

Detecting Stuffing of a User's Credentials at Her Own Accounts

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University of North Carolina at Chapel Hill

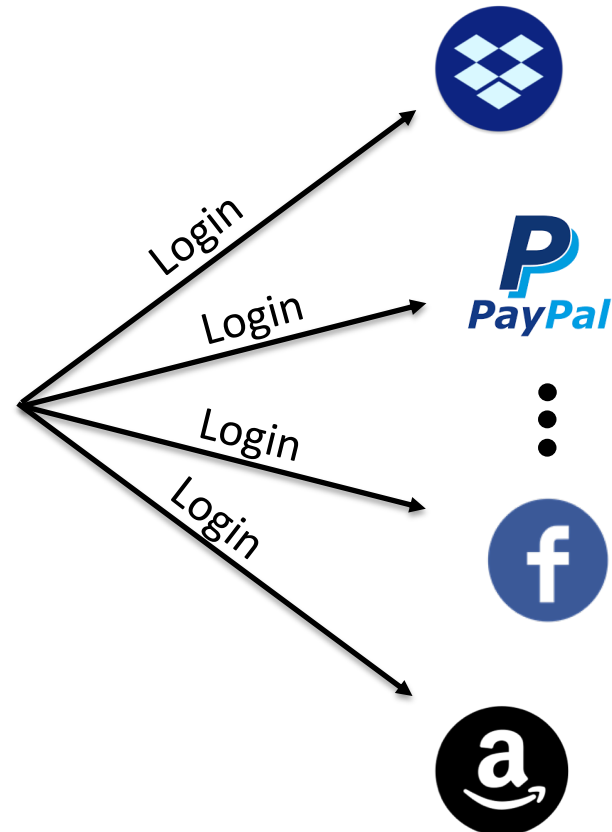


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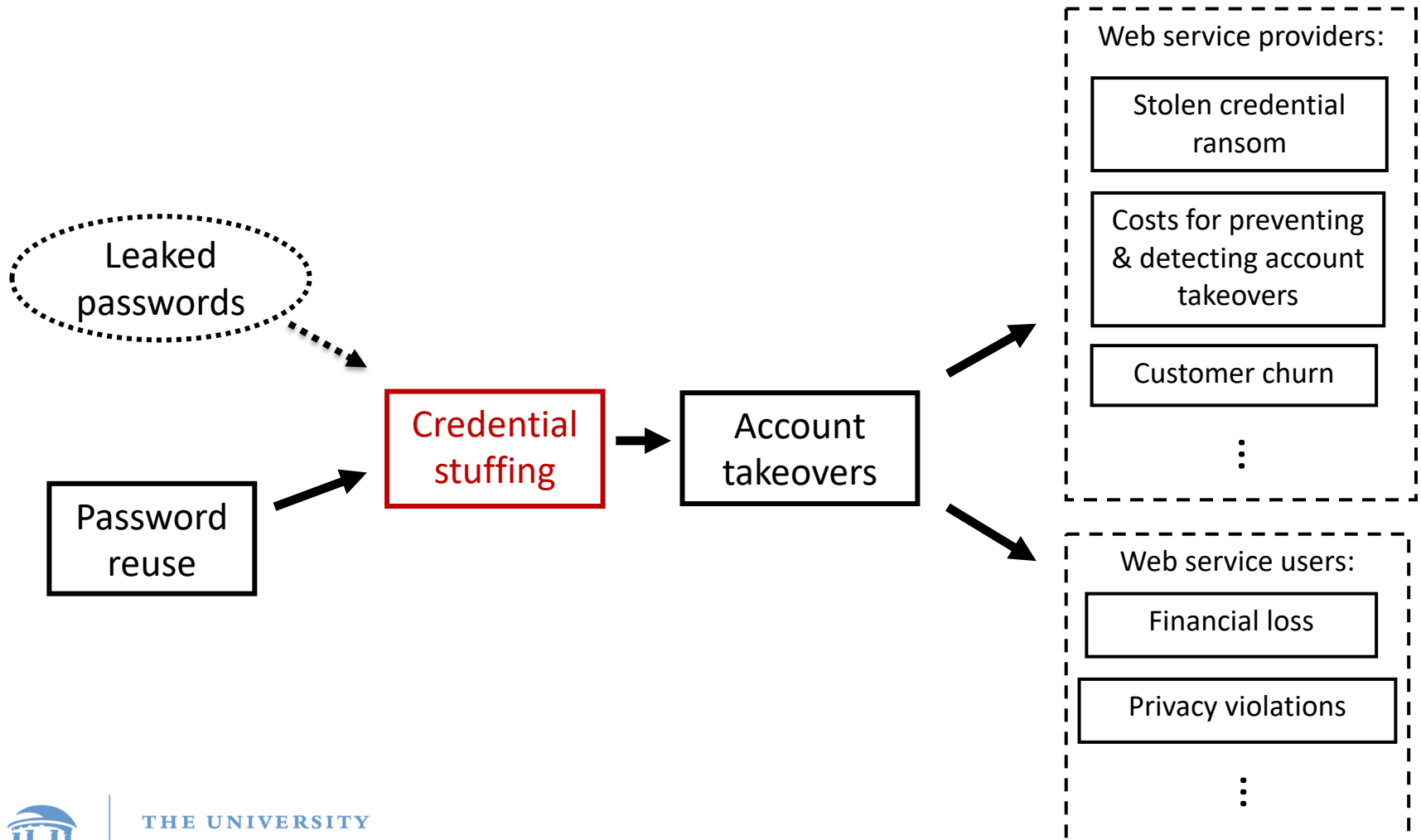
Credential Stuffing

Database breaches,
phishing, malware,
social engineering,
etc.

