER CLAIMS STUDY
2021 RANSOMWARE SPOTLIGHT REPORT
Key Findings for 5-Year Period 2015–2019
SMEs – Costs for All Ransomware Claims

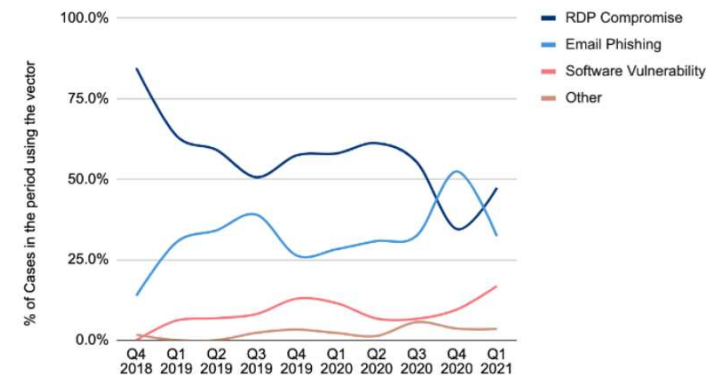


Gegevens VS 2020,
bron: taskforce ransomware

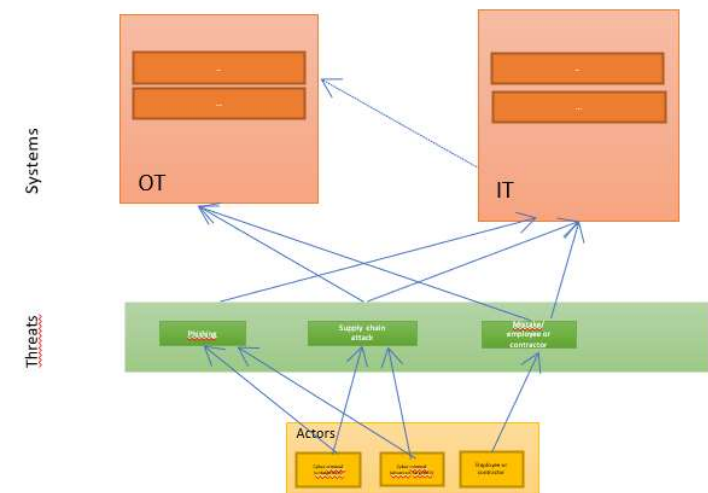
› BUILDING BLOCKS 2: THREAT SCENARIO / 3: SYSTEMS

- › Data on threats
 - › Attack vectors
 - › Data on frequency and trends

Ransomware Attack Vectors



- › Possible attack paths within an organisations own infrastructure
 - › OT and IT networks and systems
 - › (Digital) Architecture





METHODOLOGY APPLIED WITH CASE STUDIES

› INTRODUCTION QUANTIFYING METHODOLOGY

› In order to determine the quantitative risks in the case study, we applied a methodology which was developed in the Shared Research Programme Cybersecurity* a partnership between TNO, ABN AMRO, ING, Rabobank, de Volksbank and Achmea (see: <https://www.tno.nl/srpcybersecurity>).

- › Risk-based
- › Probabilistic model



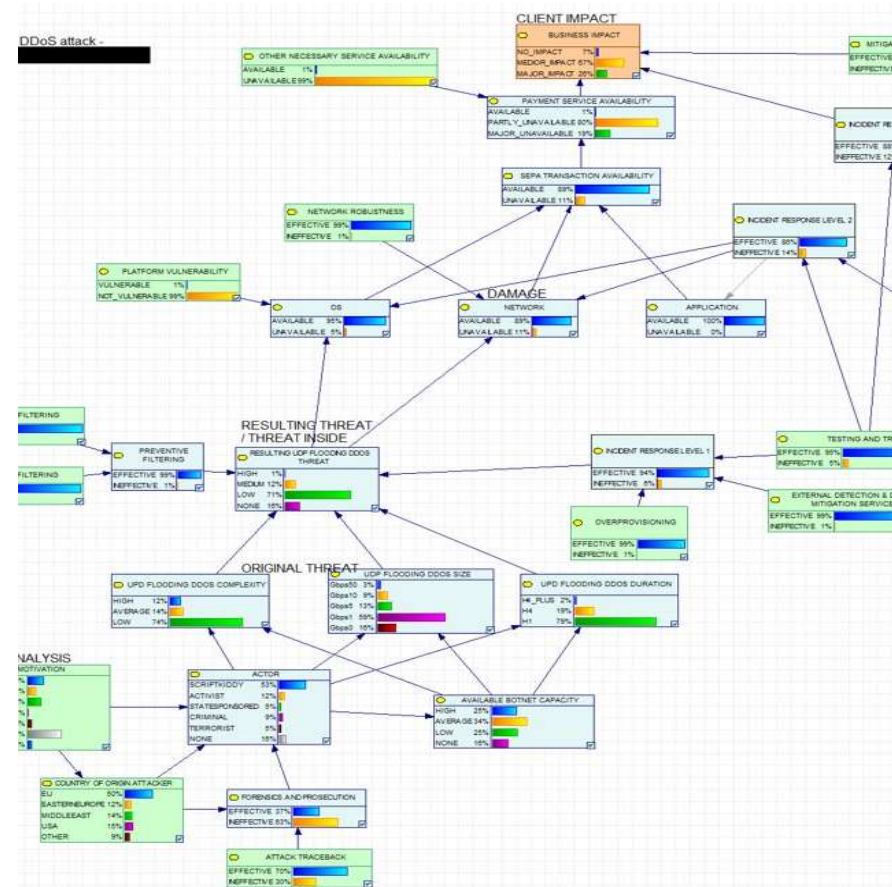
*This research programme was the predecessor of the current Partnership for Cyber Security Innovation (PCSI, zie <https://pcsi.nl/>)

