

CYSE 650 Final Project

Andy, Connor, and Wesley



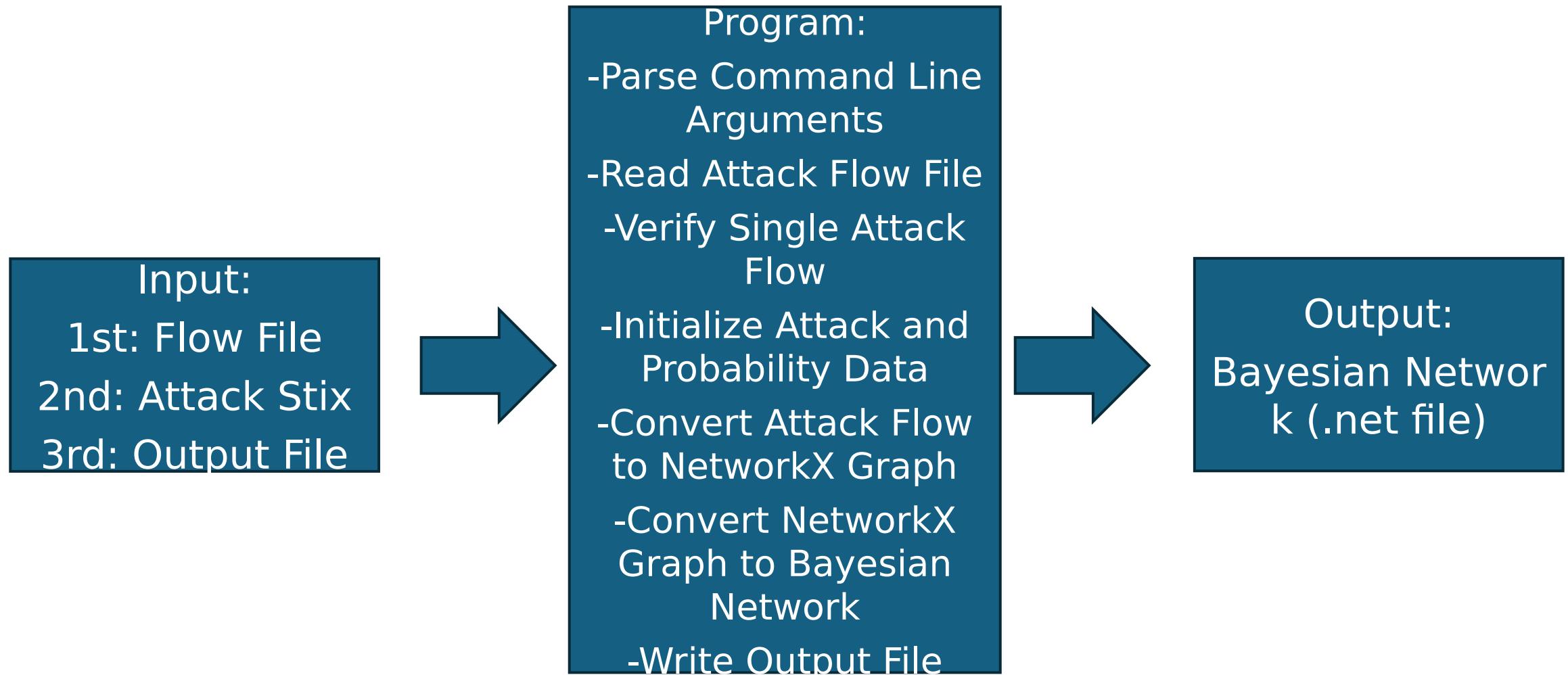
Problem Statement

Develop a framework to analyze **cyber-attack patterns and probabilities using Bayesian networks**, in an automated and user-friendly fashion.

How to Install/Configure

- GitHub Link: (More in-Depth Installation Guide in the Repo ReadME)
 - <https://github.com/Cyber-Musketeers/ATTACK-RISK>
- Step 1: Clone the repository:
 - RUN: git clone <git@github.com:Cyber-Musketeers/ATTACK-RISK.git>
- Step 2: Install poetry
- Step 3: Poetry Shell
- Step 4: Make flow
- Step 5: Export flow
- Step 6: Run main.py (ATTACK-RISK/src/main.py) with the following arguments:
 - 1st: Attack Flow File
 - 2nd: MITRE ATT&CK Stix File
 - 3rd: Output File
- Step 7: You will receive a complete Bayesian network file (.net) to be used within your program of choice.

