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Cyber Security Risk Assessment Fall 2016

Lecture 14


Quantitative Risk Analysis

Uncertainty

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Two Types of Uncertainty

- **Epistemic**
 - The world is deterministic but we don't know it
 - What is the value of an already tossed coin hidden in my hand?
- **Stochastic**
 - The world is not deterministic
 - What will be the value of tossing coin?
- **In security both aspects are present**
 - Some attacks depends on the random layout of the memory → may not work all the time
 - Some attacks took place but we don't know it yet
- **Mostly we use a Bayesian Interpretation**
 - Probability is a subjective degree of belief that we update given some information

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What to believe?

- **Linda is**
 - 31 years old, single, outspoken, and very bright. She majored in philosophy. As a student she was deeply concerned with issues of discrimination and social justice, and also participated in anti-nuclear demonstrations.
- **Which is more likely?**
 - Linda is a bank teller
 - Linda is a feminist bank teller

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Where is the fallacy?

- **What description implies**
 - $\Pr(L \text{ feminist}) = \text{High}$
 - $\Pr(L \text{ feminist bank teller} \mid L \text{ is bank teller}) = \text{Very High}$
- **What the question was about**
 - $\Pr(L \text{ is feminist bank teller}) =$
 $\Pr(L \text{ feminist bank teller} \mid L \text{ is bank teller}) * \Pr(L \text{ is bank teller})$
- **Conditional vs Absolute Probability**

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How to protect?

- **Bomber pilots can**
 - carry either a flak jacket or a parachute because of weight limitations. The probability of being strafed by enemy guns is $\frac{3}{4}$ (requiring flak jacket to survive) the probability of plane being shot down is $\frac{1}{4}$ (requiring parachute to survive)
- **What is best?**
 - Flak jacket at all times
 - Parachute at all times
 - Flak jackets 3 times out of 4 and parachute 4th time
 - Flak jackets 1st time and parachute 3 out of 4 times

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The fallacy

- **Fallacy is chances don't repeat themselves**
 - The law of large numbers is actually working on large numbers...
 - UNLESS you really know this is a process that has memory (but then probability should be described differently)
- **Pilot taking flak jacket first three times**
 - Clearly has not been shot down on the fourth one
 - So he has seen the series “strafed;strafed;strafed” → next time it is going to be “shotDown”
 - By taking the parachute the fourth time he has $\frac{3}{4}$ chance to die, $\frac{1}{4}$ of survival
 - Strafing and shooting down are independent on the previous event

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Quantitative Risk Analysis - I

- Risk = Likelihood * Impact (negative)**

The diagram illustrates the flow of risk analysis. It starts with a 'Threat' box, which leads to a 'Vulnerability' box. Above 'Threat' is a red box labeled '#Bad Guys'. Above 'Vulnerability' is a red box labeled 'Pr(Compromised | Attacked)'. Above 'Incident' is a red box labeled 'Pr(Attack | Bad Guy)'. Above 'Impact' is a red box labeled 'Pr(Incident | Compromised)'. Below 'Threat' and 'Vulnerability' is a blue double-headed arrow labeled 'Likelihood of Bad Things Happening'. Below 'Incident' and 'Impact' is a blue double-headed arrow labeled 'Impact of Bad Things Happening'. Below 'Impact' are three green boxes: 'Direct Loss', 'Secondary Losses', and 'Cost to Restore'. Arrows indicate the flow from Threat to Vulnerability to Incident to Impact, and from the loss boxes up to the Impact box.

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Quantitative Risk Analysis - II


- Fix an interval of observation (say N years)**
- Benefit = + Likelihood*Impact - NewLikelihood*NewImpact**
- Value = + Benefit - Cost for NewLikelihood - Cost for NewImpact**

The diagram shows the same flow as slide I, but with mitigation strategies. Below 'Threat' is a green box 'Remove Threats'. Below 'Vulnerability' is a green box 'Remove Vulnerabilities'. Below 'Incident' is a green box 'Reduce Likelihood'. Below 'Impact' is a green box 'Remove Impact'. Below the flow are four green boxes: 'Reduce Opportunity' (under Threat/Vulnerability), 'Reduce Likelihood' (under Vulnerability/Incident), 'Reduce Impact' (under Incident/Impact), and 'Recover from Impact' (under Impact). Above the flow is a blue double-headed arrow labeled 'Cost To Reduce Likelihood' and another labeled 'Cost to Reduce Impact'. Arrows point from the mitigation boxes up to the corresponding stages in the flow.