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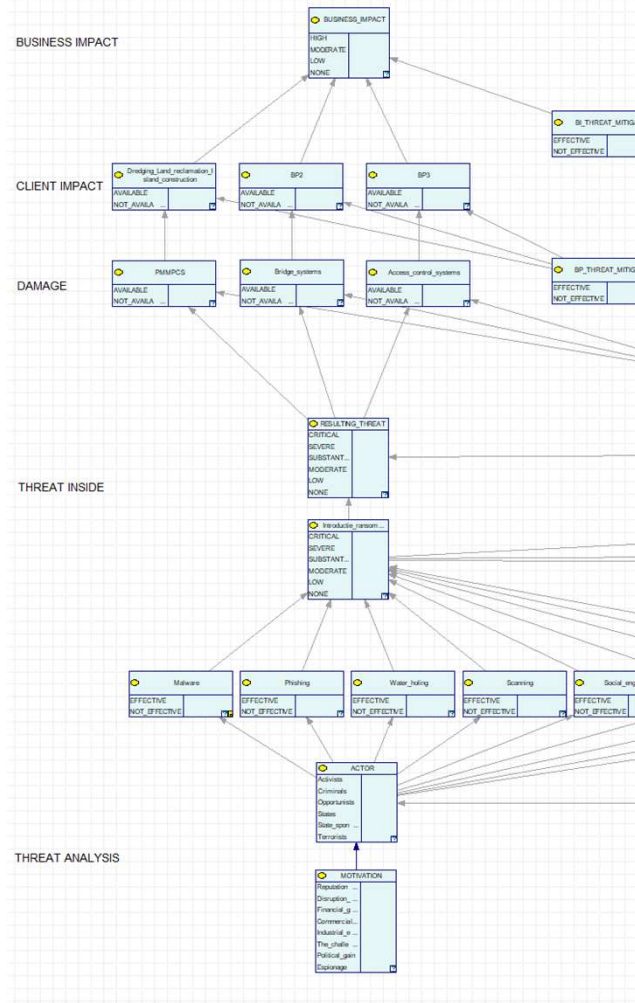
› From input (green) to business impact (orange)



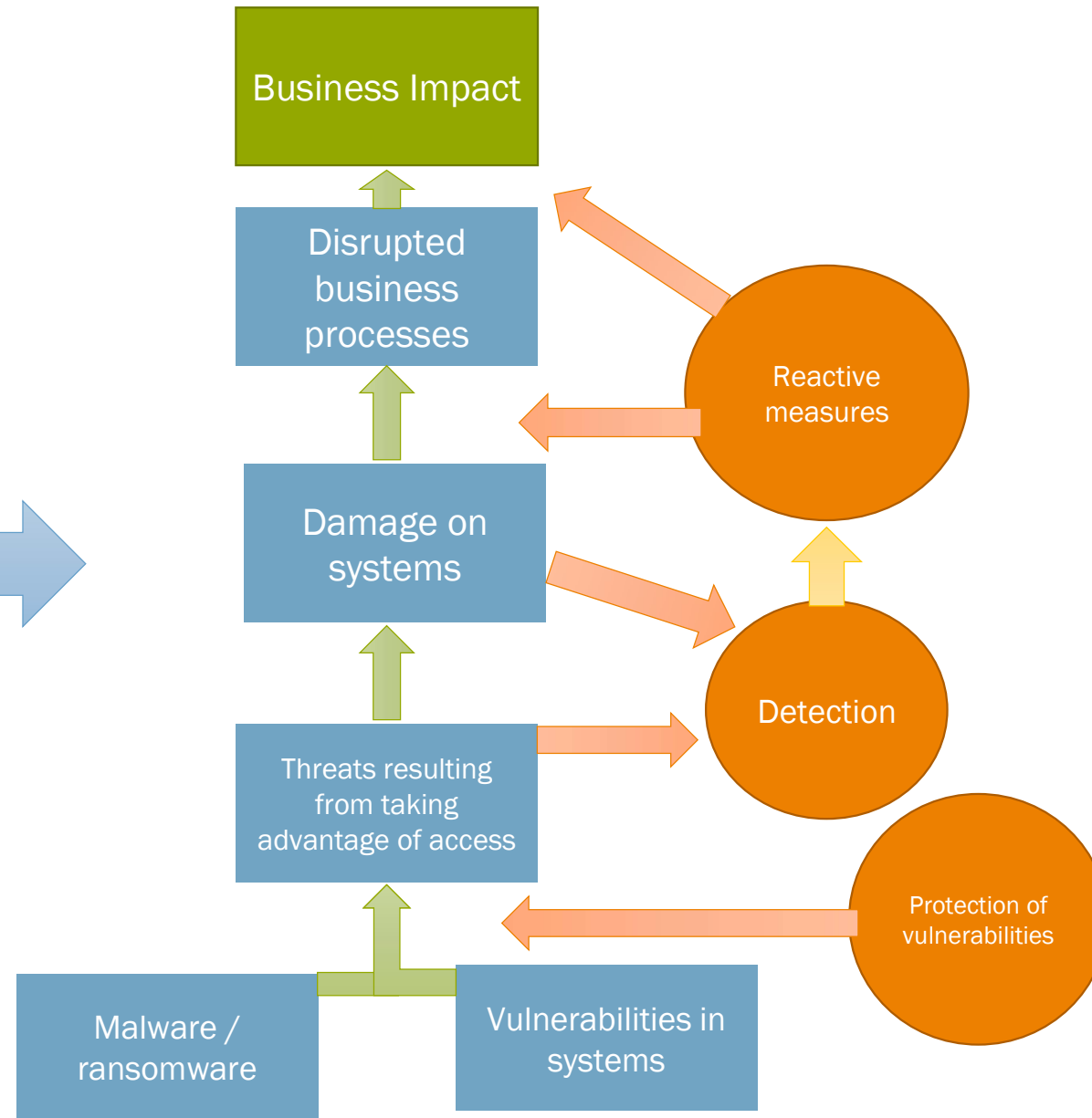
- 0 Identify relevant threats
- 1 Identify original threat and business impact to be modelled
- 2 Identify business processes and involved assets
- 3 Identify actor, motivation and means
- 4 Identify mitigating security measures
- 5 Build Bayesian Belief Network
- 6 Define probabilities with experts
- 7 Establish information sources
- 8 Develop script for automatic information feed
- 9 Put the model into operation

