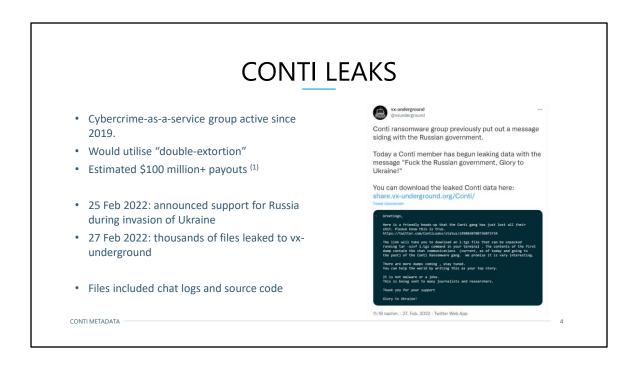


Australia Day 2022
Target tracking purchases predicting pregnancy
Government tracking and tracing
Predictive Algorithms
But really what can you do with it



Not a Threat Intel Profile

Presented with this range of data I wanted to see what I could do with it.



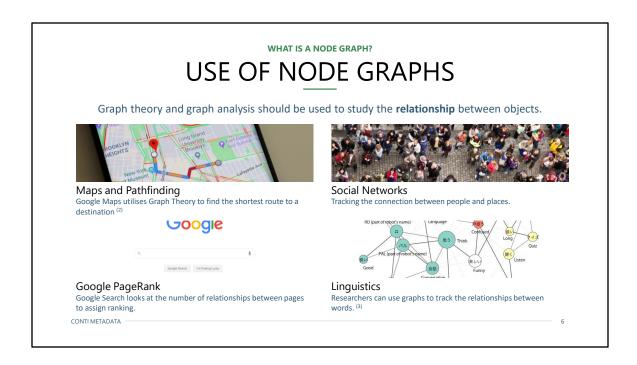
Ultimately the Conti chat files gave us over 168'000 interactions between 465 unique actors which would have to be reviewed.

Huge task for an analyst

If we could model the metadata, it gives us a place to start and target our analysis efforts.

Diving into the logs, it looks like it provides us enough metadata to get a head start Time and who is talking to who is all we need. Ultimately what is being said is irrelevant to us.

What we can do is use the metadata to look at the *relationships* between actors in order to assist with further analysis.



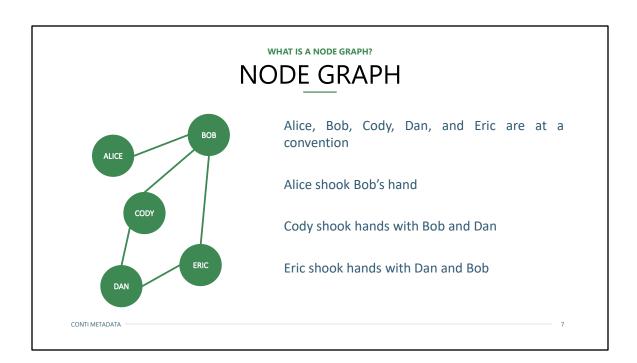
WE USE THESE GRAPHS TO LOOK AT RELATIONSHIPS

One of the first uses for Graph Theory was mapping how to get from island to island in Konigsburg without crossing a bridge twice.

Linguistics: Tracking words used by patients by non-talkative schizophrenic patients

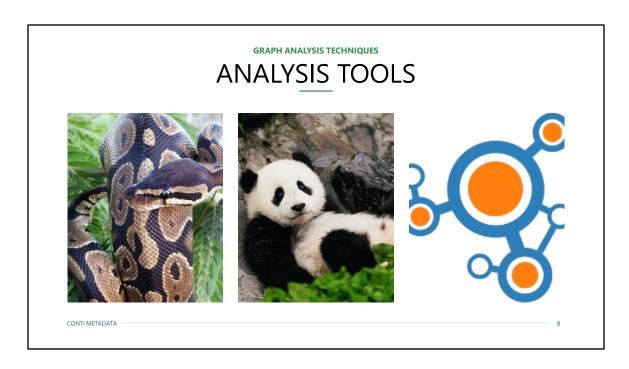
Bloodhound uses Node Graphs looking at the relationships between different domain objects.

If you collect network data in a SIEM you can build a node graph to characterise that traffic.



Read the slide

There are different types of node graphs. You can have directed graphs that show dependencies or interactions. Or weighted graphs.