Software Architecture



Software Architecture (Cont.)

- Dependencies:
 - Networkx
 - Numpy
 - Stix2
 - Mitreattack.stix20
 - Pgmpy

(Probability values are gathered from comparing the number of times TTPs appear in ATT&CK campaigns)

```
def main() -> None:
    """
    Main function of the program.
    """

    args = parse_args()
    # print_flow(args.flow_file)
    flow_bundle: stix2.Bundle = read_flow_file(args.flow_file)
    flows = flow.get_flows_from_stix_bundle(flow_bundle)
    # technically there can be multiple attack flows in a stix bundle but I am too lazy to deal with that
    if len(flows) != 1:
        raise ValueError("Expected exactly one attack flow in the file.")
    attack_data = MitreAttackData(args.attack_stix)
    probability_db = weights.ProbabilityDatabase(attack_data)
    flow_nx = convert_attack_flow_to_nx(flows[0], flow_bundle)
    flow_nx = make_nx_graph_more_readable(flow_nx)
    bayesian_network = flow_nx_to_pgmpy(flow_nx, probability_db)

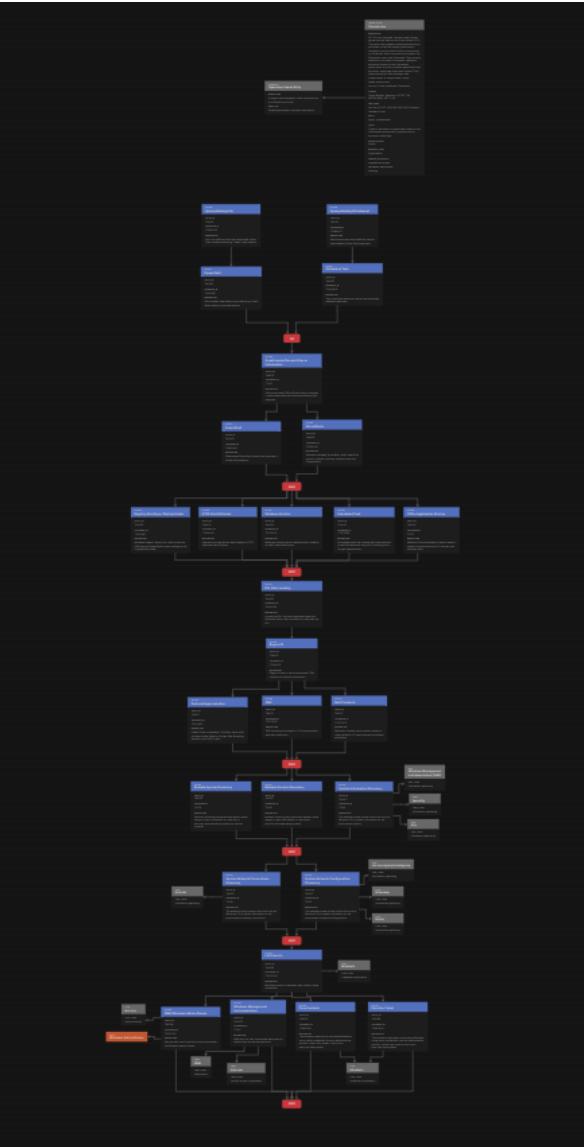
    content = pgmpy_to_unbbayes_hugin(bayesian_network)
    with open(args.output_file, "w", encoding="utf-8") as file:
        file.write(content)
    print("New bayesian network written to", args.output_file)
    print("The network can be loaded into Hugin or Unbbayes")
    print("Thanks for using the program!")
```

Tool Functionality Scenario 1: Uber Breach



The Uber breach has a more linear attack flow which can be used to showcase the implementation into the Bayesian network and the associated risk values. The flow goes from compromised accounts to exfiltration.

```
python3 ./main.py --flow_file uber.json --attack_stix enterprise-attack.json --output_file uber.net
```