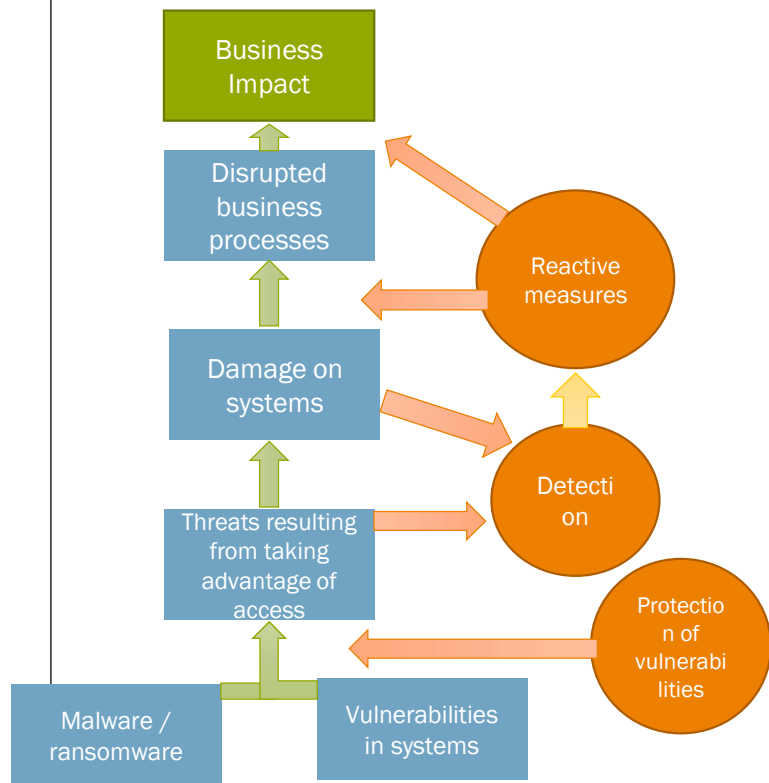
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FOCUS CASE STUDIES: RANSOMWARE

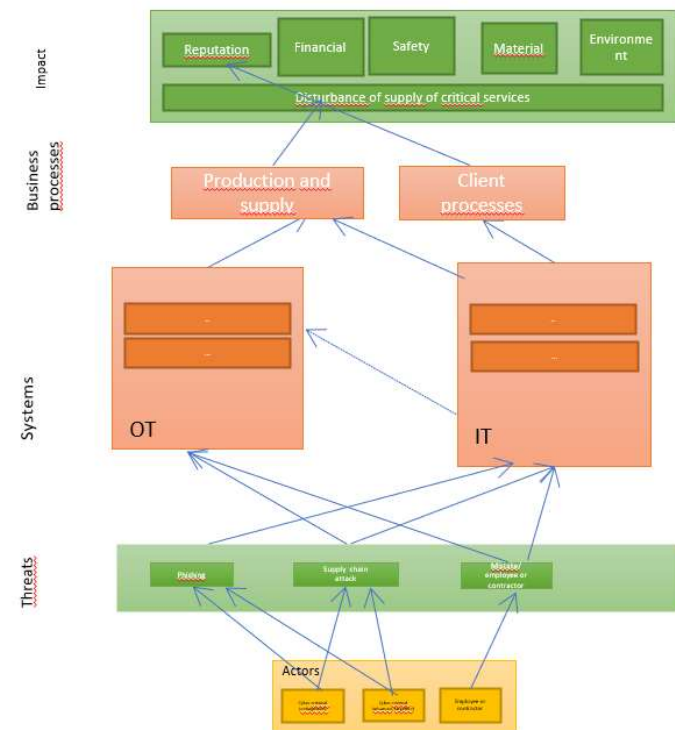
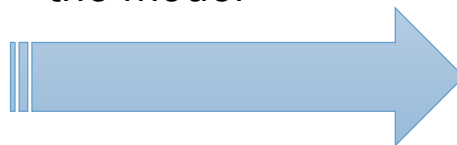