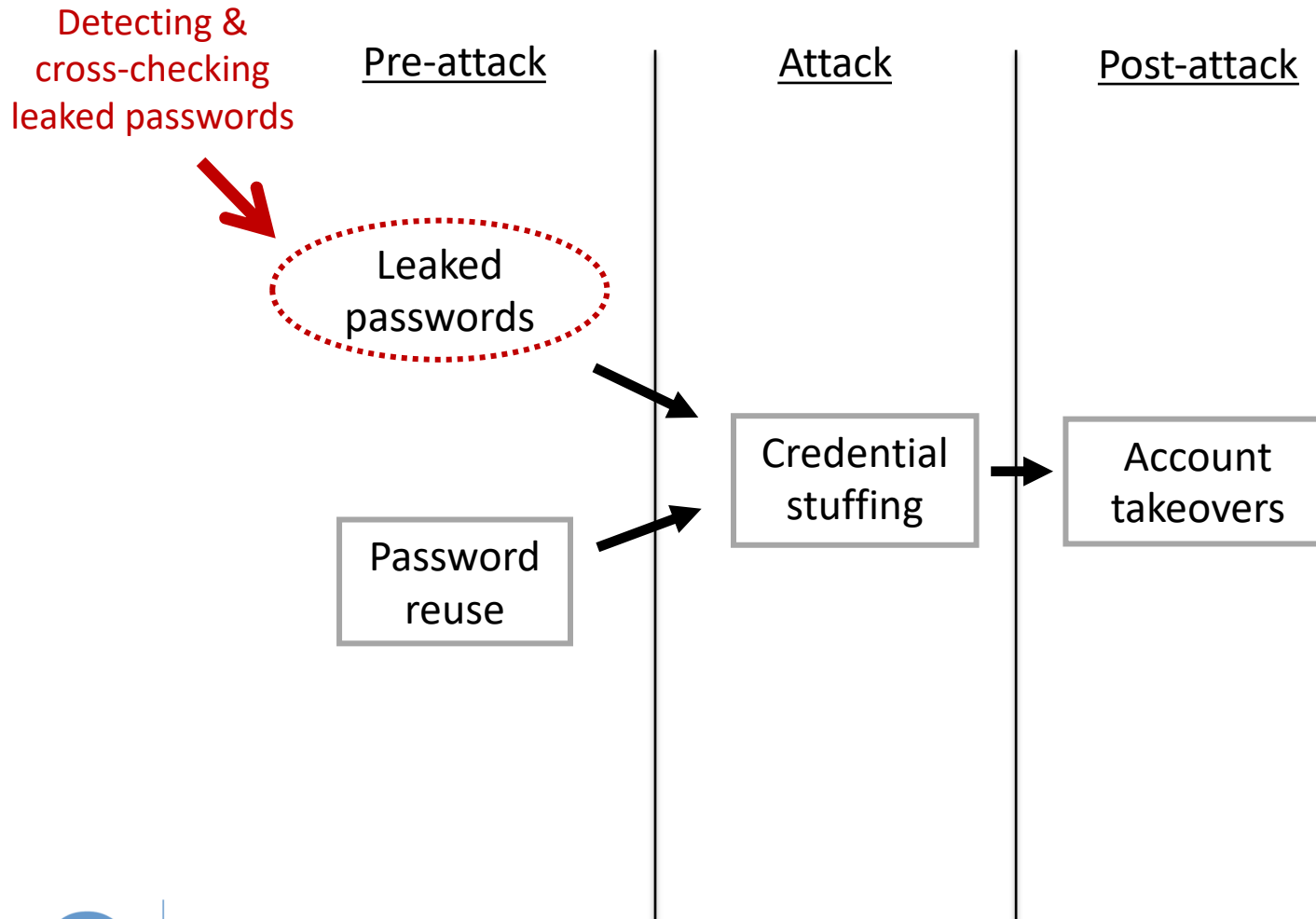
Valid user ID
password pairs



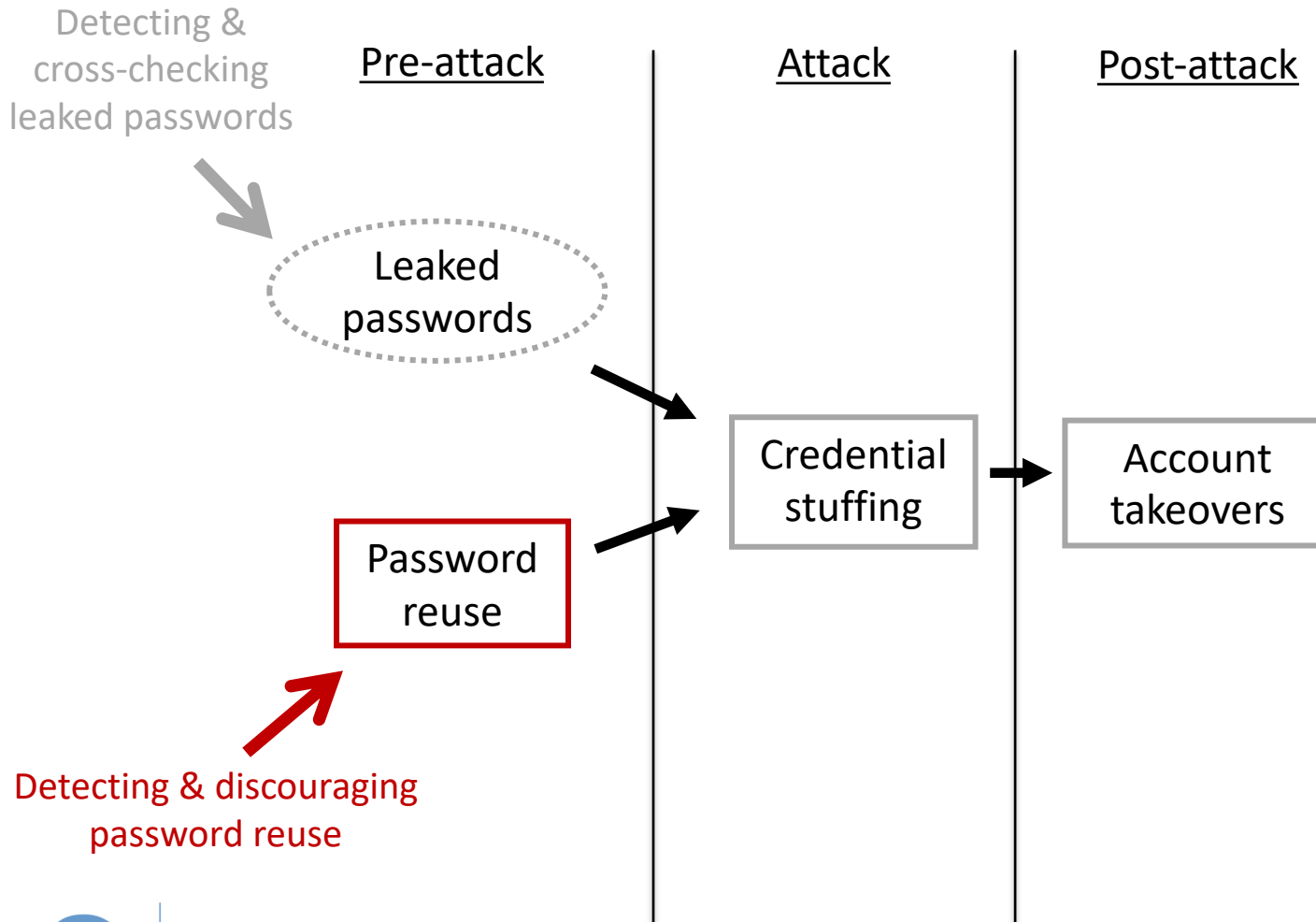
Harm of Credential Stuffing



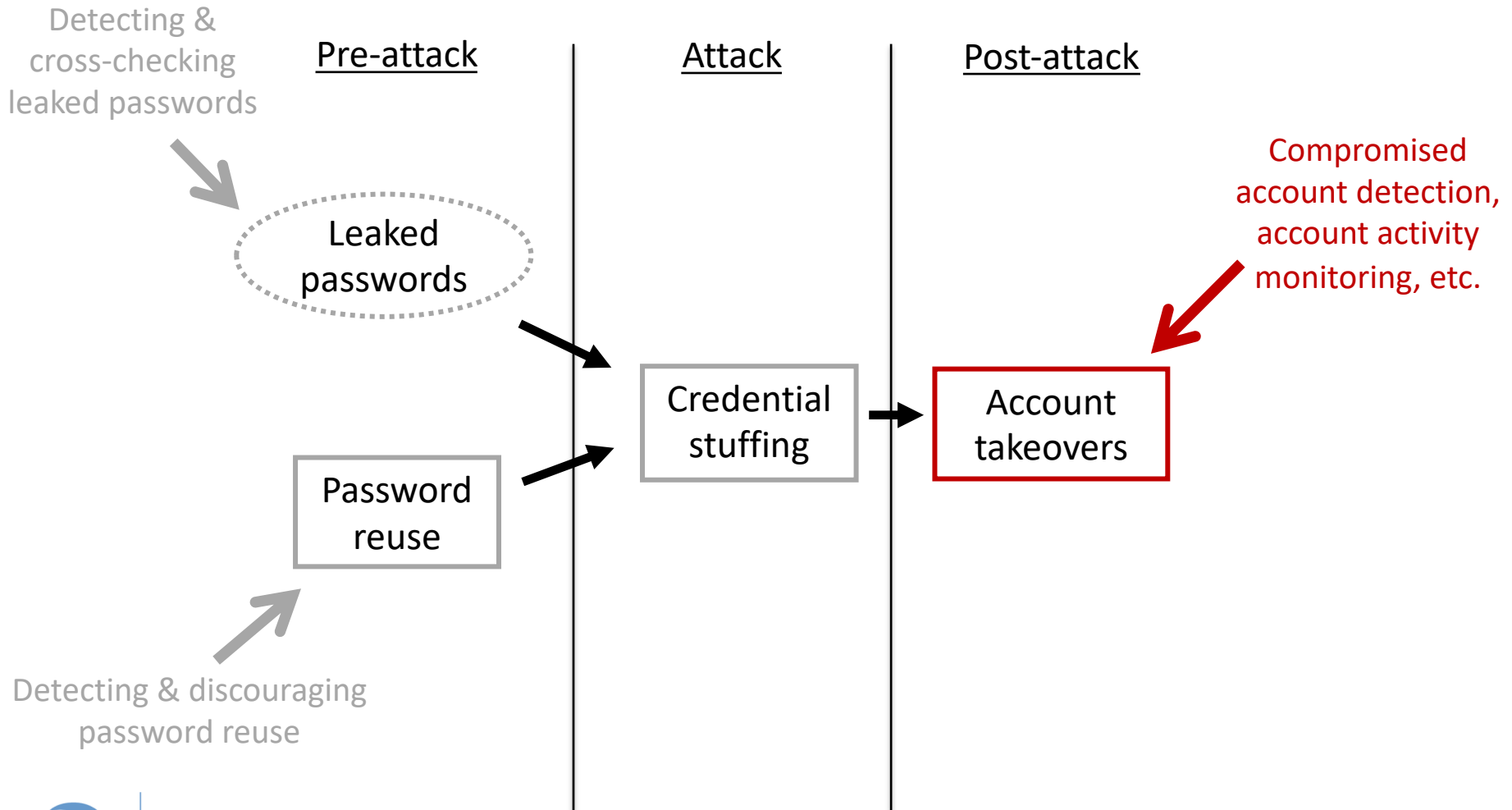
Existing Approaches



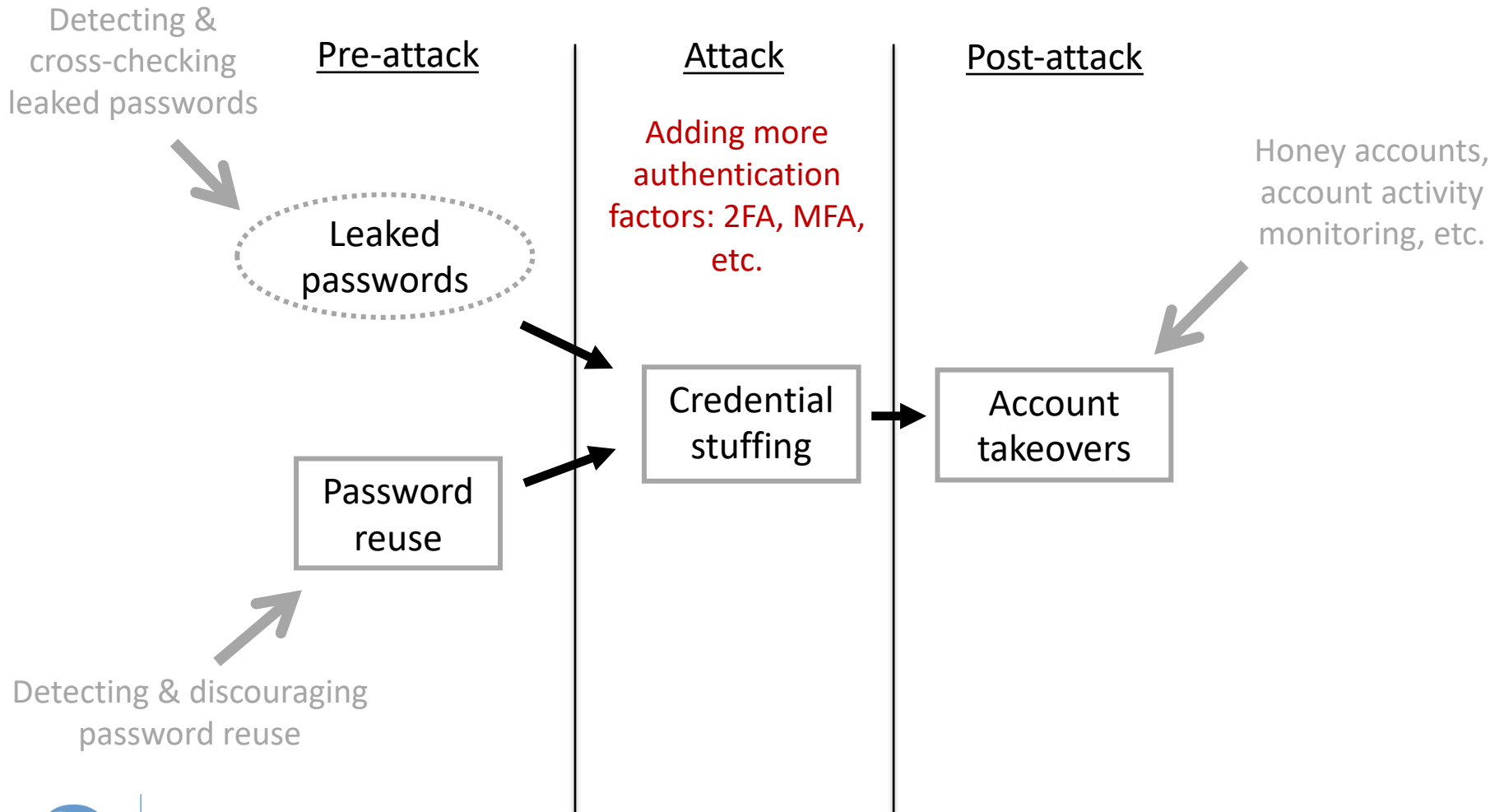
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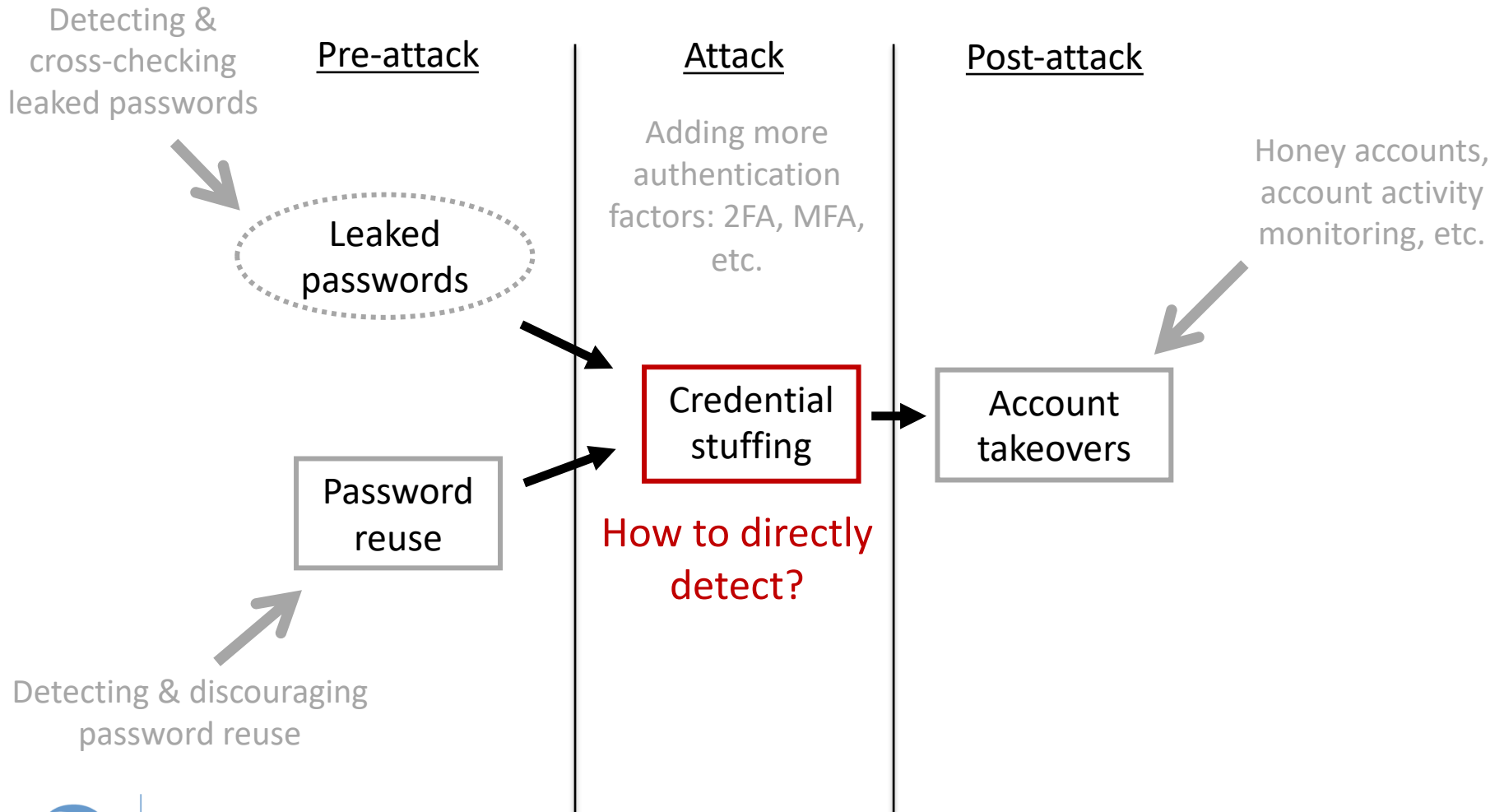
