

Why Do Nigerian Scammers Say They Are From Nigeria?

Cormac Herley

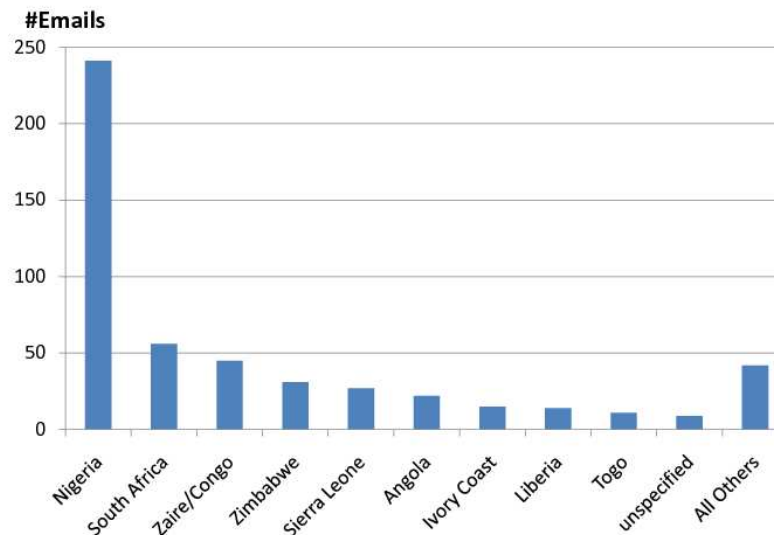
Microsoft Research, Redmond

I NEED YOUR URGENT RESPOND.
FROM Mr.Kuso Acho.
The Head of File and Auditing Department, BANK OF AFRICA (B.O.A) Ouagado
(West Africa) REMITTANCE OF US\$20, 5;MILLION CONFIDENTIAL IS THE CAS
ATTENTION
This message might meet you in utmost surprise, however, it's just my U
ner that made me to contact you for this transaction I am a banker t
aso in West Africa and currently holding the post of director A
of transferring the left over Funds (\$20.5 m
his entire family in a plane crash.
this transaction will be
according to the per
receive my
dersta

Nigerian Emails:

Who falls for these things?

- What's with the spelling mistakes, BLOCK CAPS?
- Why not Sweden, or Bolivia or New Jersey?
- Who hasn't heard of [Nigerian Scam?](#)



How does economic opportunity change as a function of victim density?

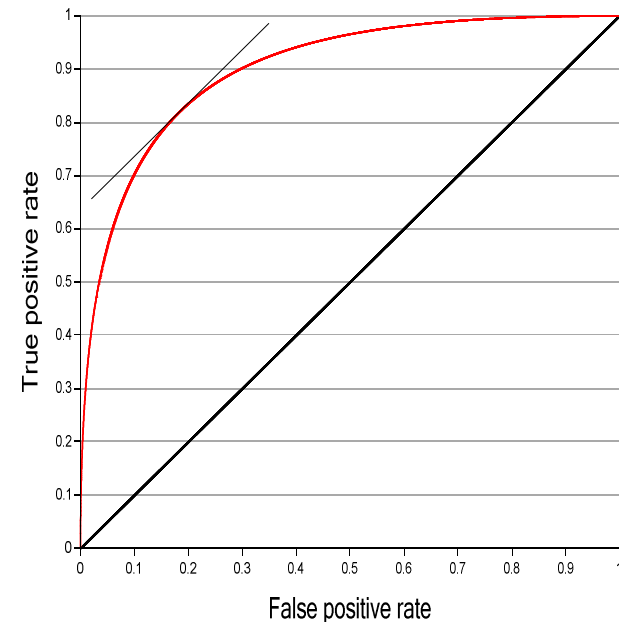
- Population with density d viable victims

$$\text{Opportunity} = d G N$$

- What if we reduce density by $2x$?
 - Does opportunity fall $2x$, $> 2x$, $< 2x$?
- *Opportunity drops much faster than density*

