



Probabilistic Mission Defense and Assurance

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Summary: Defending and Assuring the Mission

- ▶ situation: **mission is threatened**.
- ▶ task: need to **respond** adequately.
- ▶ goal: **assure mission success**.
- ▶ constraint: **without sacrificing** mission for security.



Challenges

- ▶ understand how a **threat affects a mission**.
- ▶ understand **countermeasures diminishing threats**.
- ▶ understand the **bad sides of countermeasures** causing **negative side-effects** on the mission.

Current Approaches & Problems

- ▶ **holistic approaches** deliver **intransparent** “optimal” solution.

exaggeratedly “Response XYZ is best with metric 4589.32”.

- ▶ require **unacquirable information**, do not encompass **unforeseeable events**

complex ACTs. manually intractable. automatic generation ↔ single missing links.

- ▶ optimize cost, **without** considering **negative side** of countermeasures.

shutdown of central control server is very cheap!

Our Approach & Outline

- ▶ paradigm shift. **model the mission**, not the attacker.
- ▶ **model the good and the bad** sides encompassing **uncertainty**.
- ▶ reduce problem to **mathematically well-defined** probabilistic inference **problem**.
- ▶ **decouples** assessments from generation of responses and from selection.
- ▶ delivers **directly understandable** and validated results.

The probability that our mission becomes adversarially impacted is 58% (□◻◻◻◻◻◻).

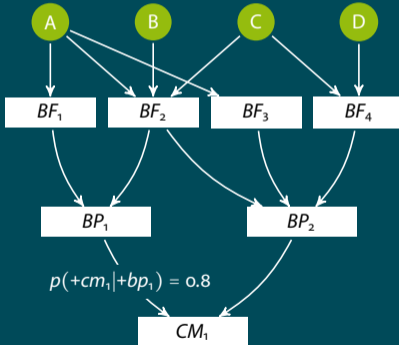
We can reduce this by 80% (to <◻◻◻◻◻◻). There exists a 30% probability of immediate conflict (<◻◻◻◻◻◻)



Probabilistic Approach

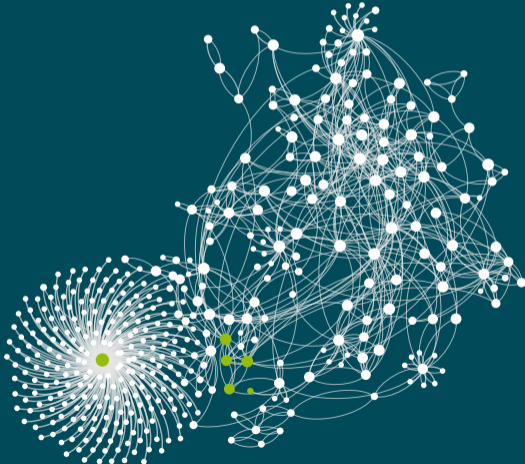
- ▶ model problem from **three different perspectives**.
- ▶ **collect** potentially disagreeing **information from multiple experts**.
- ▶ make the model able to understand disagreements;
do not enforce a bad compromise.

View 1: The Mission (or a company)



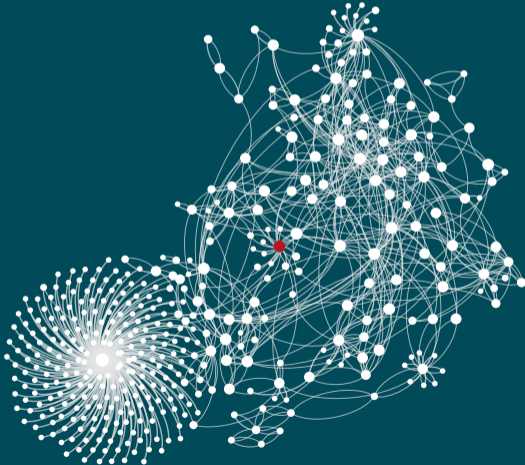
- ▶ dissect mission into **smaller pieces**.
- ✓ collected **directly from** business and mission **experts**.
- ▶ conditional probabilities are understandable and validatable
 “probability of mission failing, given BP₁ fails is 80%” 🎲🎲🎲🎲🎲🎲
- ▶ *frontend or backend?*
- **mission critical devices (ABCD)**
 only scratch surface of infrastructure

View 2: The Infrastructure



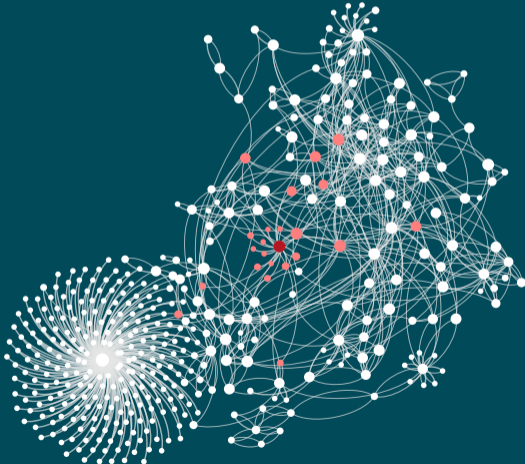
- ▶ **MCDs** are **only tip of the ice berg**
- ▶ huge **complex dependency structures**
- ✓ **automatically learnable**
- ▶ same **conditional probabilities** as before!