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What we need to estimate

- **#Threats**
 - Intentions to attack by cyber-terrorist, financially motivated criminals, hacktivists, disgruntled employees, etc.
- **Pr(Attack|Threat)**
 - If a given threat is active how many attacks are we going to get?
- **Pr(Compromise|Attack)**
 - Once we are attacked would this generate an actual compromise of the machine (so the exploit would actually work)
- **Pr(Incident|Compromise)**
 - Once we have been exploited, has this exploit been transformed into an incident that has a specific cost?
- **May not be possible to estimate everything individually**
 - E.g. some probability might be difficult to disentangle from actual data

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Example Verizon DBiR

- **Verizon Reports**
 - #Number of Incidents x Victim Type
 - #Number of Data Breaches x Victim Type
 - #Typology of attacks
- **Example in 2015**
 - Retail 370 incidents, 182 breaches
 - Professional services 916 incidents, 53 breaches
 - Cyberspies 247 incidents, web app attack 5334
- **What can be calculated?**

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What we have

- **#Threats** → *don't know*
- **$Pr(\text{Attacks}|\text{Threats}) * \text{Threats}$** → *"incidents" in Verizon terminology*
- **Exploited Vulnerabilities** → *don't know*
 - From the data I report here, the information of exploited attack is actually there in the DB, they only tell us the gross totals in this report
- **$Pr(\text{Incidents}|\text{Compromise}) * Pr(\text{Compromise}|\text{Attack}) * Pr(\text{Attack}|\text{Threats}) * \text{Threats}$** → *"data losses"*
- **Can reconstruct**
 - $Pr(\text{Incidents}|\text{Threats}) = \text{"Verizon data losses"} / \text{"Verizon's incidents"}$

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Global Data of Breach Likelihood

Distribution of Likelihood of Breaches

Probability to meet this likelihood of breach in practice

Likelihood of breaches

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Data on Three Industries

- **Average Attacks**
 - Personal Services (Finance, health) = 435.2
 - Physical Production (Agriculture etc) = 8.5
 - Industries (Utilities, Wholesale, etc.) = 112.1
- **Average Probability of Success**
 - Personal Services (Finance, health) = 0.41
 - Physical Production (Agriculture etc) = 0.62
 - Industries (Utilities, Wholesale, etc.) = 0.39

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Also distribution and type matter


Probability of Successful Breaches by Type of Industry

Incident Index	Personal Services (Finance, Healthcare etc.)	Physical Industries (Mining, Agriculture, etc.)	Industry (Utilities, Wholesale, etc.)
1	0.05	0.25	0.15
2	0.10	0.45	0.22
3	0.20	0.45	0.25
4	0.25	0.50	0.28
5	0.30	0.50	0.30
6	0.42	0.65	0.38
7	0.45	0.68	0.42
8	0.55	0.70	0.45
9	0.58	0.78	0.48
10	0.60	0.90	0.48
11	0.65	0.95	0.50
12	0.70	1.00	0.55

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Scale down to the company

- **What are we missing?**
 - We don't know how many industries are in the sample by Verizon.
 - We must investigate that into the database
 - 200 Attacks over 100 Companies means 2 attacks against your company per year.
- **Compute the final Likelihood**
 - $\text{Avg(Attacks)} * \text{Avg(Prob of Breaches)} / \text{Firms}$
- **Multiply by impact → (average) risk**

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Refining the analysis

- **Exploited Vulnerabilities**
 - If we have access to the data we can use this information to estimate the effect of countermeasures
- **$\text{Pr(Compromise|Attack)} \rightarrow \text{Pr(Compromise|Attack \& CVSS=x)}$**
 - How many vulnerabilities with a given CVSS score have been attacked
 - How many of them has been the cause of a data breach?
 - If we remove the vulnerabilities with highest probability → reduce likelihood
- **Approximate calculation also possible**
 - Assume that vuln with CVSS=10 yields a compromise with $\text{Pr}=1$
 - Conservative but may be an overkill

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What About Extreme Risks?

- ***So far we calculated averages of success***
 - Personal Services (Finance, health) = 0.41
 - Physical Production (Agriculture etc) = 0.62
 - Industries (Utilities, Wholesale, etc.) = 0.39
- ***Which is the 90-percentile of success for breaches?***
 - Personal Services (Finance, health) = 0.58
 - Physical Production (Agriculture etc) = **0.95**
 - Industries (Utilities, Wholesale, etc.) = 0.66