Attackers have False Positives too

	Attack	Don't Attack
Viable	TP	FN
Non-viable	FP	TN



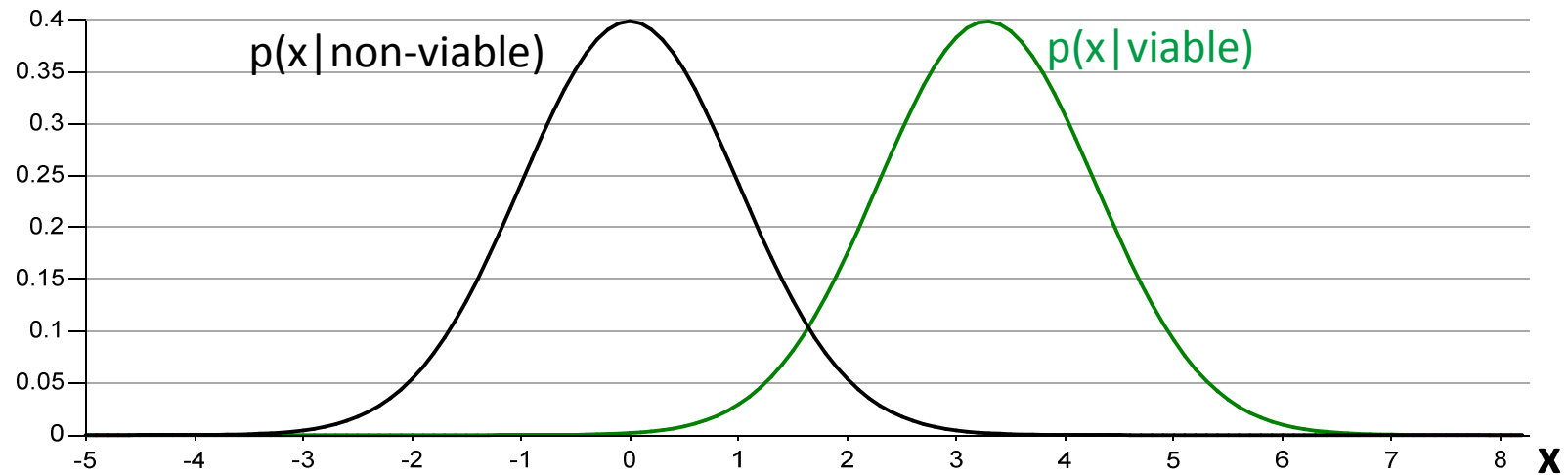
Expected Return:

$$E\{R\} = d t_p G N - (1-d) f_p C N$$

d = victim density
 N = population
 t_p = true pos. rate
 f_p = false pos. rate
 G = net gain (success)
 C = Cost (fail)

Viability is not observable

- Encapsulate all *observable* info about viability in x

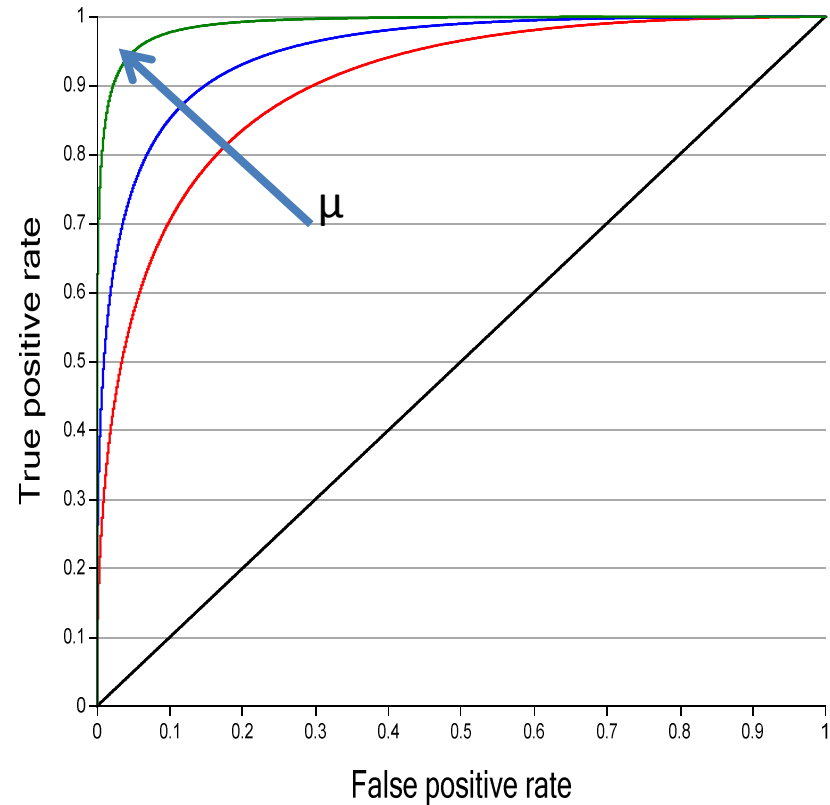


- Model:
 - $p(x|\text{non-viable}) \sim N(0,1)$
 - $p(x|\text{viable}) \sim N(\mu,1)$
- ***Rich does not mean viable!***