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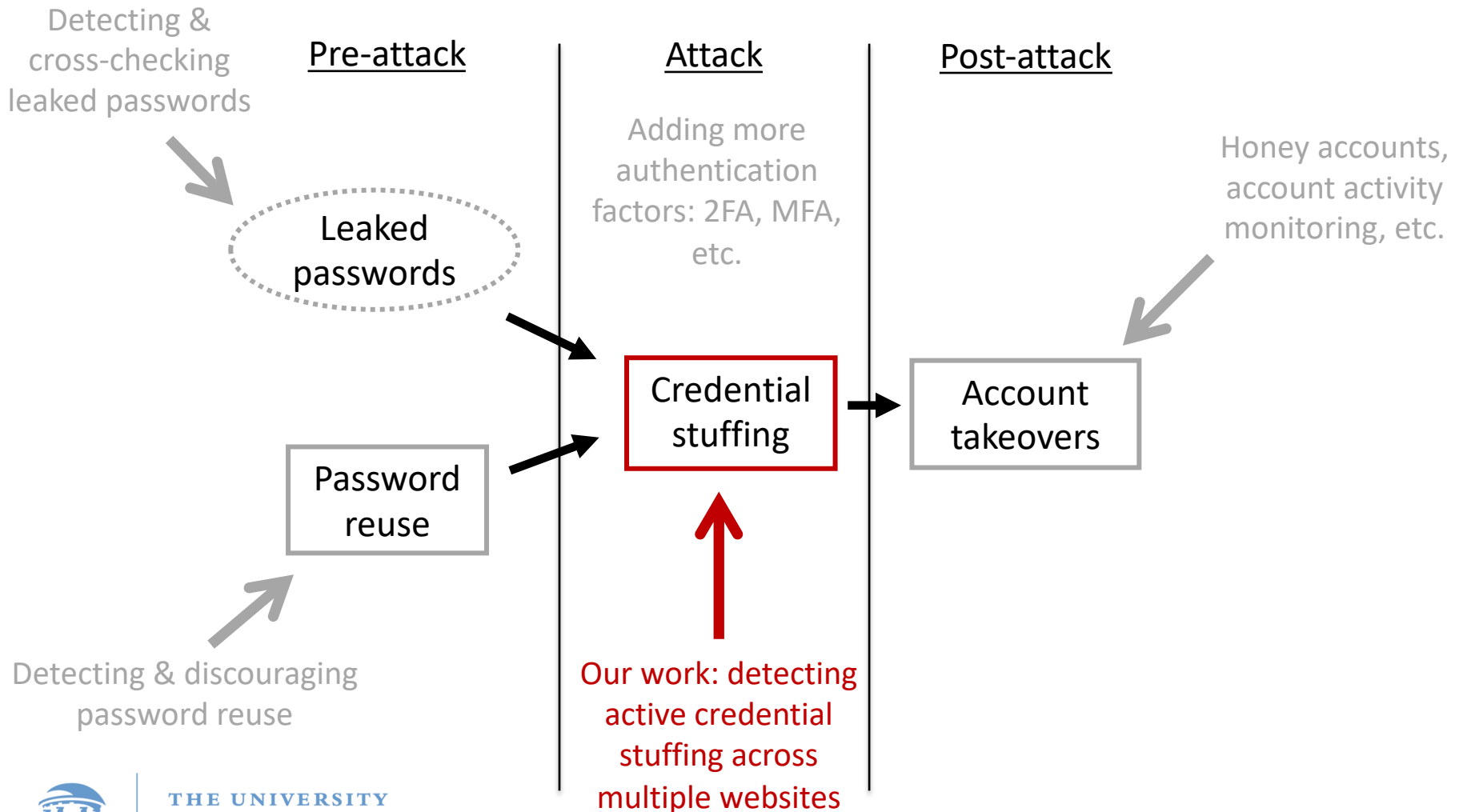
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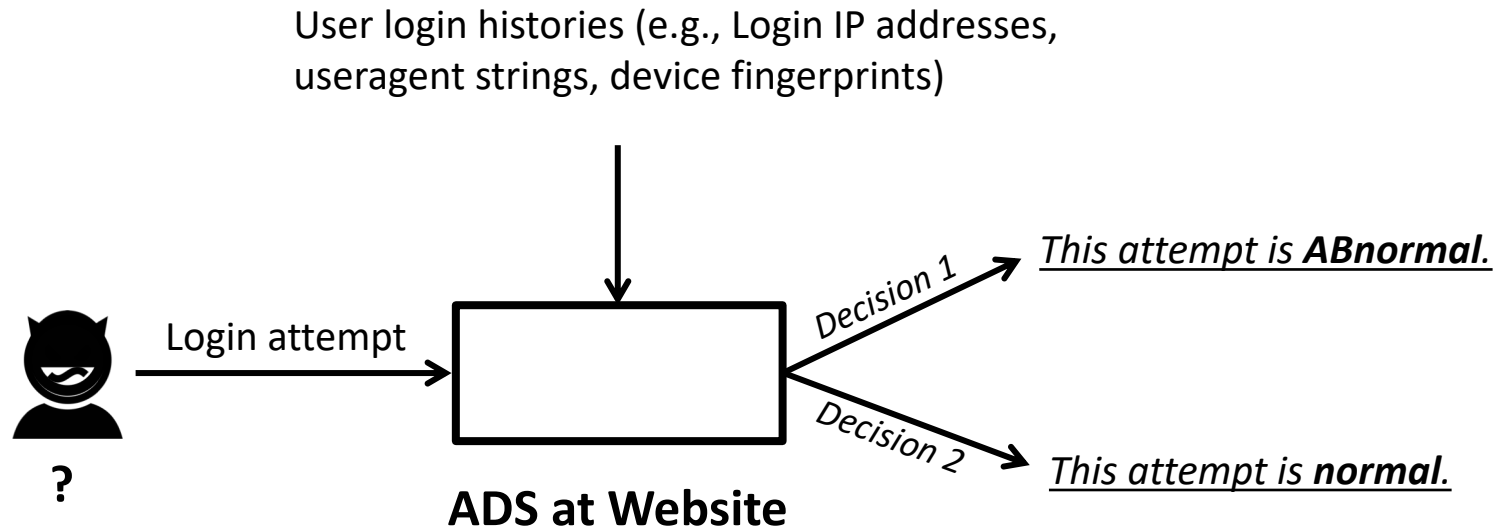
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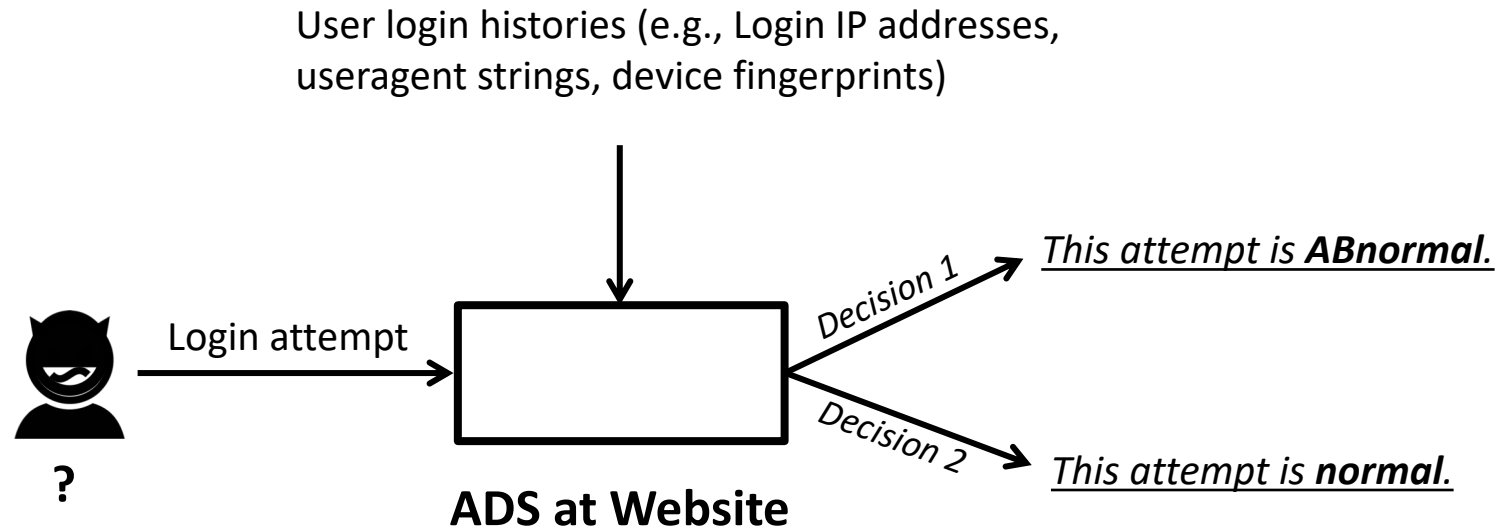
Our Work