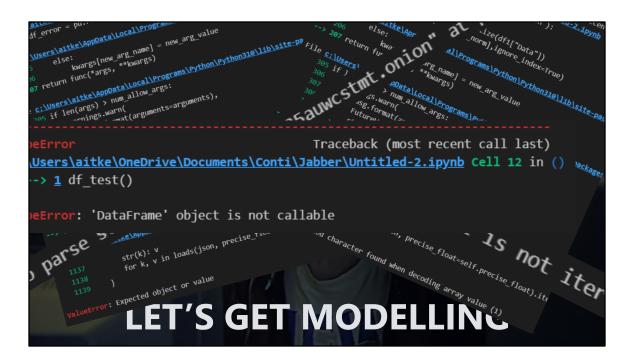
Three primary tools will be used for Analysis.

Python, because it's better than trying to do this all in PowerShell (you may laugh, but that's what I used to do. Because we weren't allowed Python on our work computers)

(this one is a ball python)

Pandas, to help us manipulate the JSON data and it has some basic statistical tools built in.

NetworkX, which doesn't have a cute animal logo and should be ashamed of themselves. Package made for drafting complex networks.

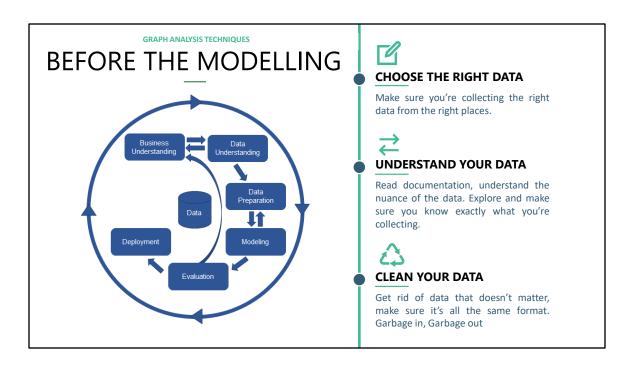


So, I allocated myself 24 hours to get cracking on this project

Totally not because I had a university assignment due the next day



So 18 hours into my 24 hours I've finally completed the work I needed to do *before* modelling and my housemate found me crying in the shower.



This is CRISP-DM – Reference 4. If you ever work with data in your job for anything, check this model out and use it as a structure for your workflow.

Collect your data, have a play, explore it.

Describe it, understand the qualities and features.

Cleaning the data is the lengthiest task. (side note: there is a standard for JSON – ISO/IEC 21778 – developers who do not follow it should be shot. I will take no questions on this matter)

GRAPH ANALYSIS TECHNIQUES

CREATING THE GRAPH

```
1. node_series = conti_df['to_short']
2. node_series = node_series.append(conti_df['from_short'])
3. node_series = node_series.drop_duplicates()

4. G = nx.Graph()
5. G.add_nodes_from(node_series)
6. G = nx.from_pandas_edgelist(conti_df, source='to_short', target='from_short')

7. pos = nx.spring_layout(G,scale=500,k=50/np.sqrt(G.order()))
8. d = dict(G.degree)
9. nx.draw(G, pos, node_size=[d[k]*20 for k in d], with_labels=True)
```

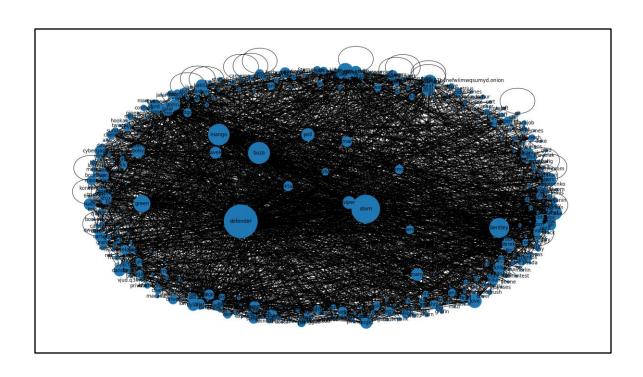
There will be very little code in this presso, I promise. But this is for the people who want to know what happens behind the scenes.

Essentially there are three parts to creating the graph. Identifying the nodes (1-3)

Creating the graph structure (4-6)

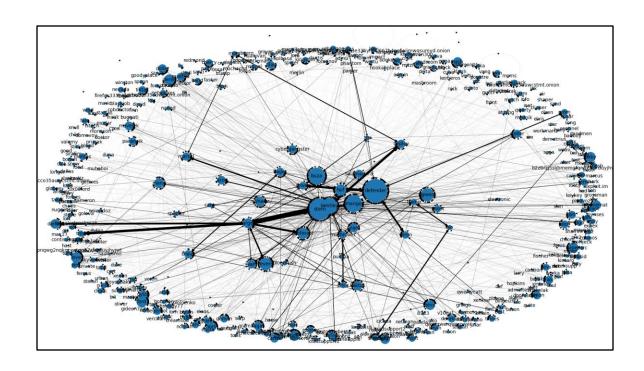
Finally drawing the graph (7-9). What we're saying in 7-9 is that we want the more important nodes to be bigger.