ROC Curves:

true positives vs. false positives

- Increasing μ gives better detection
 1. Curve is monotonic
 2. Slope decreases monotonically
 3. Area Under Curve (AUC) = Prob. viable ranked higher than non-viable



AUC = 0.99
AUC = 0.95
AUC = 0.9

Optimal Operating Point (OOP)

Expected Return:

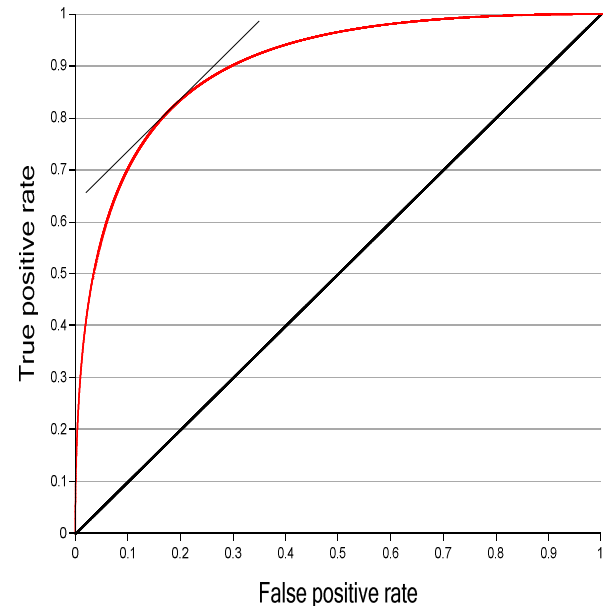
$$E\{R\} = d t_p G N - (1-d) f_p C N$$

Maximize $E\{R\}$:

$$\frac{dt_p}{df_p} = \frac{1-d}{d} \frac{C}{G}$$

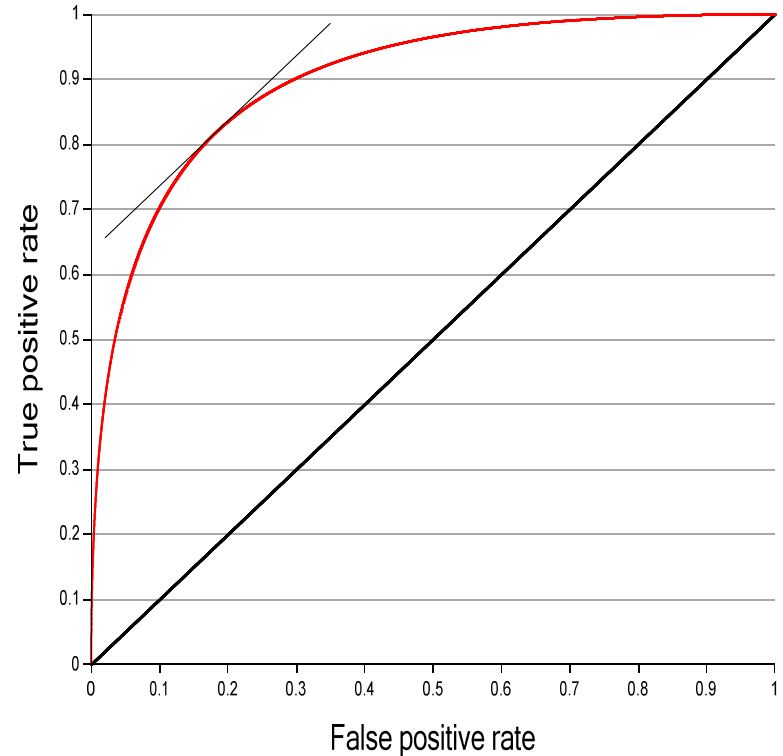
Slope at OOP:

- Inversely proportional to d
- Inversely proportional to G/C

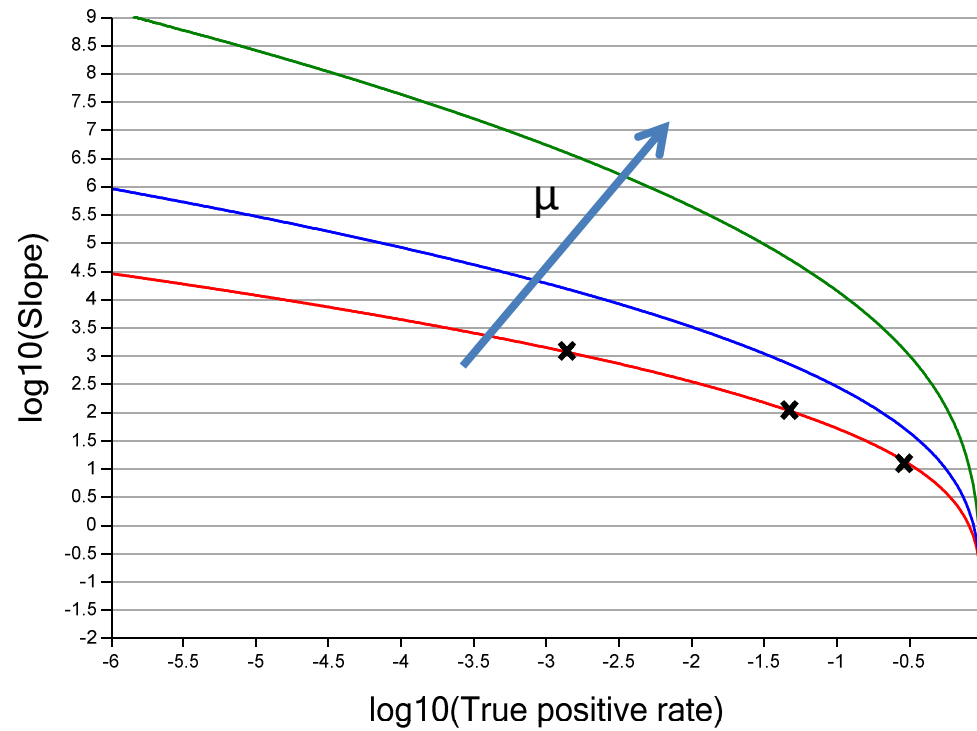


Consequences

- ***As $d \rightarrow 0$ slope at OOP increases***
 - Slope at OOP: $\frac{dt_p}{df_p} = \frac{1-d}{d} \frac{C}{G}$
- ***As slope increases fewer attacked***
 - Viable users attacked = $d t_p N$
 - Non-viable attacked = $(1-d) f_p N$
- ***Opportunities with lower densities harder to exploit***



Slope vs t_p



Example slope of:

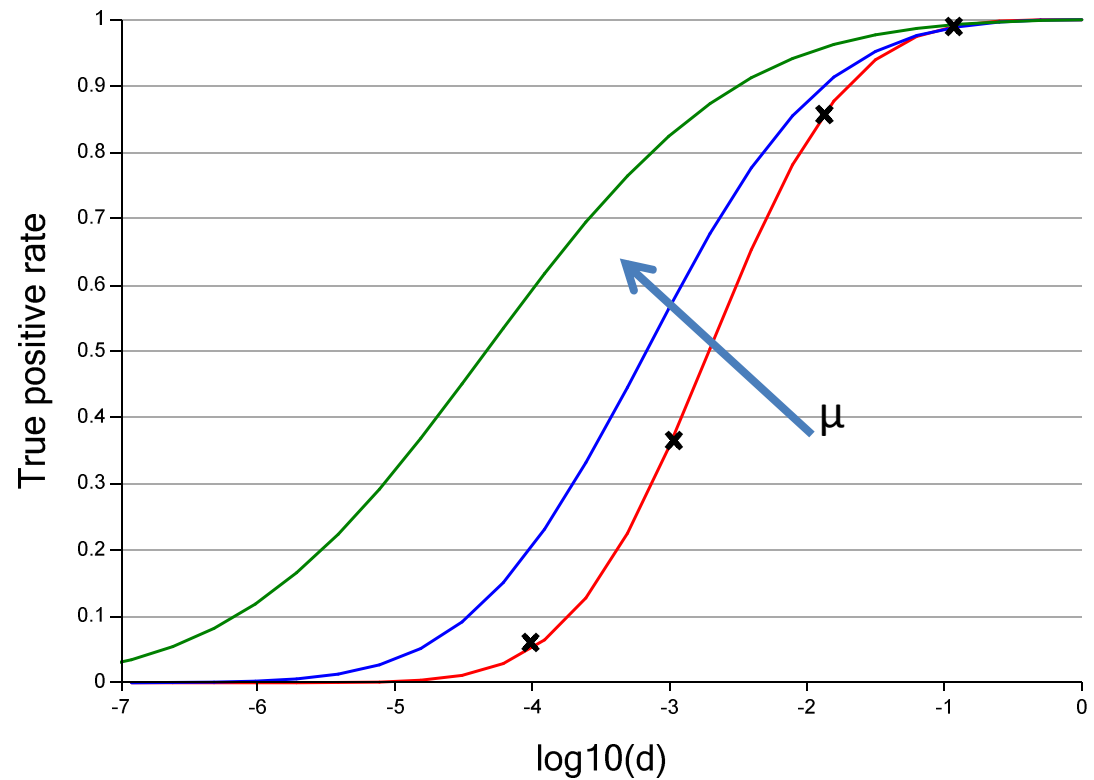
- $10 \Rightarrow t_p = 0.36$
- $100 \Rightarrow t_p = 0.05$
- $1000 \Rightarrow t_p = 0.0019$