Interpretation
model for our
use case



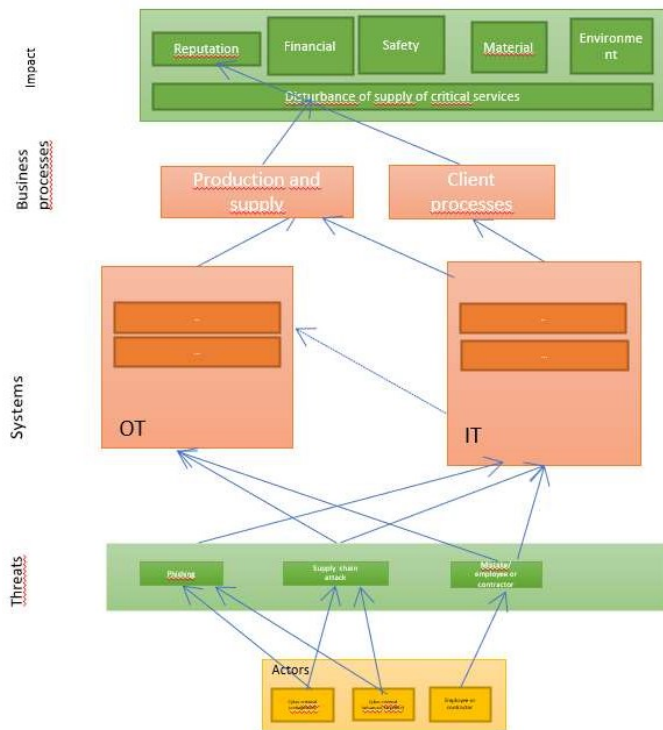


Together with experts, we refine the model

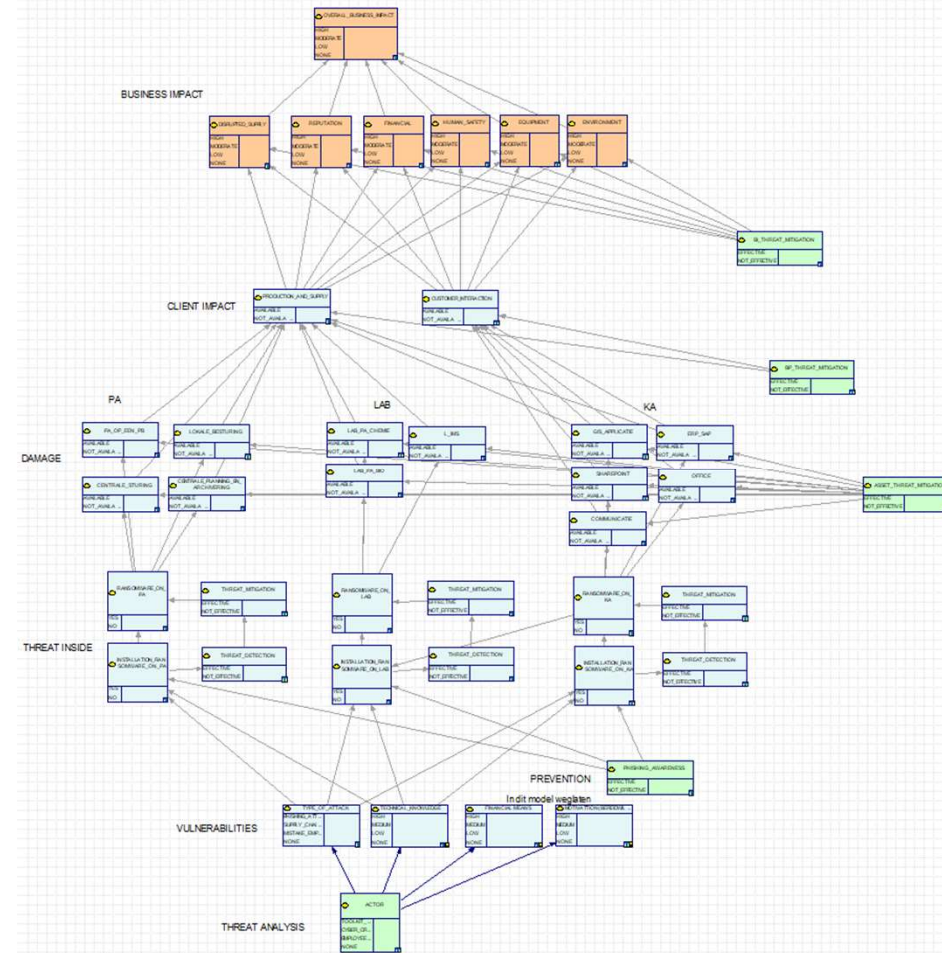


RESULTING MODEL

RESULT OF CASE STUDY: RANSOMWARE MODEL



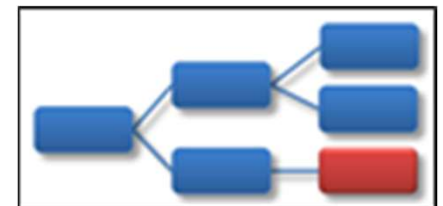
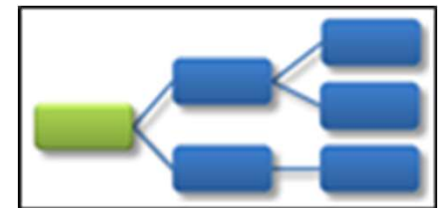
Turn the model into a Bayesian Belief Network



› TYPES OF VARIABLES IN THE MODEL

INPUT, ANALYSIS, GOAL

- Input
 - In the model, these variables don't have 'parents'
 - Parameters have to be filled out by the user
- Analysis
 - Have both 'parent' and 'child' variables
 - 'Probability table' determines relation from input 'parent(s)' to output 'child'
- Goal
 - Terminus of the model, the variable that you want to determine (e.g. Impact)
 - 'Probability table' determines influence of 'parent' variable(s)