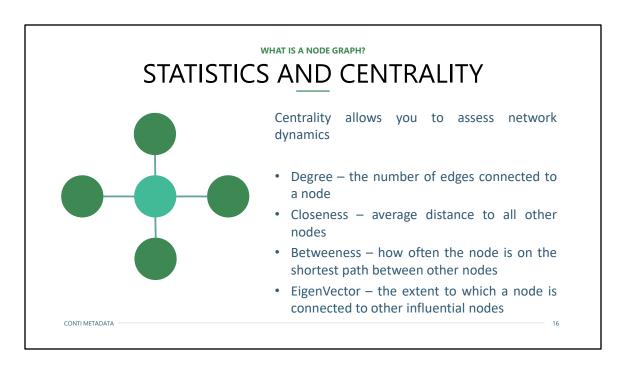
Great! So now we're going to have a pretty graph that will give us all the answers!





Graphs with this many nodes become tricky to purely use as analysis tools. Those magical answers you thought you would find in nine lines of code probably wont be there. But there are some small tricks...





Really the graph is just a means to an end. What matters is the stats you can pull from the graph. Graphs allow you to identify *centrality*. Centrality tells you who is important in a network

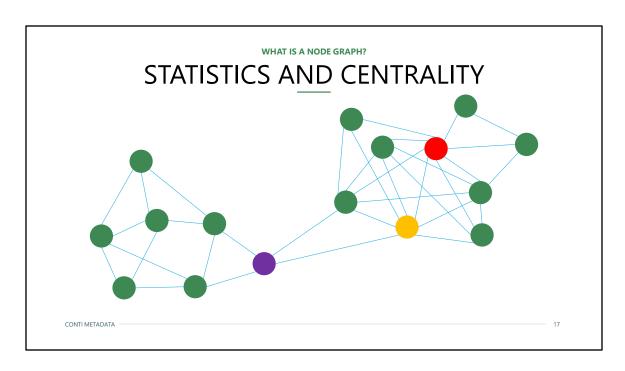
Degree: exposure to the network. Opportunity to directly influence.

Closeness: Who diffuses information to the network.

Betweeness: Informal Power. Broker Resources. Controls flow of information.

Eigenvector: Also used to identify informal power. Particularly useful in dense

networks. "Not what you know, but who you know"



Red: Degree

Yellow: Closeness

Purple: Betweenness

GRAPH ANALYSIS TECHNIQUES

GATHERING STATISTICS

```
1. df_metrics = pd.DataFrame(dict(
2.    in_degree = nx.degree_centrality(G),
3.    eigenvector = nx.eigenvector_centrality(G),
4.    closeness = nx.closeness_centrality(G),
5.    betweenness = nx.betweenness_centrality(G)
6. ))
```

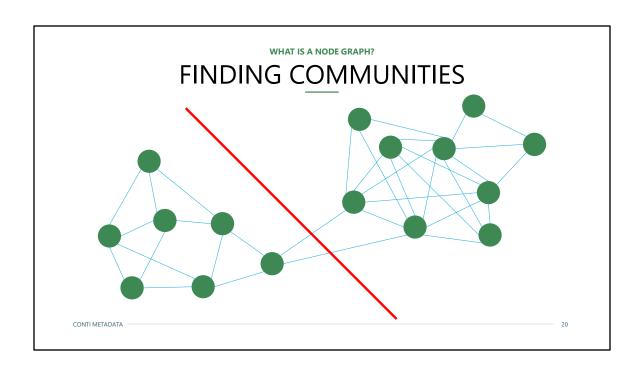
		CEN	ΓRALITY		
NAME	CENTRALITY	NAME	CENTRALITY	NAME	CENTRALITY
Defender	1.058	Stern	0.463	Defender	0.293
Stern	0.859	Defender	0.475	Stern	0.167
Mango	0.522	Bentley	0.414	Mango	0.065
Buza	0.511	Mango	0.413	Buza	0.058
Bentley	0.450	Buza	0.407	Bentley	0.043
Green	0.303	Green	0.402	Ford	0.031
Revers	0.281	Professor	0.372	Revers	0.025
Hof	0.225	Revers	0.372	Green	0.022
DEGREE		CLOSENESS		BETWEENNESS	
Exposure to the network. Opportunity to directly influence.		Who diffuses information to the network.		Informal Power. Broker Resources. Controls flow of information.	