True Positive rate vs victim density

t_p vs. d (at $G/C = 100$)

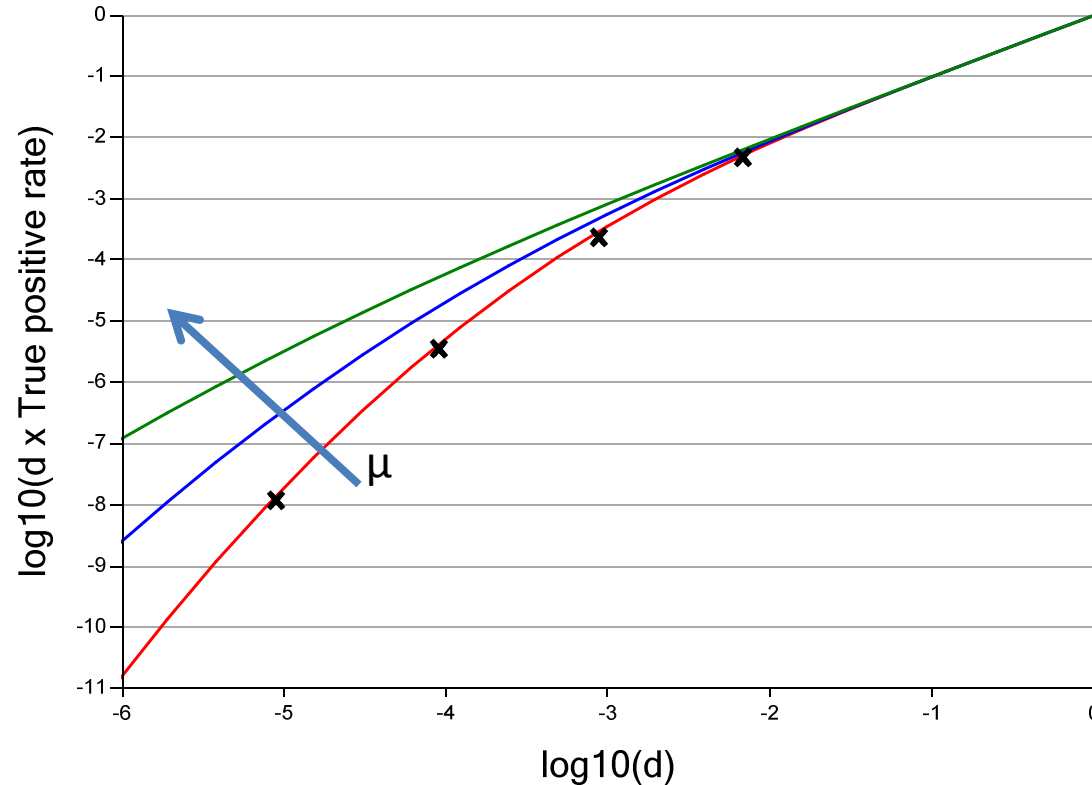
If fix G/C can plot t_p vs d

$$\frac{dt_p}{df_p} = \frac{1-d}{d} \frac{C}{G}$$



- ***True positive rate falls fast with density***

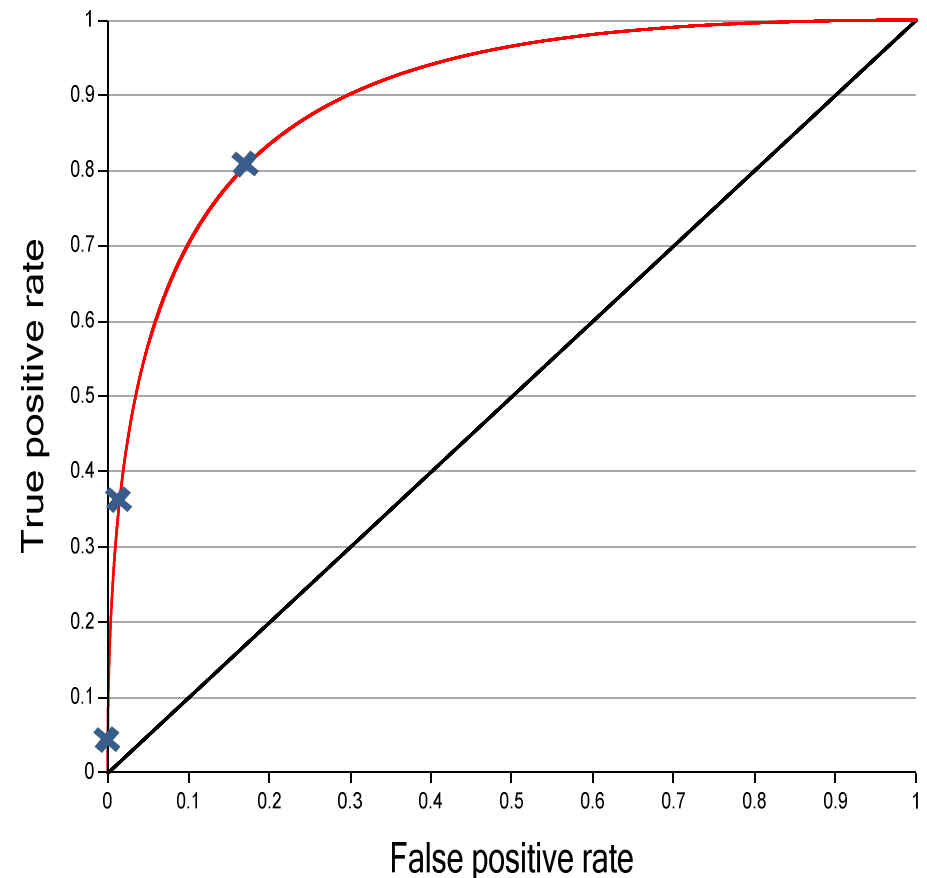
Victims found vs density (G/C=100)



- **Victims found falls *much* faster than density**
 - $d = 10^{-5}$ gives $d t_p = 10^{-8}$
 - E.g. Population 200 million, 2000 viable, 2 profitably victimized

Diversity is more important than strength