› EXAMPLE INPUT VARIABLE PHISHING AWARENESS

› What is the likelihood of an employee **not** falling for a phishing scam

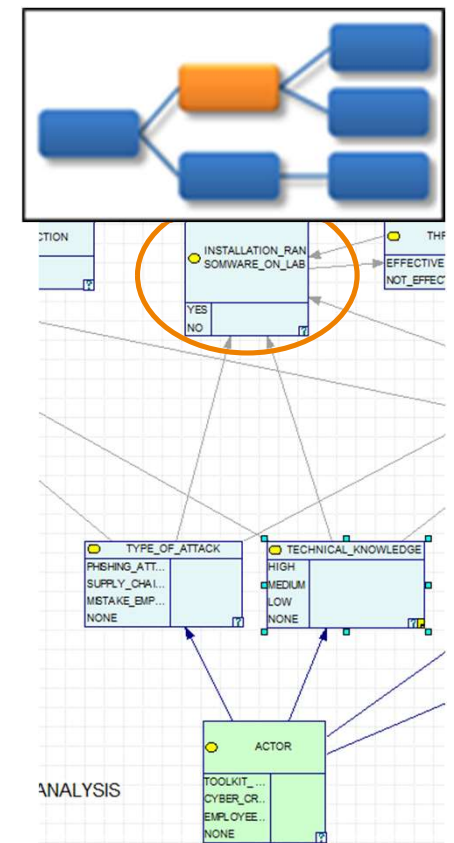


Phishing awareness	Probability
Effective	30 %
Not_effective	70 %

EXAMPLE ANALYSIS VARIABLE INSTALLATION OF RANSOMWARE

- Given a specific type of attack, what is the probability of the successful installation of ransomware?

	Type of attack	Phishing					
	Phishing awareness	Effective			Not effective		
	Technical expertise (attacker)	High	Middle	Low	High	Middle	Low
Instal. Rans.	Yes			1%	30%		
	No			99%	70%		



EXAMPLE ANALYSIS VARIABLE INSTALLATION OF RANSOMWARE

- Given a specific type of attack, what is the probability of the successful installation of ransomware?

		Installation Ransomware?					
		Effective			Not effective		
Instal. Rans.	Type of attack						
	Phishing awareness						
	Technical expertise (attacker)	High	Middle	Low	High	Middle	Low

› OUTPUT

The acquired model can be used in multiple ways!

- › Determine the probability distribution of one or more goal nodes
 - › Straightforward calculation of probable states based on input

- › Given the actual end state, what was the likely input state?
 - › In case of a certain actual impact, what was the likely attack path? (actor, type of attack, ...)

- › Sensitivity analysis
 - › What if...?
 - › Is it worthwhile to invest in certain measures? Or do they barely have any effect?
 - › Is it crucial to understand parts of the model in more detail? Or do they have limited bearing on the output?

HANDLING DYNAMICS IN CYBER RISK



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A system dynamics approach

System Behaviour Paradigm

The *observed behaviour* of any eco-system is a function of the operational *structure* that drives it (Paich 2009):

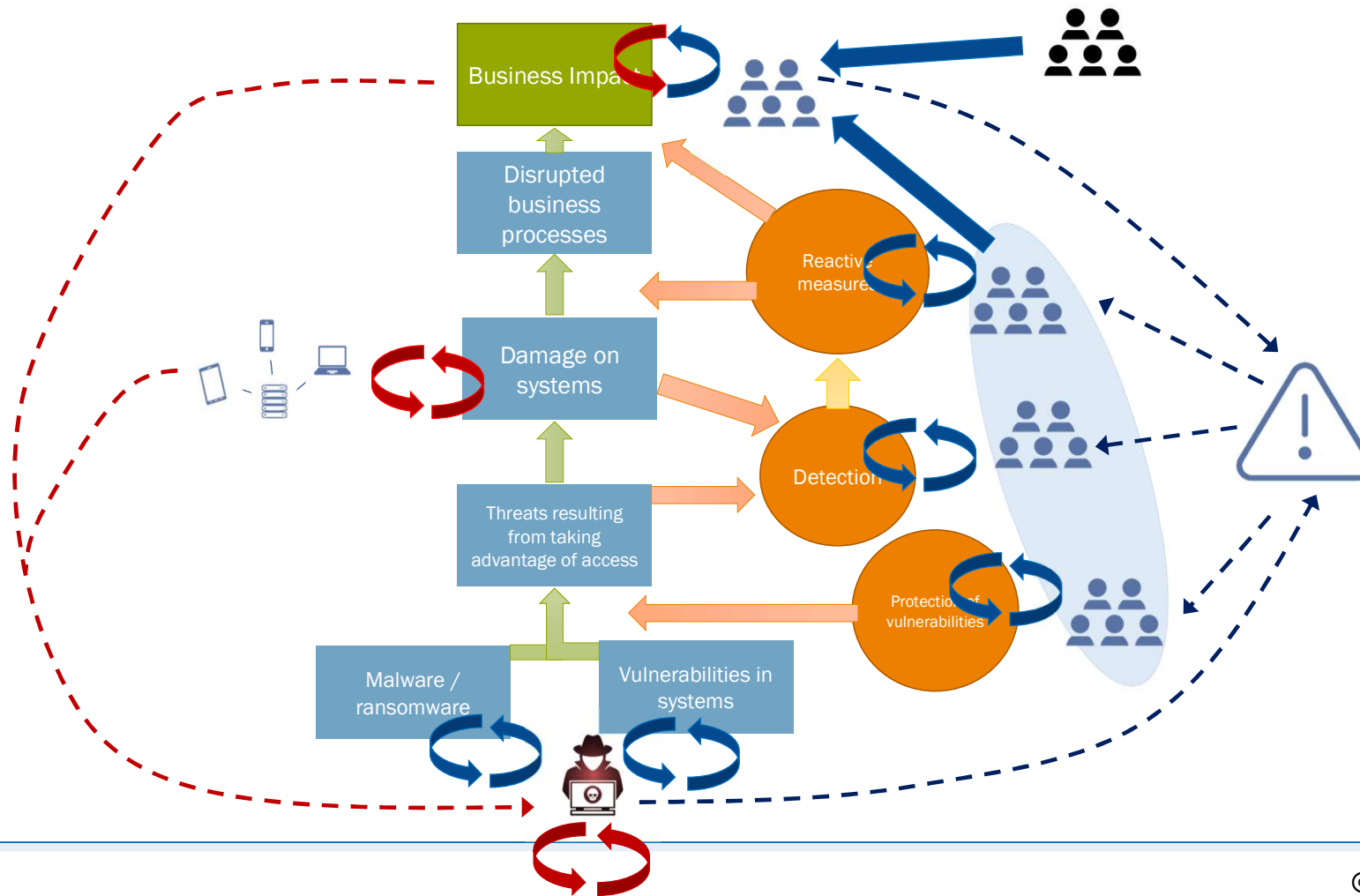
- *What eco-system structure (feedback, accumulation and time delays) drives its outcome of interest (behavior over time)?*
- *How can we influence this structure (what policies) ?*