View 3: The Impact



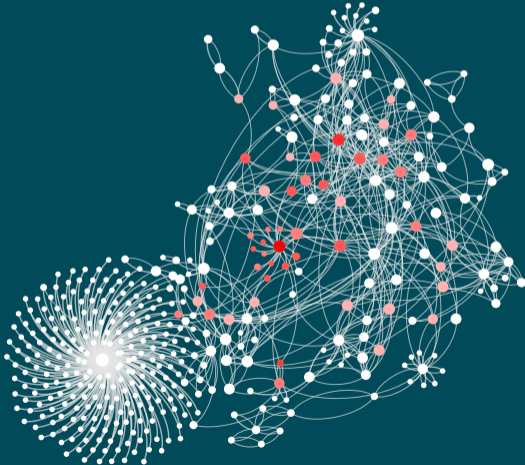
- ▶ something **fails** or is **attacked**.
- probability of **local impact**.
- ▶ leads to **global impact**

View 3: The Impact



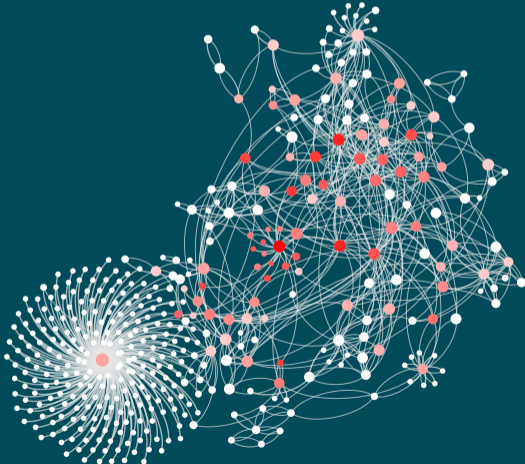
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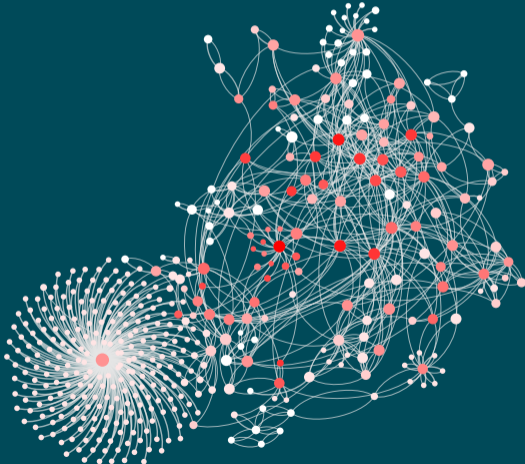
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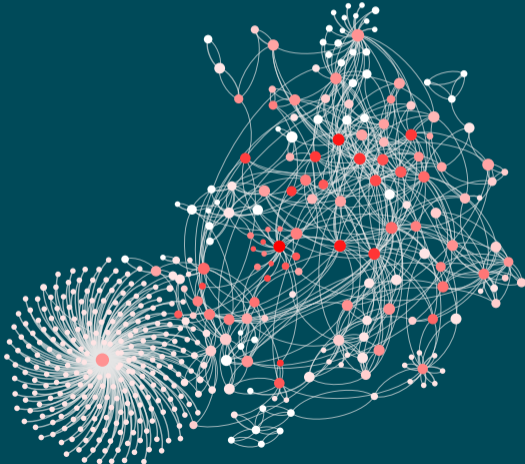
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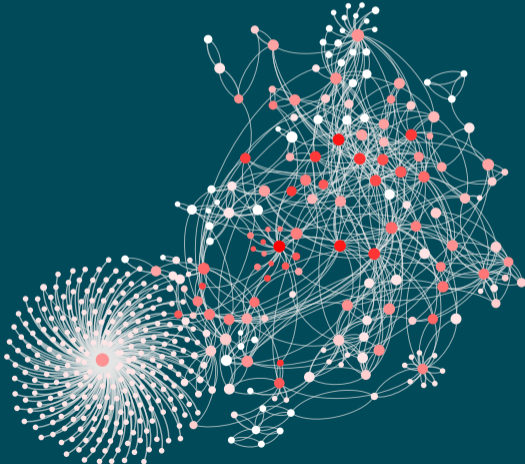
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- probability of **local impact**.
- ▶ leads to **global impact**
- ▶ might even **spread...**
- ▶ ...to dependent nodes...
- ▶ ...to dependent nodes...
- ▶ until **everything** is impacted.
- ▶ how to assess?