- Suppose d , G/C s.t. slope = 1
 - $t_p = 0.82$, $f_p = 0.18$
- Now divide into 10 attacks with density $d/10$
 - $t_p = 0.36$, $f_p = 0.015$
- Thus, no change in #viable targets, or G/C
 - Viable users attacked drops 2x

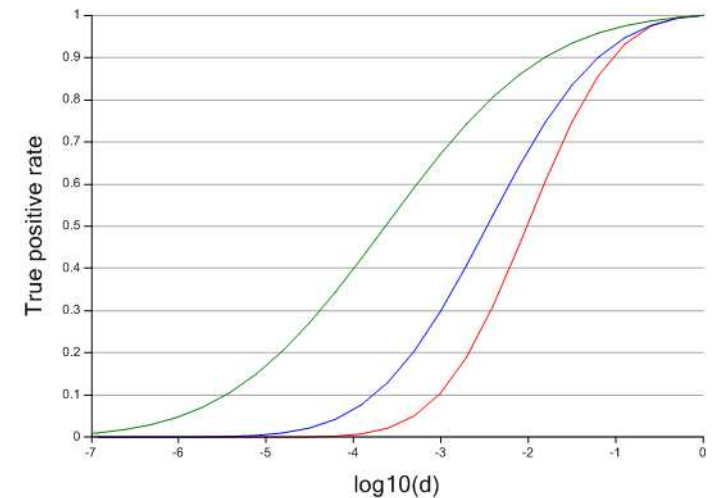


Everyone vulnerable, no-one attacked?

- Divide population into Q attacks each with d/Q
- Viable victims attacked:

$$\frac{d}{Q} \sum_{k=1}^Q t_p(d/Q) N = d t_p(d/Q) N$$

- $t_p(d/Q) \rightarrow 0$
- $d=1$ (everyone vulnerable)
- Q large (no-one attacked)