Anomaly Detection Systems (ADS)



Anomaly Detection Systems (ADS)

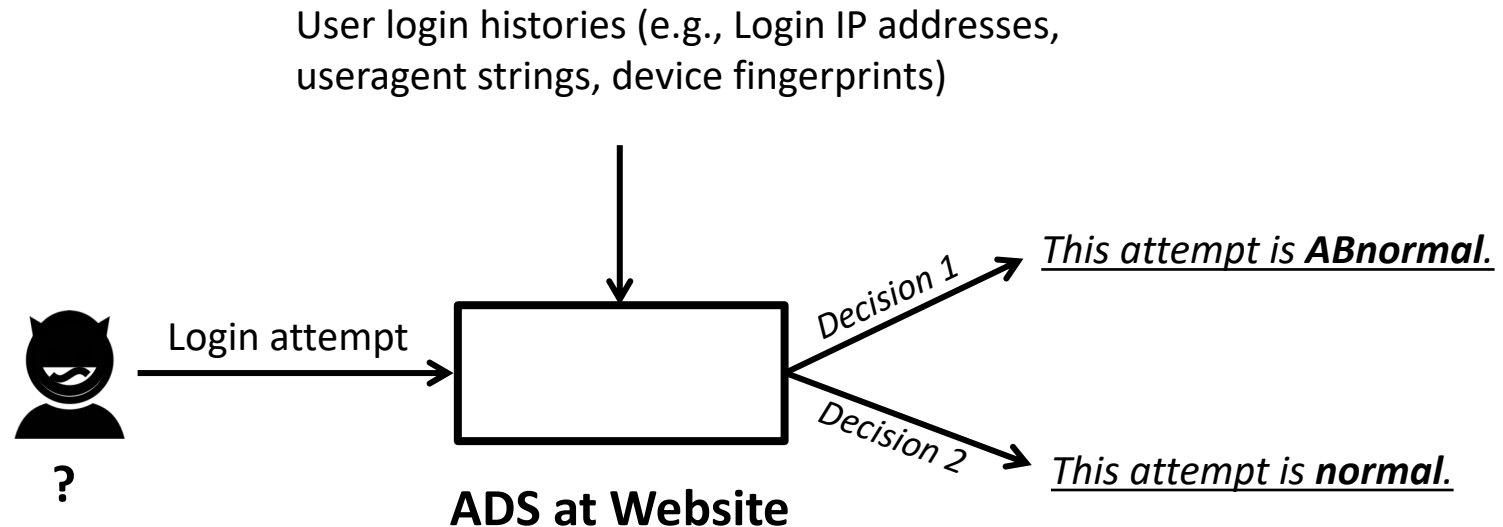


Naïve ADS:

- *Strange IPs = "abnormal"*
- *Strange devices = "abnormal"*



Anomaly Detection Systems (ADS)

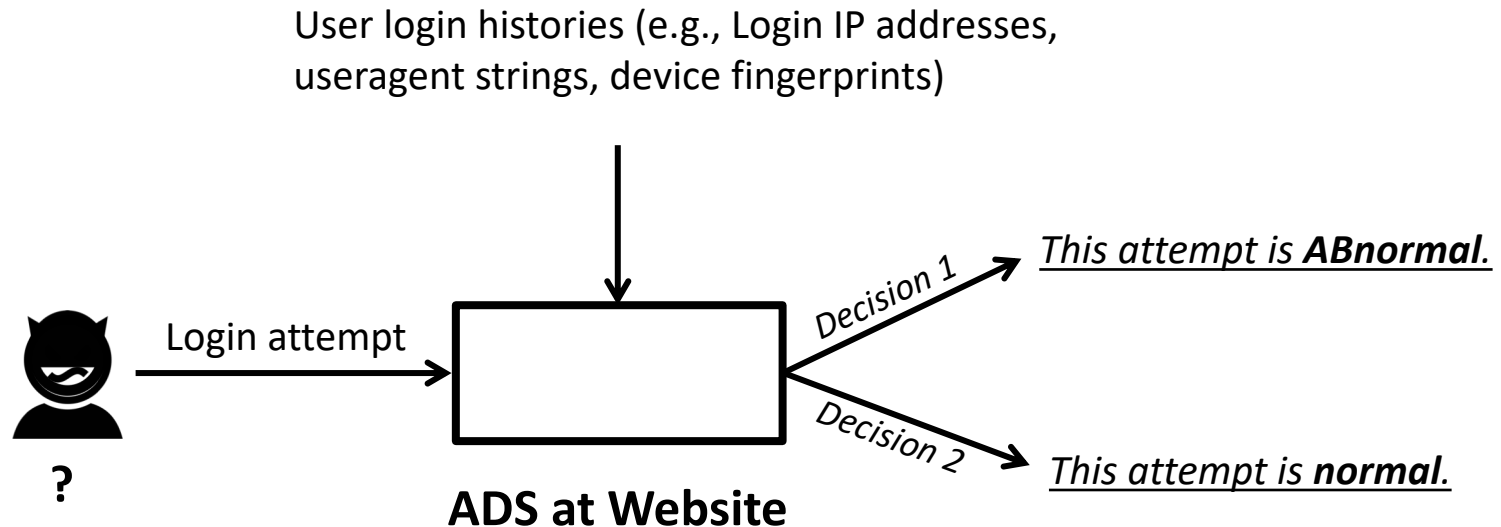


More sophisticated ADS:*

- *Multiple login features*
- *Attackers' different capability levels*



Anomaly Detection Systems (ADS)