Problems of “Spreading” Algorithms



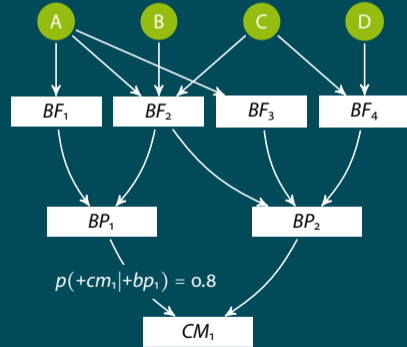
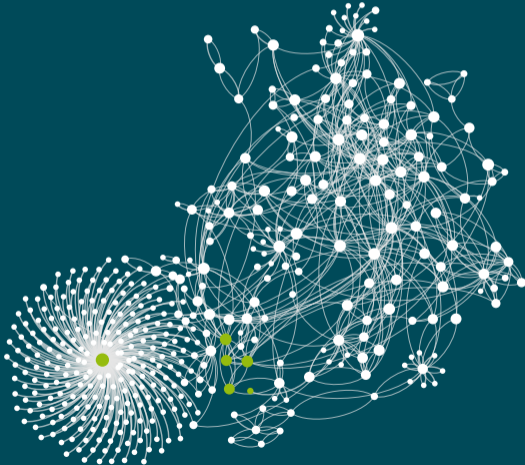
- ▶ various novel “spreading”-algorithms exist.
- ▶ novelly designed, hand-crafted.
- ✗ unclear behavior.
- ✗ sense for parameters missing.
- ✗ no clear definition for interpreting results.
- only **deeply trained experts** can **parametrize** models and **understand** results.

Problems of “Spreading” Algorithms

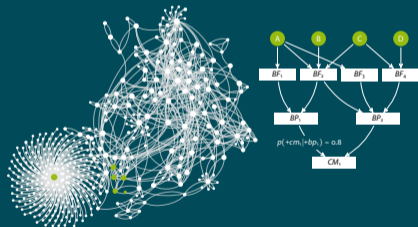


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- ✓ reduce to mathematical problem!

Mission Impact Assessment is a Probabilistic Graphical Model

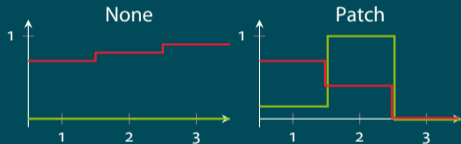


Mission Impact Assessment as a Probabilistic Inference Problem



- ▶ probabilistic inference
projects local impacts globally on the mission.
 - ✓ well-defined mathematical problem.
 - ✓ **validate the model**, not the algorithm.
 - ✓ **parameters** define their own **semantic**.
 - ✓ **results** are **directly understandable** by everyone.
- **model** adversarial **threats**,
countermeasures **positive & negative**
intuitively and locally.

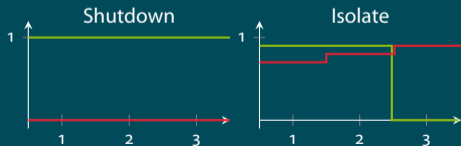
Modeling Defense and Threats Locally: direct impact example



▶ **vulnerability** creates probability of **adversarial impact** **varying over time**: short-, mid-, long-term

▶ **shutdown** suffocates **AI**, but **nothing works**.

▶ **patching** causes prob. of conflict: **operational impact** short: **installation** conflict, mid: **reboot** required, long: vulnerability **removed**.

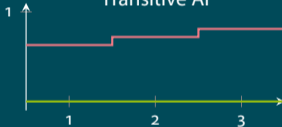


▶ **isolate**: no *local* "positive" effect. negative=shutdown.