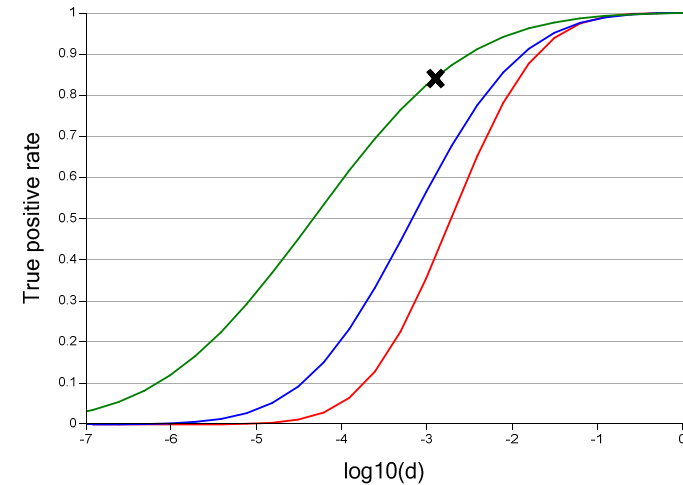
Optimism does not pay

- Attacker thinks:
 - $d = 10^{-3}$
 - $G/C = 100$
 - $AUC = 0.99$
- But attacker gets:
 - $d = 10^{-4}$
 - $G/C = 20$
 - $AUC = 0.9$

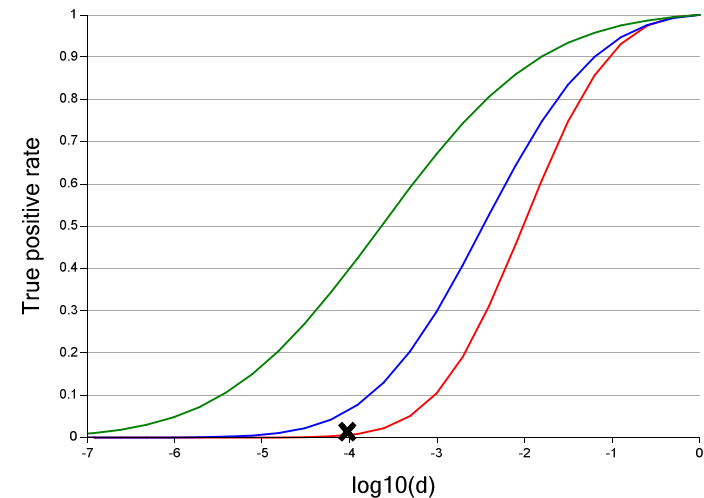
Factor difference:

$$\frac{10^{-3} \times 0.826 \times 100}{10^{-4} \times 0.006 \times 20} = 6883$$

$G/C = 100$



$G/C = 20$



Three factors affect return

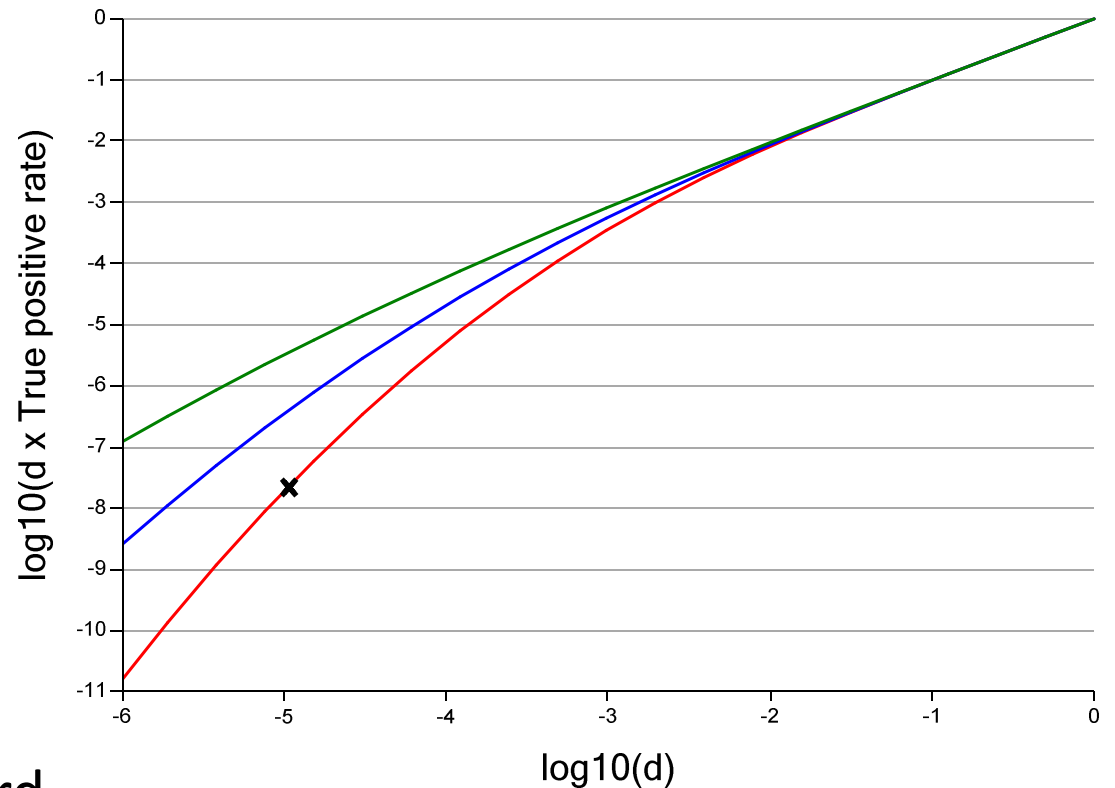
- ***Density d***
- ***Profitability G/C***
- ***Ability to detect (i.e. ROC curve)***