Security strategies and programs are significantly affected by *the dynamic nature of cyber risk*.

Examples are:

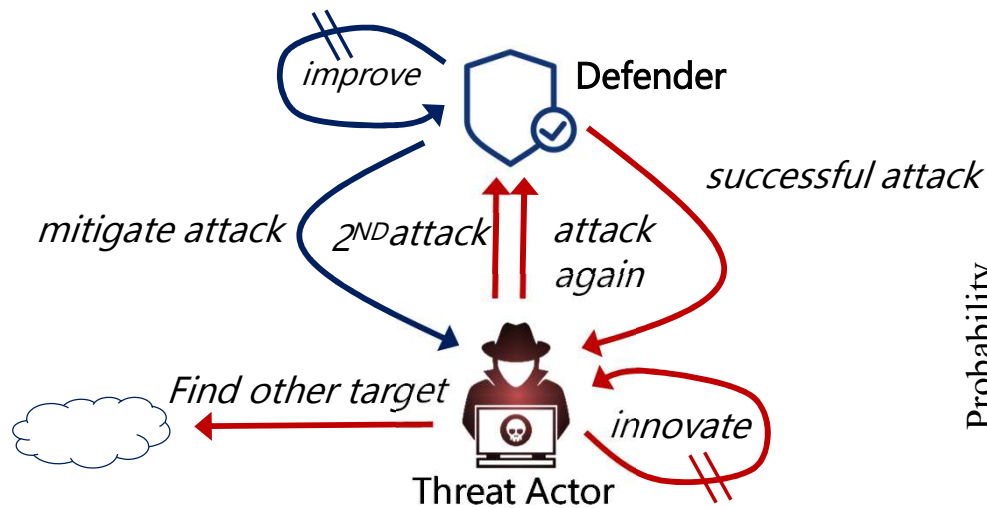
- Evolving attacker tactics
- Changing organizational characteristics
- Limits to resource availability
- Emerging impact of incident response and recovery
- Etc.

Dynamics: attacker - defender interaction & response



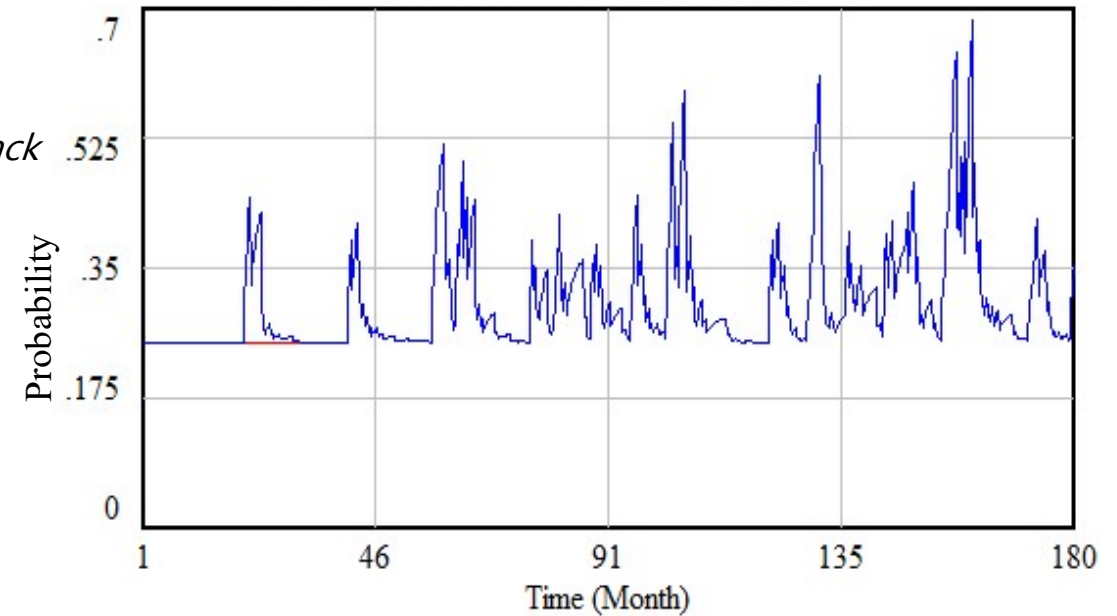
Example: attacker's behavior: structure and behavior