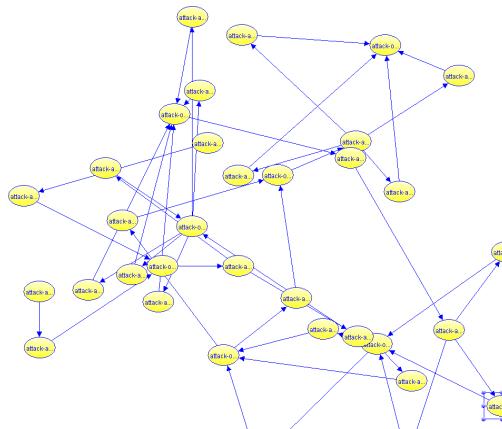


Tool Functionality Scenario 2: Cobalt Kitty

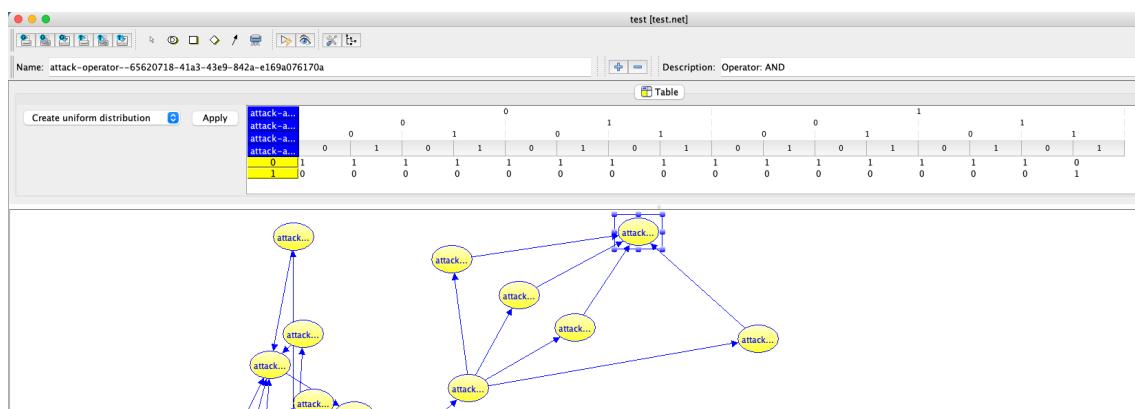
- This attack flow is much more complex and has multiple actions within certain layers
- Ultimately showcases the risk associated within Cobalt Kitty and the associated TTPs within each layer.

```
python3 ./main.py --flow_file ck.json --attack_stix enterprise-attack.json --output_file ck.net
```

Results



Cobalt Kitty



Uber

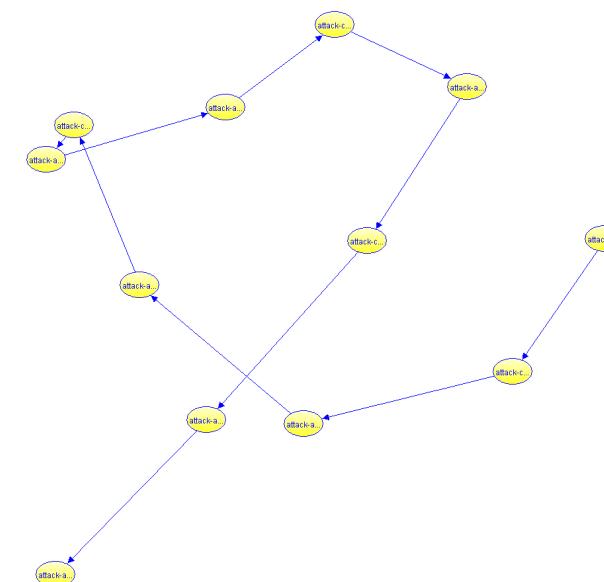


Table	
Create uniform distribution	Apply
	attack-action
0	0.99
1	0.01

Limitations

- If a technique doesn't show in any ATT&CK campaign, or if a tactic does not have an associated technique, we are assuming that it has a **0.01%** chance probability (Uber output)
- Methodology for probabilities is weak (**less than half** of TTPs have a campaign!)
- For condition blocks within ATT&CK flows, we are only considering values of "**True**"
- We are only accepting **one** attack flow at a time
- **Slow** startup time

Future Works

- Improve weight calculation to consider more than “if it showed up in many campaigns, it must be likely”
- Handle “false” determinations in “Attack Condition” blocks
- Better weight calculation for “Default” probabilities
- Less-naïve implementation of CPTs
- Better node names