“Researching attacker”:*

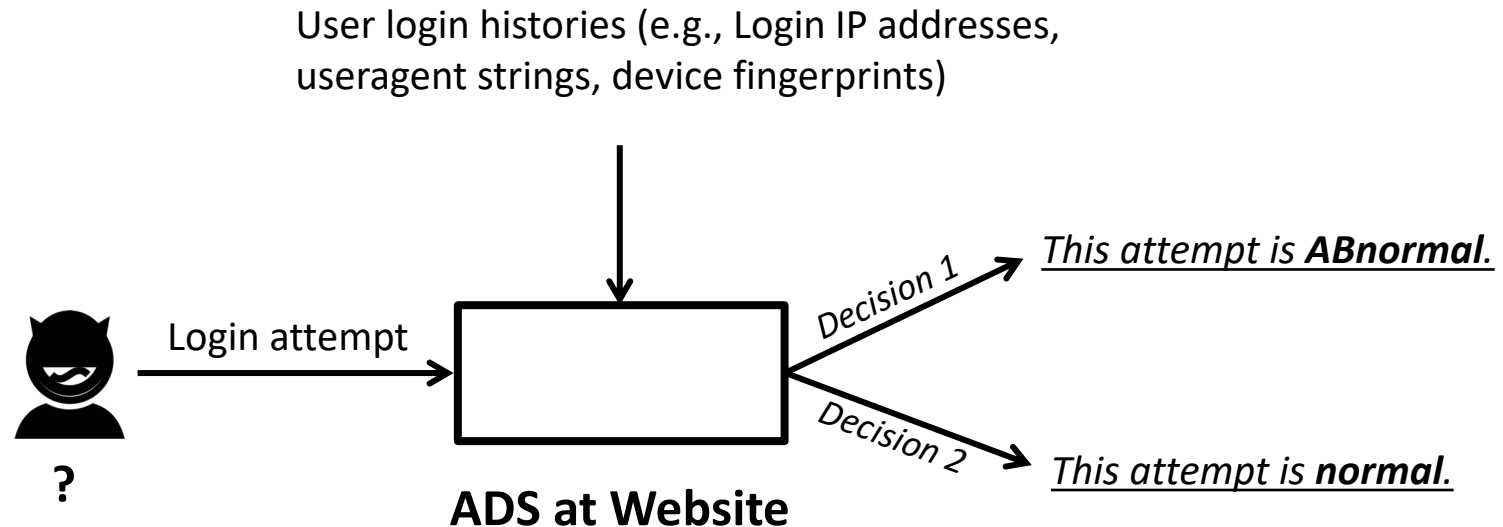
- Hold users’ *correct* passwords
- Try to access users’ accounts from *same countries* of legitimate users

“Phishing attacker”:*

- Hold users’ *correct* passwords
- Try to access users’ accounts from *same countries* with *same browser user-agent strings* of legitimate users



Anomaly Detection Systems (ADS)



ADS:

- leverages users' login patterns (IPs, browser agentstrings, etc.)
- helps a website to distinguish malicious login attempts
- **NOT** an authentication factor that directly decides whether a login attempt is successful or not.



Evidence Trail from Credential Stuffing

c = “alice@yyy.com : ***alicepwd***”,
a leaked username-password pair
possessed by the credential stuffer



**Credential
Stuffer**

Websites where Alice
has accounts



alice@yyy.com : *alicepwd0*
ADS



2FA

alice@yyy.com : *alicepwd*
ADS
2FA



alice@yyy.com : *alicepwd*
ADS



Evidence Trail from Credential Stuffing

$c = \text{"alice@yyy.com : **alicepwd**"}$,
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Websites where Alice
has accounts

$\neq \text{alicepwd}$



Credential
Stuffer

Login attempt with c



alice@yyy.com : alicepwd0
ADS: *abnormal*



2FA

alice@yyy.com : *alicepwd*
ADS
2FA



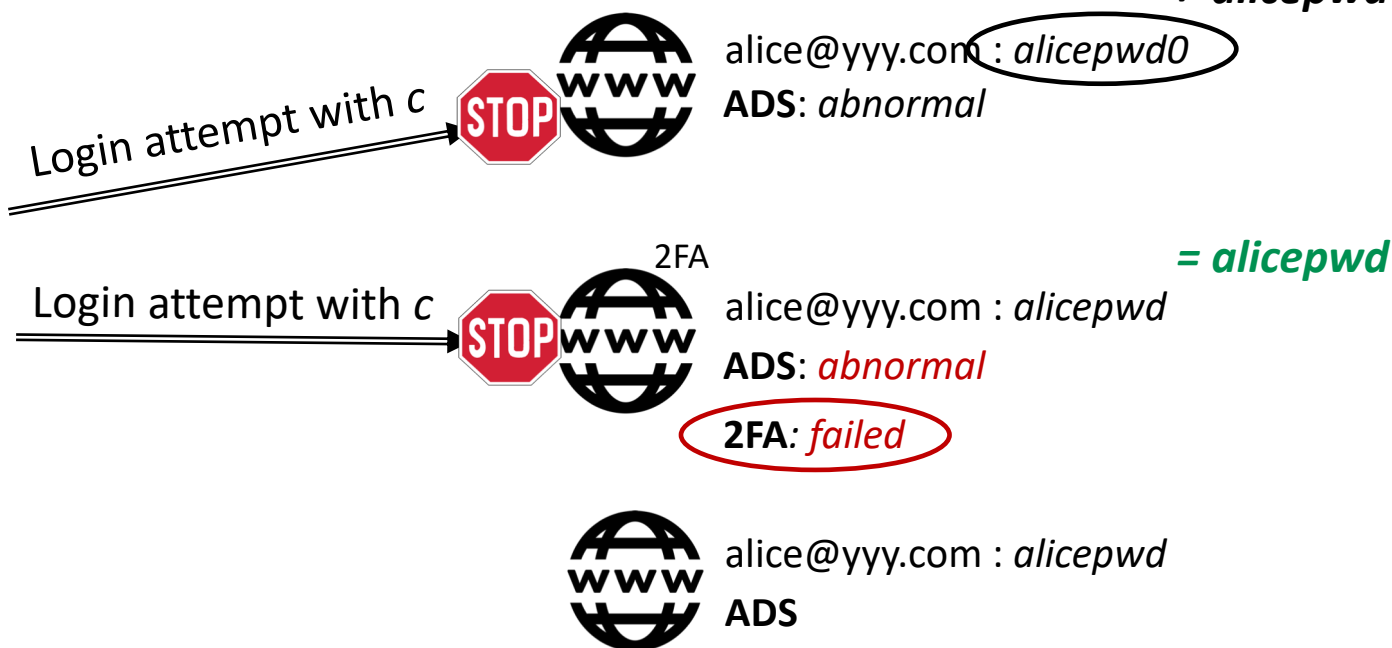
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ADS



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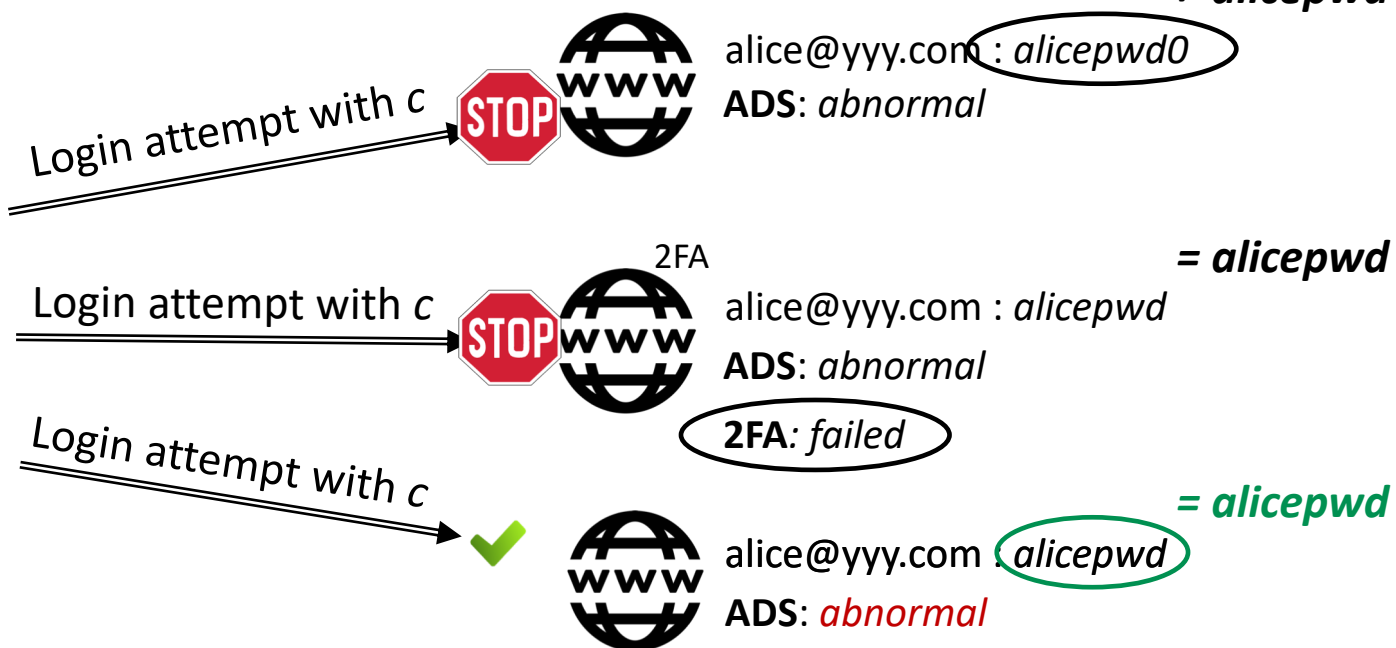
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Evidence Trail from Credential Stuffing

c = “alice@yyy.com : *alicepwd*”,
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Websites where Alice
has accounts

≠ alicepwd

The “trail” left by credential stuffing attacks are those passwords submitted in abnormal login attempts that fail:

- Without 2FA
 - ADS reports “abnormal”; the submitted password is incorrect
- With 2FA:
 - ADS reports “abnormal”; the submitted password is incorrect
 - ADS reports “abnormal”; the submitted password is correct but 2FA fails

pwd

pwd



ADS: *abnormal*



Our Framework

$c = \text{"alice@yyy.com : *alicepwd*"}$,
a leaked username-password pair
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**Credential
Stuffer**

Websites where Alice
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alice@yyy.com : *alicepwd0*
ADS



2FA

alice@yyy.com : *alicepwd*
ADS
2FA



alice@yyy.com : *alicepwd*
ADS



Our Framework

$c = \text{"alice@yyy.com : *alicepwd*"}$,
a leaked username-password pair
possessed by the credential stuffer



Credential
Stuffer

Login attempt with c



Websites where Alice
has accounts

alice@yyy.com : *alicepwd0*

$\neq \text{alicepwd}$

ADS: *abnormal*

SUSPICIOUS: {}



2FA

alice@yyy.com : *alicepwd*

ADS

2FA



alice@yyy.com : *alicepwd*

ADS



Our Framework

$c = \text{"alice@yyy.com : **alicepwd**"}$,
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Credential
Stuffer

Login attempt with c



Websites where Alice
has accounts

alice@yyy.com : **alicepwd0**

ADS: **abnormal**

SUSPICIOUS: { **alicepwd** }

\neq alicepwd

collect



2FA

alice@yyy.com : **alicepwd**

ADS

2FA



alice@yyy.com : **alicepwd**

ADS



Our Framework

c = “alice@yyy.com : **alicepwd**”,
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Credential
Stuffer