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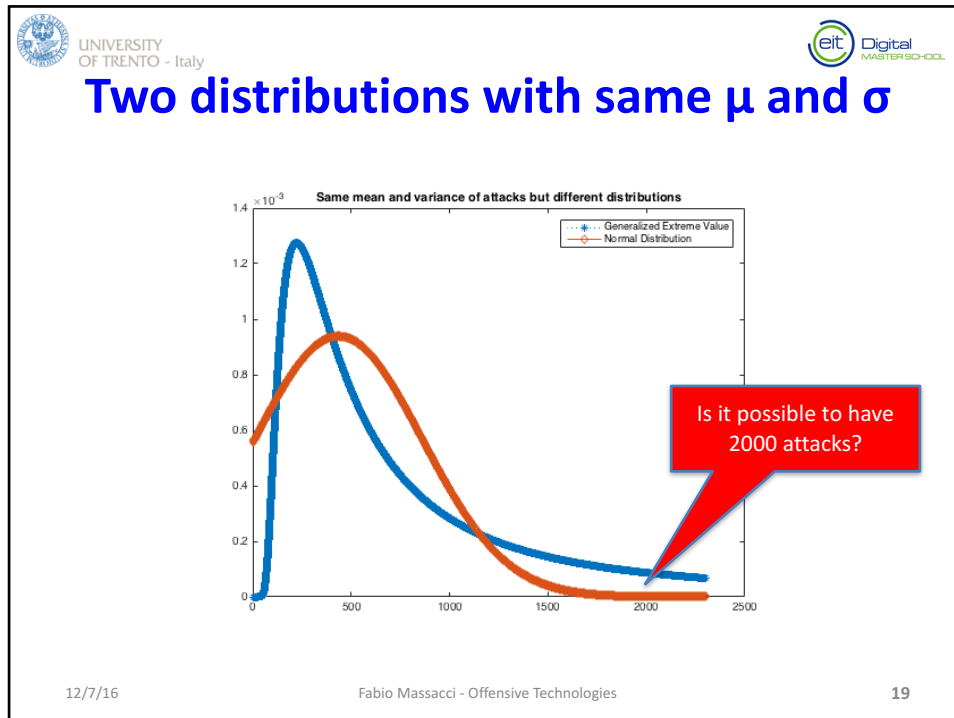
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The Problem

- ***With Normal/Poisson distributions***
 - as we we go away from the average → there are very few dangerous cases
 - So we can reasonably use the average, at most moderate with the standard deviation
- ***Our data is not normal***
 - we have very fat tails of the distribution → dangerous cases may not be so few
 - Extreme risks may be less rare than we thought
- ***We need to estimate “worst cases”***

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Estimating Extreme Risks

- **Simplified version of Basel-II approach**
 - Used by banks to estimate rare but anyhow big risks
 - Banks do a double convolution: estimate likelihood and estimate losses \rightarrow we take impact as given
- **Approach**
 - Collect data of value of the variable of interests (attack, percentage of success, vulnerabilities, etc.)
 - Fit data into a distribution (try both slim and heavy tails)
 - Calculate at the alpha-percentile of the distribution
 - This is the number we use for the calculation of the final risk


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Poisson Distribution

- **Key Idea**
 - Probability that n incidents will happen in a give time interval → decreases linearly with the size
 - $\Pr(k) = \Pr(k-1) * \lambda/k$
- **Typically very thin tail**
 - Large number of incidents are very very rare
- **Cumulative Distribution**
 - $\Pr(X < x) = e^{-\lambda} \sum_{i=0}^x \lambda^i / i!$
- **Parameter estimation from data**
 - $\lambda = 1/n \sum_{j=0}^n x_j$

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Pareto Distribution

- **Key Idea**
 - Power Law for distribution of income
 - The people with a (large) pot of money m are progressively fewer and fewer i.e. they are only a/m^b
 - Used to model large losses (m) in property and liability insurance → the larger the b the more likely we have people with large losses
 - Typical values of b for earthquakes (1), fire industry (1.5), general liability (1.8), occupational injuries (2), motor liability (2.5)
- **Cumulative Distribution**
 - $\Pr(X < x) = 1 - (a/x)^b$
- **Parameters estimation from data**
 - $a = \min(x_j)$
 - $b = n [\sum_{j=1}^n \log(x_j/a)]^{-1}$

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Generalized Extreme Value Distribution

- **Key Idea**
 - Try to captures the possible maxima (or minima) of a batch of random values
 - If the tail goes esponentially to zero → collapse to normal/Poisson distribution
 - If the tail goes polynomially to zero → Student's t distribution or Frechet's distribution
 - If the tail is bounded → Beta distribution
- **E.g. Cumulative Distribution (Frechet)**
 - $\Pr(X < x) = e^{-b/(x-a)^C}$

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Estremal Values for Likelihood

- **Data**
 - Use the “Incidents” in Verizon DBiR terminology
- **Goal**
 - We want to know the worst possible number of attacks, at 95% percentile for different type of small firms
 - Banks have to calculate at the 99.9% (but we don't have enough data here)
- **Process**
 - Compare Distributions
 - Actual (the empirical distribution), Poisson, Generalised Extreme Value, Pareto Tails
 - Find best distribution
 - We do this “visually”, should be done with statistical tests → advanced courses
 - Return the inverse value of the 95% percentile

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Extremal Number of Attacks


95%	Administr.	Consumers	Industry	Personal	Production
Empirical	26	179	18	50*	4
Normal (fit)	24	164*	13	50	3
Poisson	15	80	9	34	3**
GEV	30**	374*	16**	50*	1343
ParetoTails	24	169	17	49	4

Starred nodes correspond to the distributions that seem to fit best (from the plots)

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Further reading

- **Chapters 10, 11 on Textbook**
- **Chapters 1-3, Claudio Franzetti, “Operational Risk Modelling and Management”, CRC Press**
- *Ross Anderson’s book*
- *L. Allodi, F. Massacci. Comparing vulnerability severity and exploits using case-control studies. ACM Trans. on Information and System Security, 17(1):1 (2014).*

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