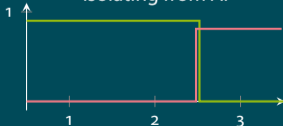
Modeling Defense and Threats: transitive-effects example



Transitive AI



Isolating from AI



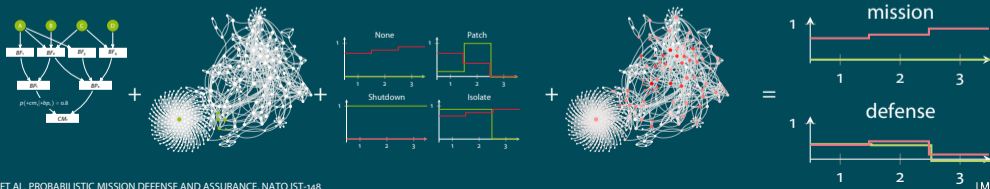
- ▶ node A depends on affected node X. impact “**spreads**”.
- **transitive adverserial impact**
(not modeled, *assessed automatically*)
- ▶ **isolate X** for short- and mid-term
blocks adverserial impact for X (*assessed automatically*)
- ▶ but **blocks required information flow** towards A
- **operational impact** on A

Probabilistic Inference

- ▶ local impact models create **impact time-profiles**.
- ✓ considers **adversarial and self-inflicted impact** on the mission.

- ▶ **probabilistic inference projects local impacts** to the **global mission impact**
- ✓ directly **understandable**, interpretable and reportable.

- ▶ no novel spreading-algorithms, well defined mathematical problem
- ✓ **models can be validated** directly. **no holistic validation** required.



Probabilistic Inference

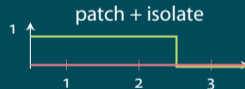
- ▶ assessment for the **current situation**, **benefits of our response** and its **negative side effects**.



- ▶ *probability of impact on the mission over the time.*
- ✓ based on **acquirable** and automatically **learnable data**.
- ✓ accept **disagreeing information sources** and **directly reflect expertise**.
- ✓ captures **unforeseen events and uncertainty** “*what all could happen*” through **transitive impacts**.

Experiments

- ▶ **experimented on live data** with artificially placed adversarial impacts in Acea ARETi.
- ▶ based on the placed impacts, **doing nothing was the best option** in short and mid term.
- ▶ **patching and isolating** first gains superiority in **long term perspective**.




- ▶ $\{\max_i(OI^i) = 0.8\} > \{\max_i(AI^i) = 0.75\}$
- ▶ delivered **non-trivial results** and precisely raised awareness for the **compromise between the good and the bad**.

Further Usage

- ▶ **no vendor lock in:** assessments are an **independent quality control**. no reference required.
e.g., evaluate response plans from other applications, handbook approaches, intuition, guidelines, obliged responses
- ▶ **selecting “the best”** is = **find multi-dimensional semi-optimum** among all evaluated
note: applicable on any subset. no holistic evaluation required
- ▶ use PGM to find POIs to **strategically block** communications
min-cut graph problem based on edge-contribution probability

Conclusion

- ▶ **probabilistically** well-defined **mission impact defense** and assurance approach.
- ▶ encompasses uncertainty, missing links and disagreeing information sources.
- ▶ encompasses **the good and the bad** sides
- ▶ available and **easily acquirable data**.
- ▶ **validated results, understandable without context** and **without biased interpretations**.
- ▶ suitable for **reporting** along command chains and in documents for **justifying decisions**.
e.g., understanding does not require vulnerabilities, attack paths, or algorithmic properties.
probability of 83% remains 83% and is equivalent to tossing .



Thank You.

... questions?

Independent Assessment

- ▶ every assessment can **stand on its own**.
- ✓ understandable knowing neither the algorithms nor the models.
- ✓ **no holistic reference set** required to “interpret” results. **no biased interpretation**.

*“Probability of impact onto mission is **83%**” instead of “Mission impact is **4546.345**”.*

- ▶ therefore: **response-proposals may origin from anywhere**.
e.g., other applications, handbook approaches, intuition, guidelines, obliged responses
- ▶ assessment is an **independent quality control**.

Decoupled Selection

- ▶ problem: given multiple assessments, **select a single response**.
- ▶ multiple dimensions might disagree.
- ▶ difficult to design a “cost function” among dimensions.
- ▶ well-defined problem: **unweighted multi-dimensional minimization** finding **best compromise**.
- ✓ **no bias** towards one dimension required.
- ✓ **does not require all possible responses** to be evaluated.

$$\hat{RP} = \min_{\epsilon} \left(\left\{ \bigcap_{d \in \vec{d}} RP_{\epsilon}^d \right\} \neq \emptyset \right). \quad (1)$$

Decoupled Generation

- ▶ problem: **how to obtain adequate responses?** use PGM for generation!
- ✓ no bad generation possible: **assessment is independent** judgment.
- ✓ “one does not have to find *the* best, only one whose assessment is independently acceptable.”
- ▶ **randomly sample**: shutdown critical, directly affected, patch affected, isolate patched nodes.
- ▶ where to **strategically block communication**, i.e., firewall rules?
mathematical problem in PGM: **find min-cut** in model based on minimal edge-contribution probability.