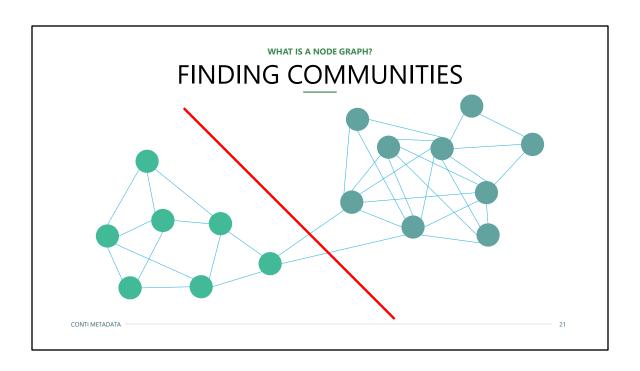
Purely by looking at the metadata statistics, we can start to find some information about Conti.

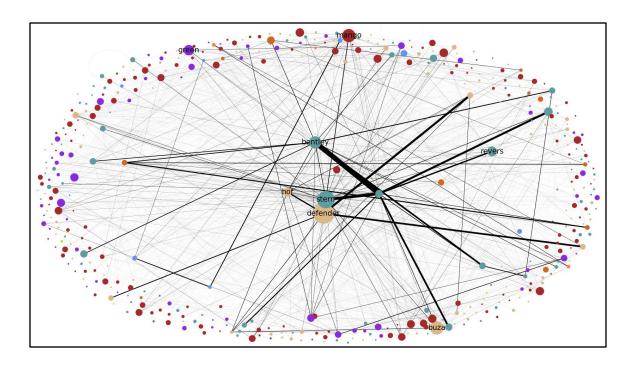
There are a few nodes with high authority. These are obviously highly influential nodes within the network. These nodes are the centre of information. It also means the network isn't highly resilient. Loss of these nodes could lead to a communication breakdown.

The closeness start middling, but like degree trend downwards rapidly. Our players with the highest degrees are highest on the closeness. Are these individuals regularly in contact with everyone in the group? Interestingly Bentley moves up and Professor appears on this list. Do these players serve roles as middle-managers? Maybe they have larger teams?

Defender and Stern present with the highest betweenness scores. This really represents that there are no other information channels within the network. To get from one side of the network to the other you probably have to go through them.







Here's our network with some community detection applied. For ease of reading I've only allowed the names of the people that we discussed earlier.

Community detection is difficult to apply to this graph. Stern and Defender appear to regularly interact with all elements of the network. The algorithms don't handle that well. Maybe we should drop Stern and Defender and see what shakes out?

Stern and Defender appear to have their own core groups that they manage. Additionally there is an unnamed blue dot that Bentley and Stern both interact with regularly? Why?

Mango and Green appear to be the only nodes of significance in their community, cementing the idea that they might be team leaders.

Who is this orange dot? They appear to be somewhat central, but not within the upper leadership echelons?

OUTSIDE RESEARCH DEFENDER STERN BENTLEY MANGO BUZA Identified as one Mid-level Mid-level Conflicting Another senior of the leaders of manager or manager and information. member. senior developer. coder. the group. Described as Representative Often described Instructed junior Handled QA from Emotet "COO" as the "CEO" members on works. (KrebsOnSecurity) compared to Stern's "CEO". duties. "Catankerous Typically tasked Technical taskmaster" -Supposedly in by Stern on manager in Handled finances and internal KrebsOnSecurity charge of antirandom side charge of coders. antivirus works. projects. (Check Point) logistics. CONTI METADATA

KrebsOnSecurity did a big series of four pieces Forescout Check Point Rapid7

Mango: "Blockchain", "hacker social network"

TAKEAWAYS



EXPLORE CENTRALITY

- Remove key nodes
- Discover pathways
- Identify key individuals



UNDERSTAND COMMUNITIES

- Who is the core of a group?
- Who is missing?



PLAY WITH THE DATA

- Explore what you're looking at
- Add, remove, chop, change
- Try new things

CONTI METADATA

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- 6. <u>forescout.com/resources/analysis-of-conti-leaks/</u>
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- $\textbf{8.} \quad \underline{\text{https://www.rapid7.com/blog/post/2022/03/04/graph-analysis-of-the-conti-ransomware-group-internal-chats/}\\$

CONTI METADATA