Websites where Alice
has accounts



alice@yyy.com : *alicepwd0*
ADS: *abnormal*
SUSPICIOUS: { *alicepwd* }

Login attempt with c



2FA

alice@yyy.com : *alicepwd* = **alicepwd**
ADS: *abnormal*
2FA: *failed*
SUSPICIOUS: { *alicepwd* }

collect



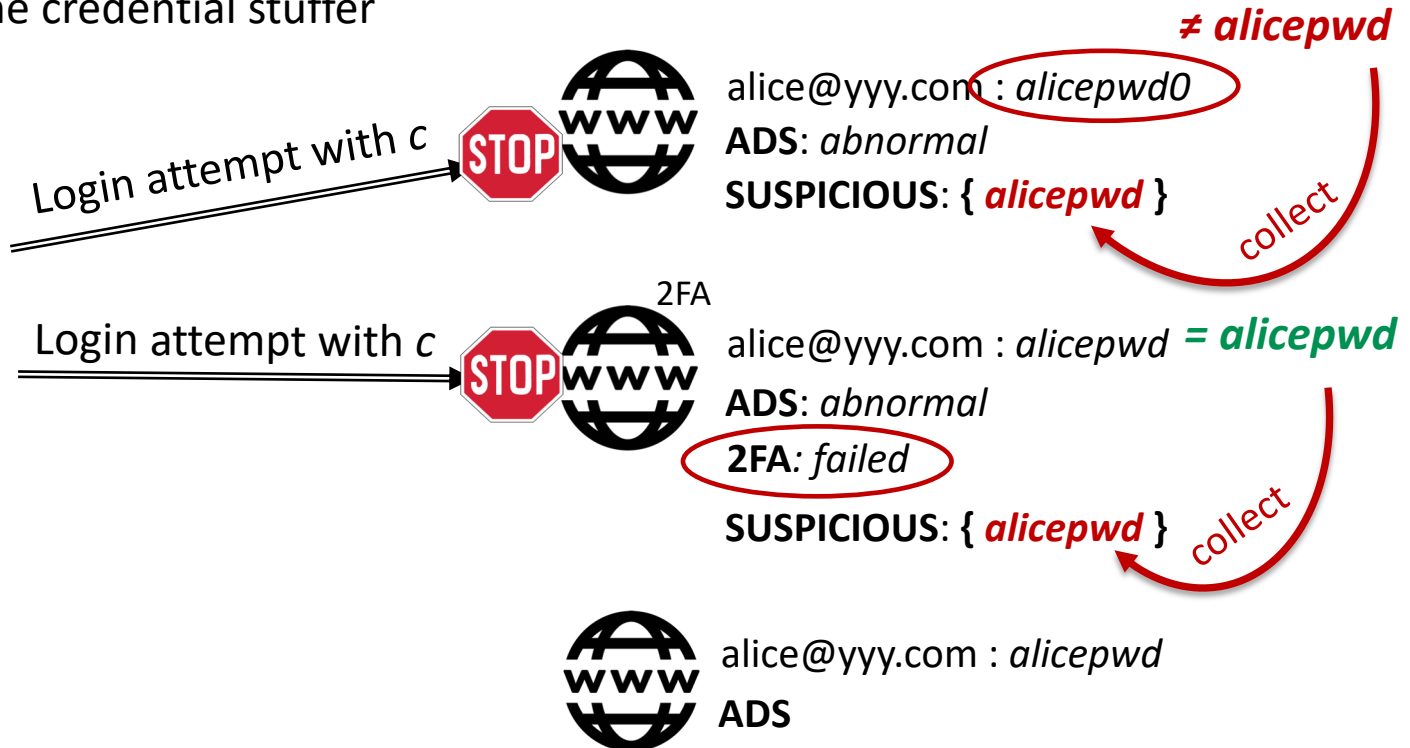
alice@yyy.com : *alicepwd*
ADS



Our Framework

$c = \text{"alice@yyy.com : **alicepwd**"}$,
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Websites where Alice
has accounts



COLLECTING phase



Our Framework

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**Credential
Stuffer**

Login attempt with c

Websites where Alice
has accounts



alice@yyy.com : *alicepwd0*
ADS: *abnormal*
SUSPICIOUS: { **alicepwd** }



2FA

alice@yyy.com : *alicepwd*
ADS: *abnormal*
2FA: *failed*
SUSPICIOUS: { **alicepwd** }



alice@yyy.com : **alicepwd** = **alicepwd**
ADS: **abnormal**



Our Framework

$c = \text{"alice@yyy.com : **alicepwd**"}$,
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*Have you collected
"alicepwd" for
"alice@yyy.com"?*

Login attempt with c

Websites where Alice
has accounts



alice@yyy.com : *alicepwd0*
ADS: *abnormal*
SUSPICIOUS: { **alicepwd** }



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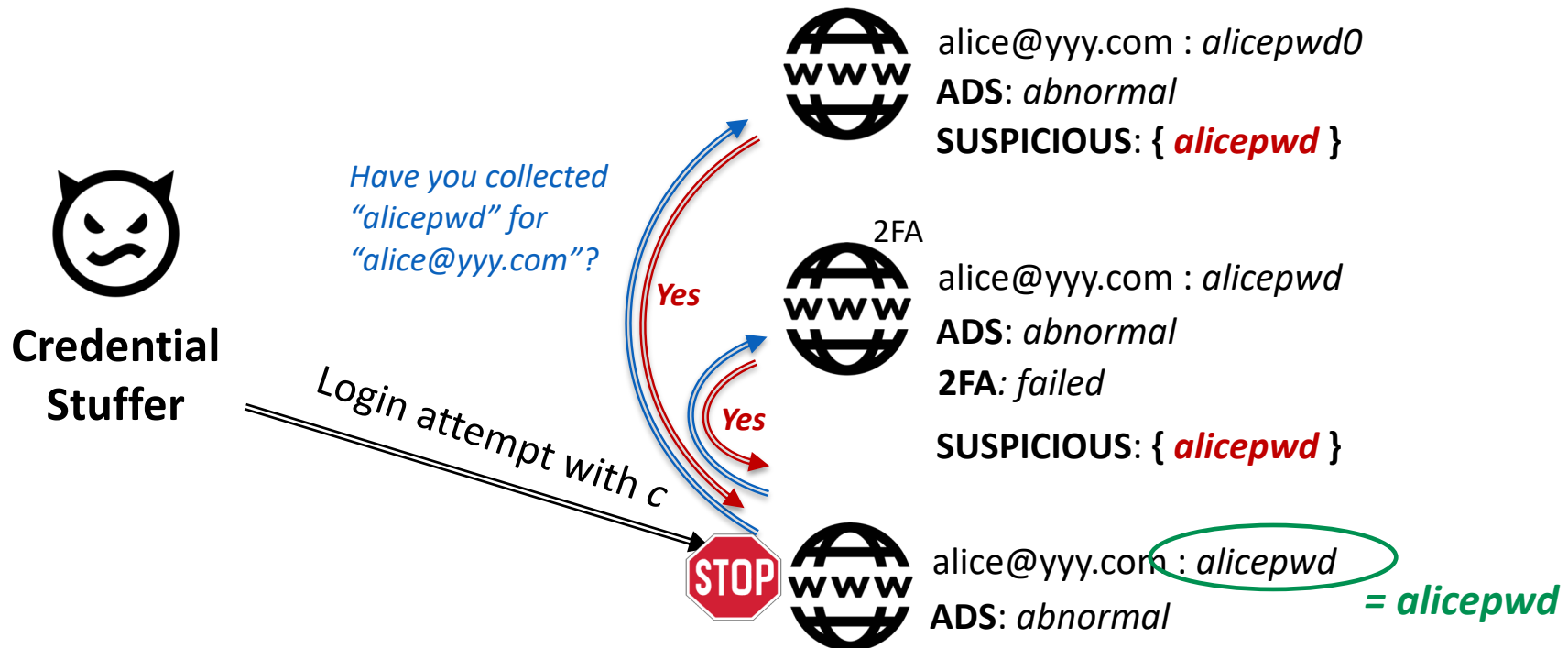
alice@yyy.com : **alicepwd** = **alicepwd**
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Our Framework

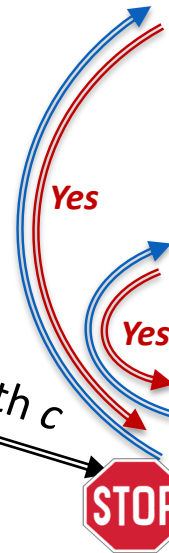
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SUSPICIOUS: { **alicepwd** }



alice@yyy.com : *alicepwd*
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alice@yyy.com : **alicepwd**
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= alicepwd

COUNTING phase



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Our Framework

$c = \text{"alice@yyy.com : alicepwd"}$,
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Websites where Alice

Two important questions:

- **False detection rate (FDR)**
 - *What if a (forgetful) user "guesses" her own passwords at her accounts?*
- **True detection rate (TDR)**
 - *What if a credential stuffer tries to circumvent detection by trying a smart attack strategy?*



alice@yyy.com : alicepwd

ADS: abnormal

= *alicepwd*



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COUNTING phase

Conservatively Estimating FDR & TDR

- ***A forgetful user as a MDP****:
 - ***Maximizing the probability of triggering a false detection (false detection rate)***



Conservatively Estimating FDR & TDR

- A forgetful user as a MDP*:
 - Maximizing the probability of triggering a false detection (false detection rate)
- ***A credential stuffer as a MDP*:***
 - ***Minimizing the probability of getting detected while maximizing the number of account takeovers (true detection rate)***



Conservatively Estimating FDR & TDR

Phishing attackers*: valid passwords from same countries with same browser user-agent strings of legitimate users