Structure

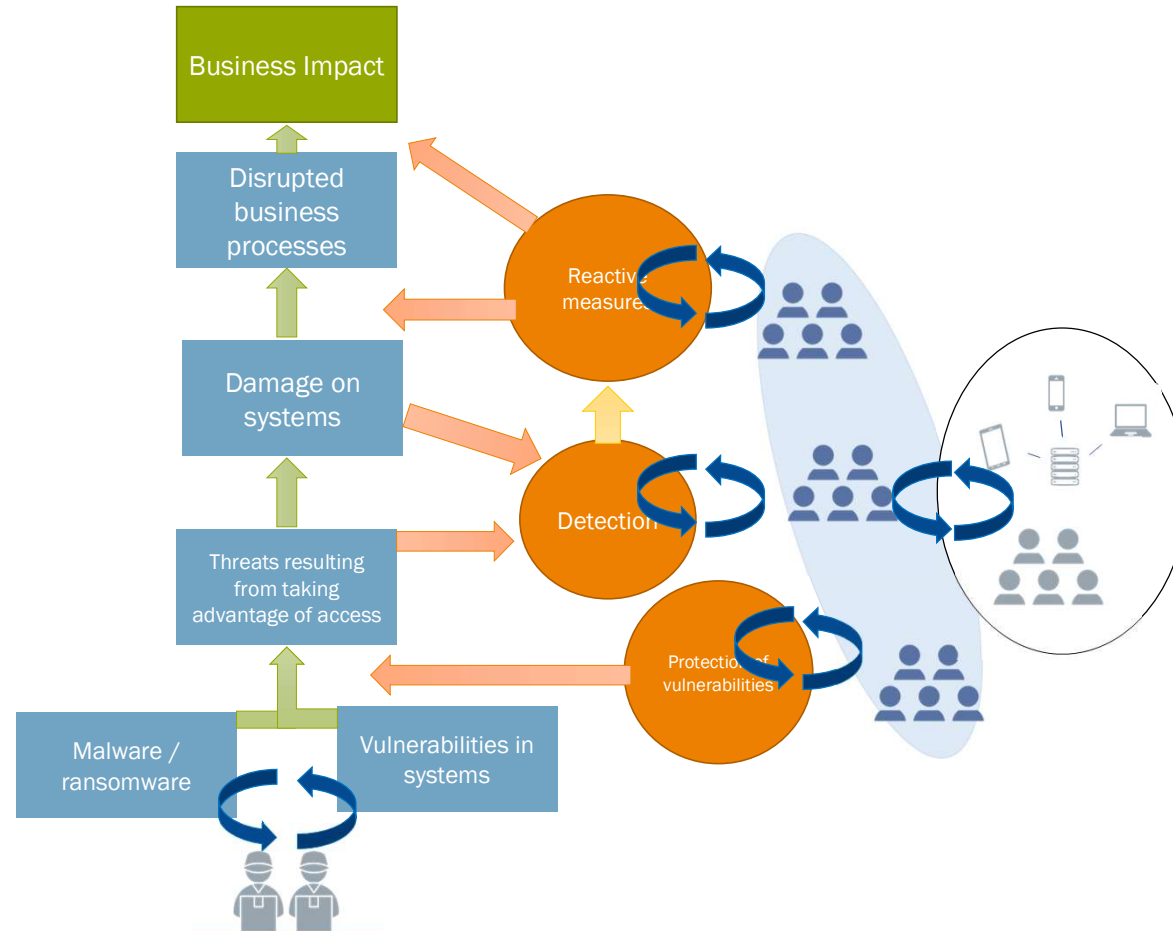


Behavior

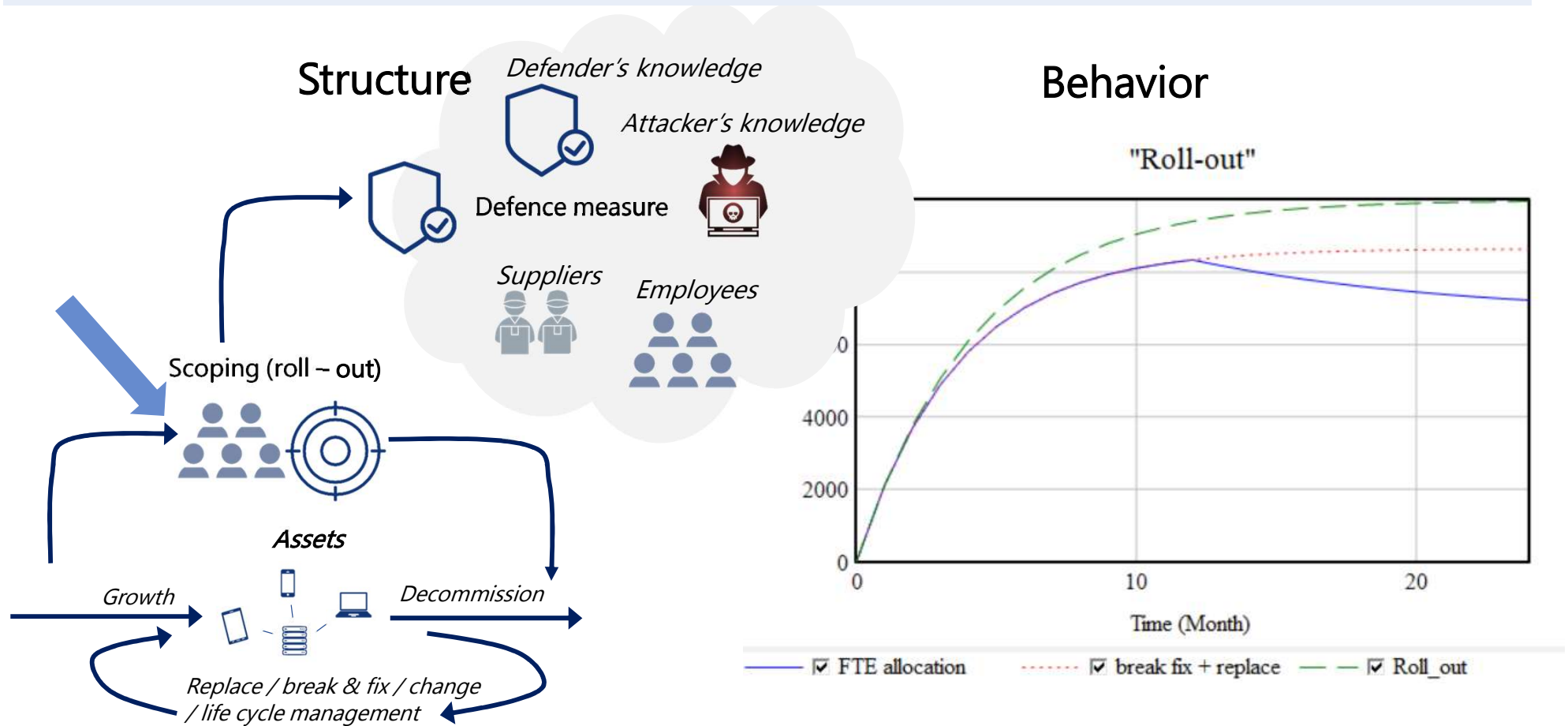
DDOS attack probability graph



Dynamics: changing organization: suppliers, asset base, insiders



Example: security capability roll-out: structure and behavior



Why using a system dynamics approach?

Ultimately these dynamic effect *determine the success* security strategies and programs.

An system dynamics approach *accounts for these effects* by:

- 1) Providing explainable and understandable insights in the structures that causes these dynamics
- 2) Compute and simulate realistic future performance of relevant performance indicators while using these structures
- 3) Identifying relevant organizational levers that can influence these dynamics