How about $G/C = \text{“A Gazillion”}$?

- At OOP return is like $(G/C+1)$ sided coin
 - Gain G with prob C/G
 - Otherwise lose C
- Variance \approx Mean
- Suppose $G/C = 1000$
 - $\text{binocdf}(0,1000,0.001) = 36.8\%$ chance no victim after 1000 attacks.
- Keynesian attacker is more conservative
 - Slope = $\frac{1-d}{d} \frac{1}{k_{max}}$
 - where $k_{max} = \text{max avg between hits}$

How about Classifier Accuracy = 99.999%

- How did you get to be so good?
- How learn to distinguish viable from non- when viable are so rare?
- Need many samples of each for training
- Iterative improvement hard



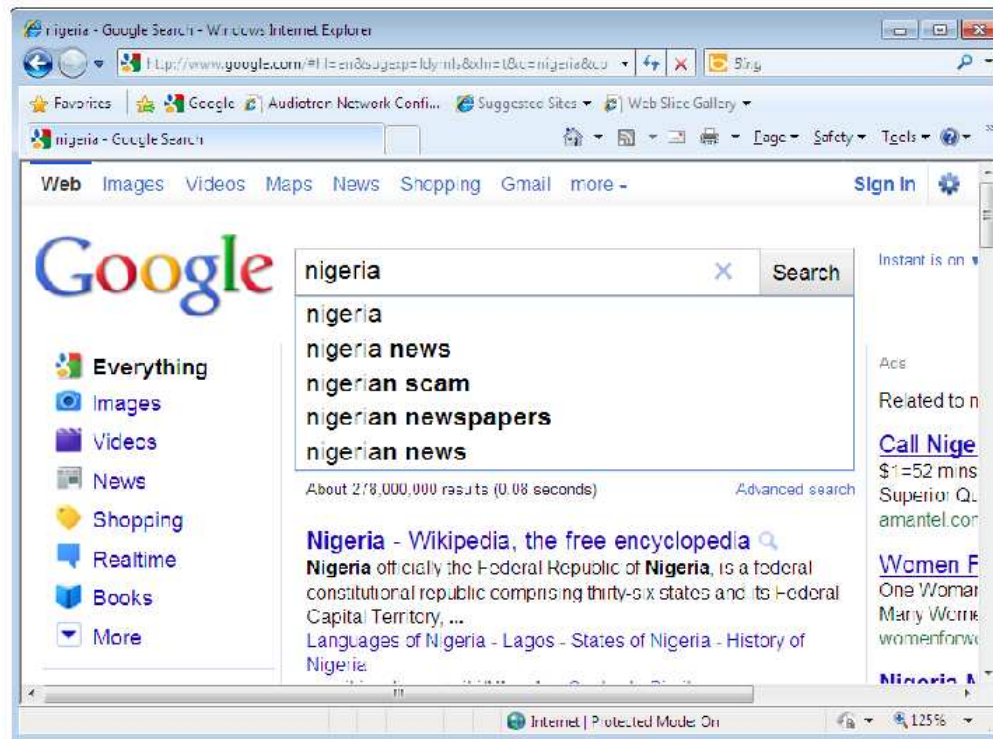
Catch-22 at low densities

- **Need to find them to learn how they can be found**
 - Must distinguish viable from non-viable with great accuracy
 - Need many viable samples to learn to distinguish.

Nigerian Emails:

Who falls for these things?

- Who hasn't heard of [Nigerian Scam](#)?



- ***Ideally: attack only those who haven't heard of it.***

Why do Nigerian Scammers say they are from Nigeria?

- *Initial email has cost ≈ 0 per user*
- *Follow-up has cost > 0*
 - *Detector = wording of initial email*
- *When d is low*
 - *Only small fraction of vulnerable can be found*
 - *Repelling FP's more important than finding TP's*

Conclusions

- ***Economic Opportunity falls far faster than victim density***
- ***Extreme difficulty for low density attacks***