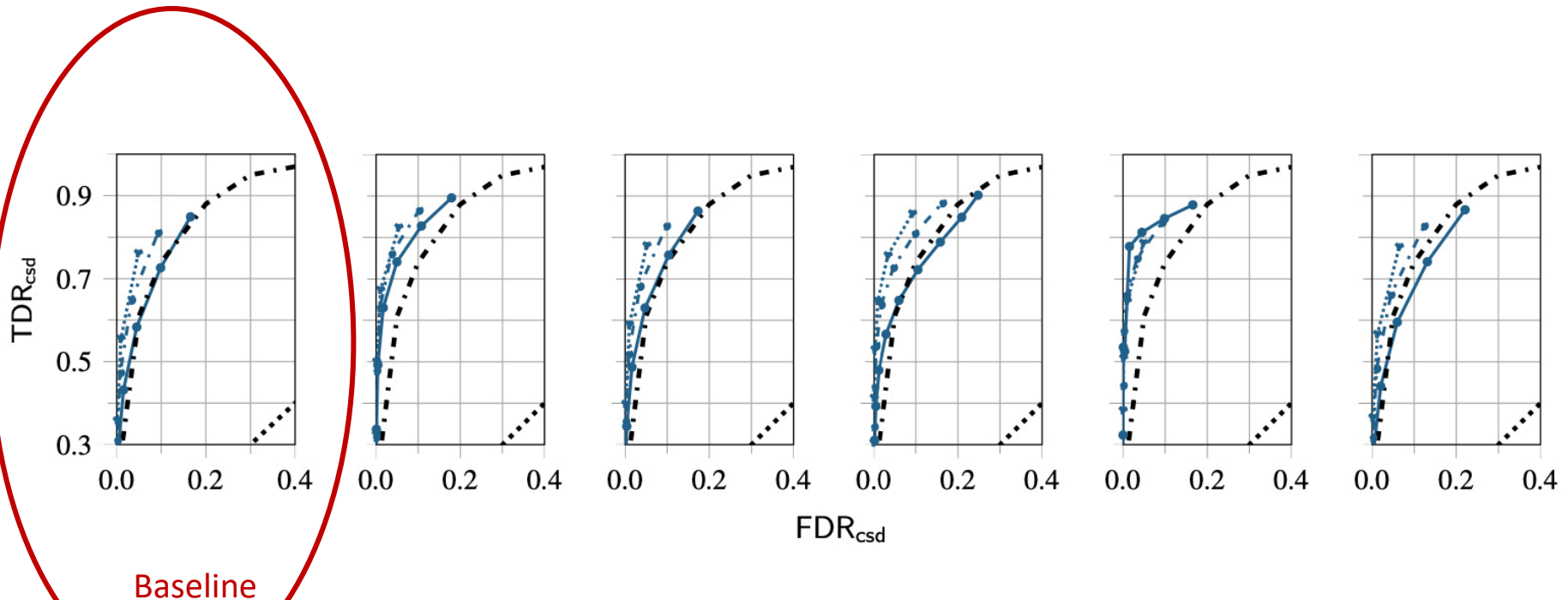
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* Freeman et al. (NDSS 2016)

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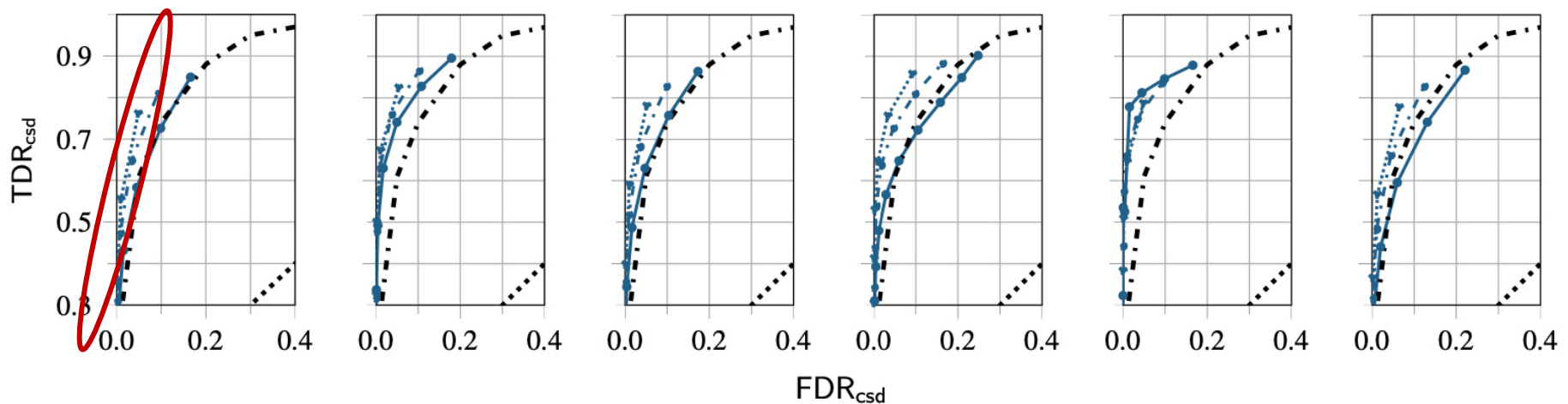
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- **Blue curves**: each for a different *ADS threshold* in the collecting phase



Baseline



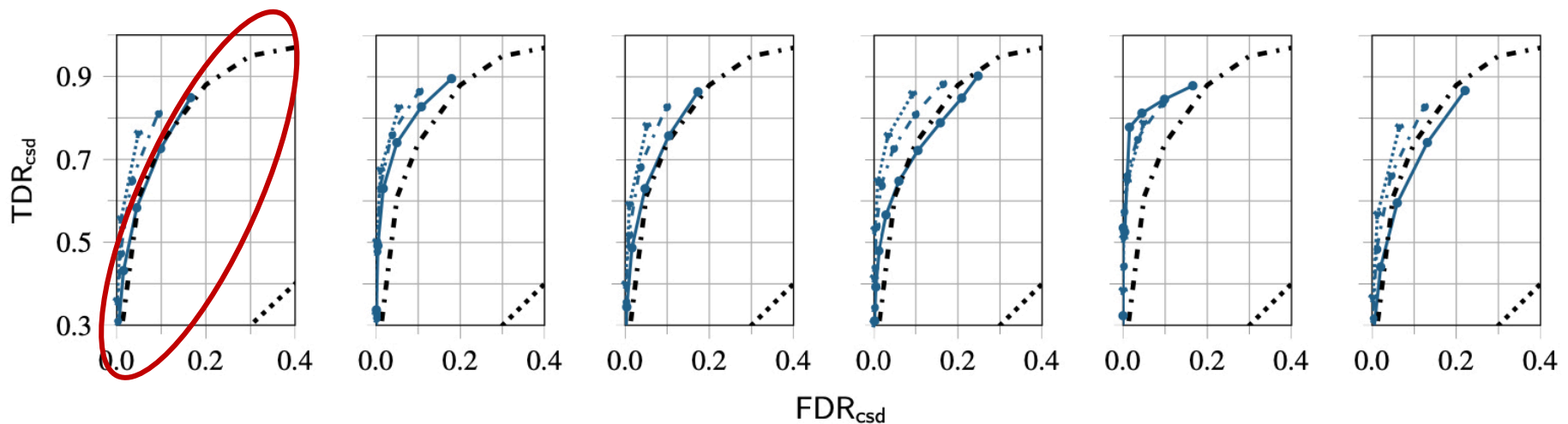
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- **Black, dashed curves**: corresponding ADS's accuracy in detecting abnormal logins



Baseline



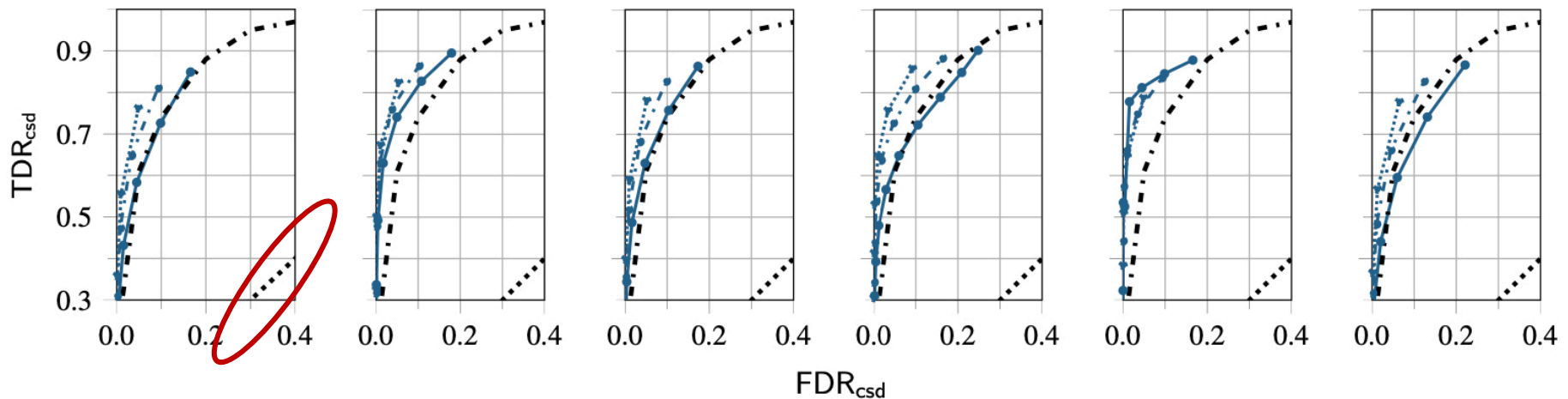
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- **Black, dotted lines**: random guessing