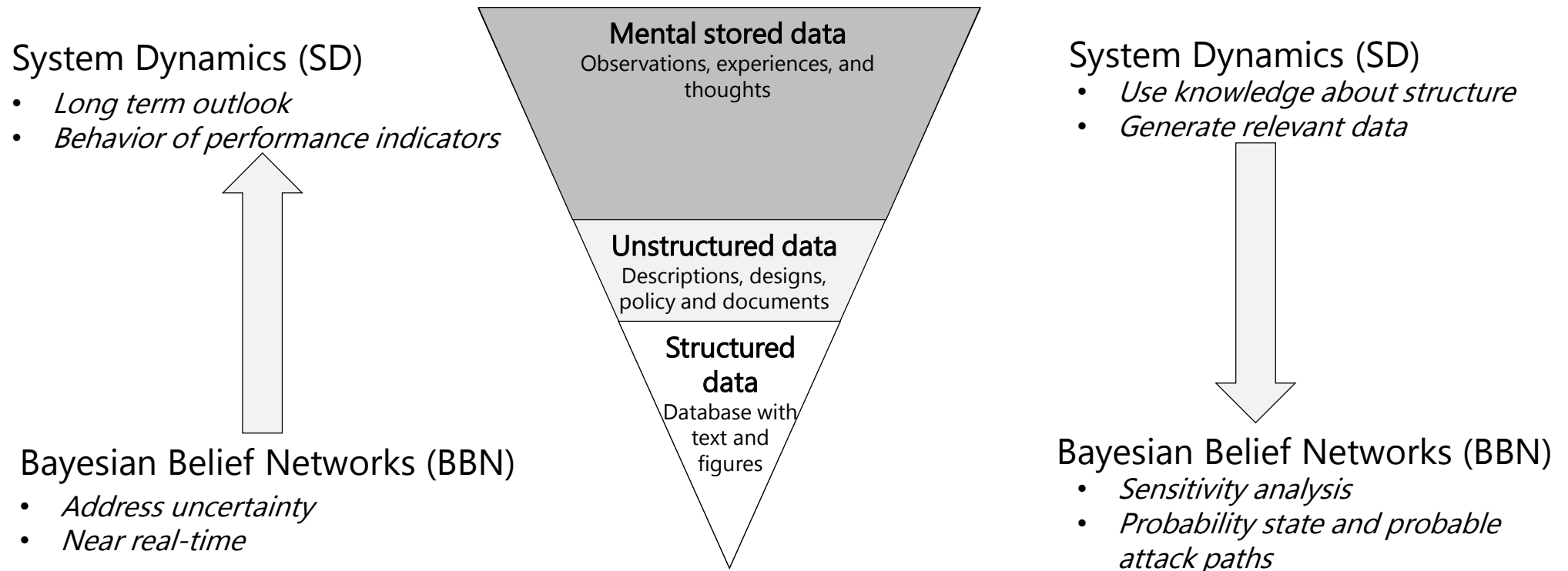


CONCLUSIONS

The contribution of cyber risk quantification

- Substantiate and rationalize cyber risks.
- Enable decision makers to prioritize cyber risks amongst other risks.
- Give credence to cyber-security strategy and program.

Cyber risk quantification: Bayesian Belief Networks (BBN) and System Dynamics (SD) complement and reinforce each other



› **NEXT STEPS**

Research project NCSC and TNO

- › Finalise and describe approach and models
- › Explore inclusion of models and tools in processes NCSC
- › Stimulate data collection process

TNO innovation
for life

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Nationaal Cyber Security Centrum
Ministerie van Justitie en Veiligheid

» **THANK YOU**