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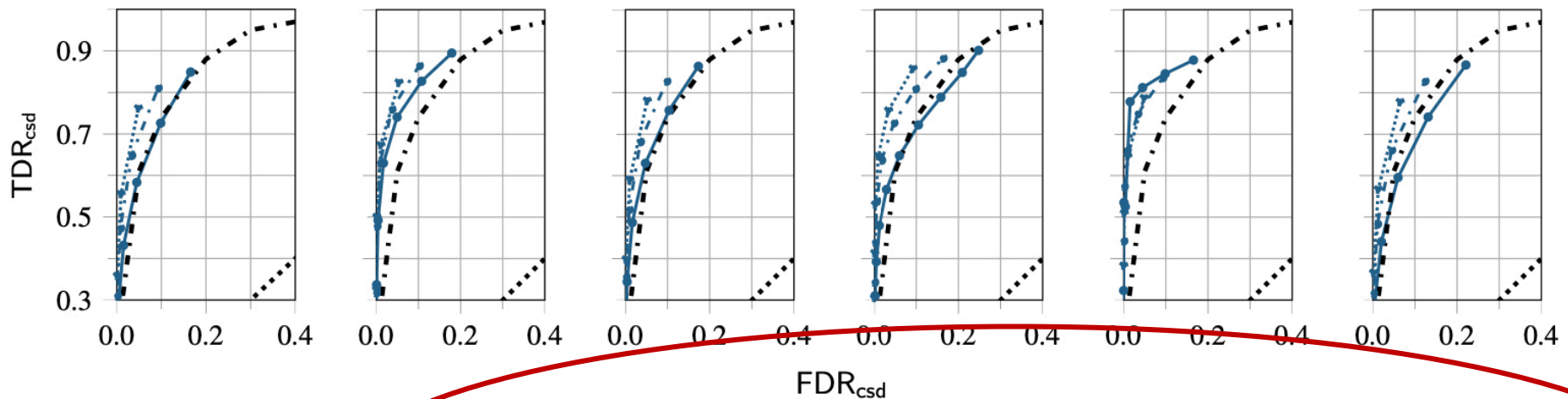
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Baseline

Less pwd
reuse

of pwds
+1

of accnts
+ 10

of 2FA
+ 5

Higher ADS
detection rates
in the *counting*
phase



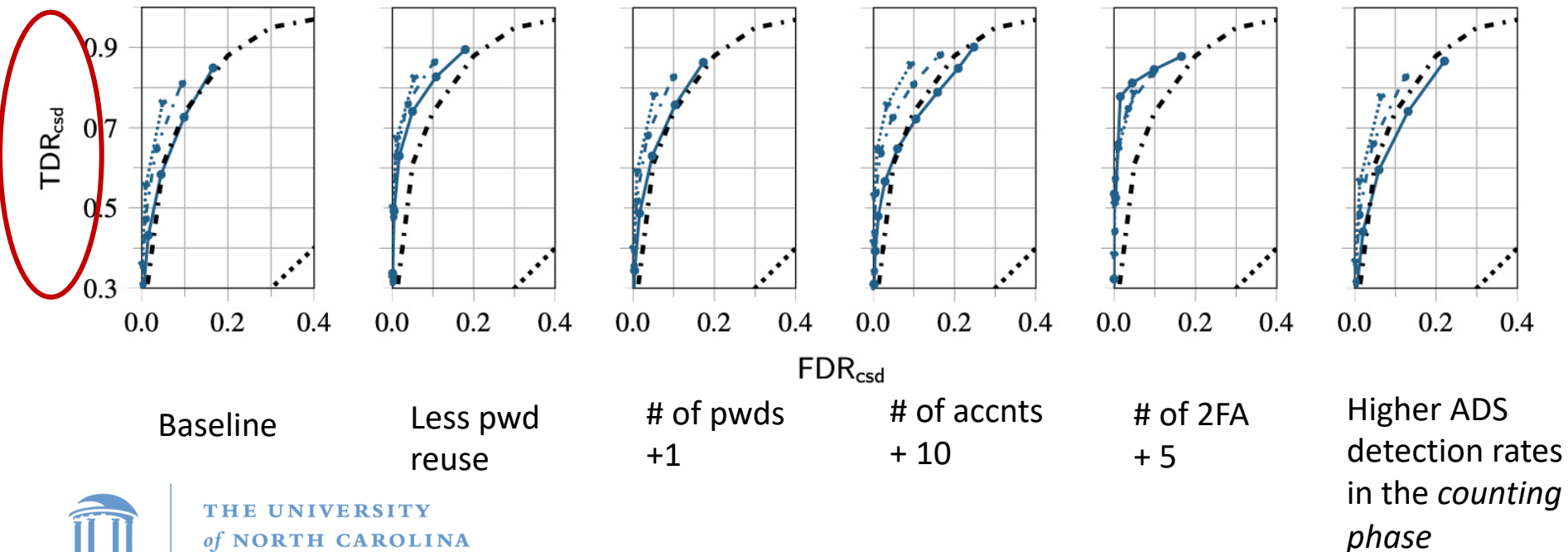
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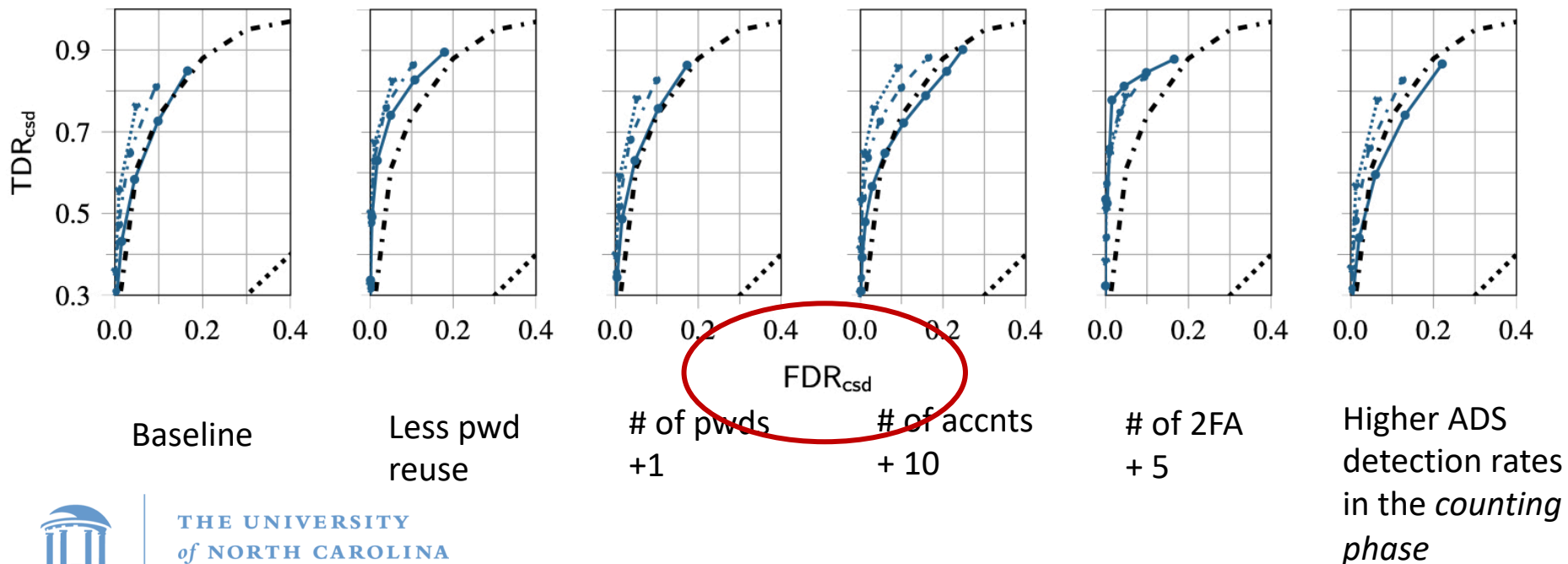
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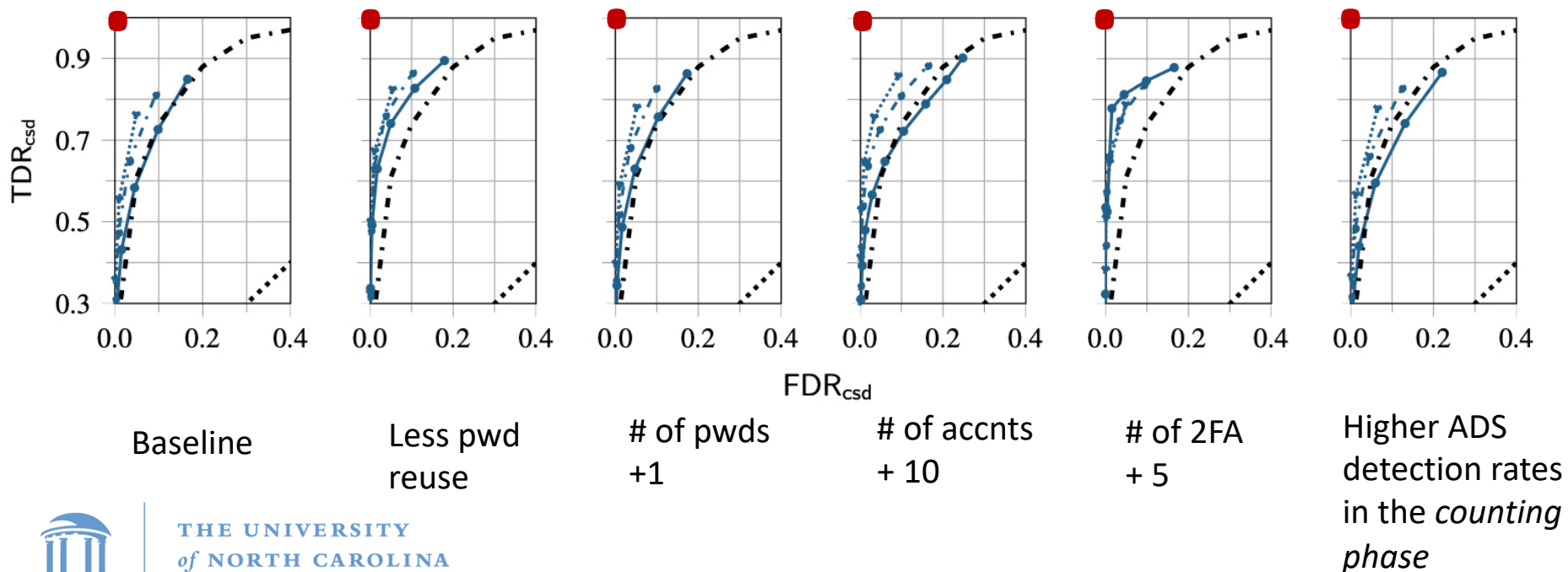
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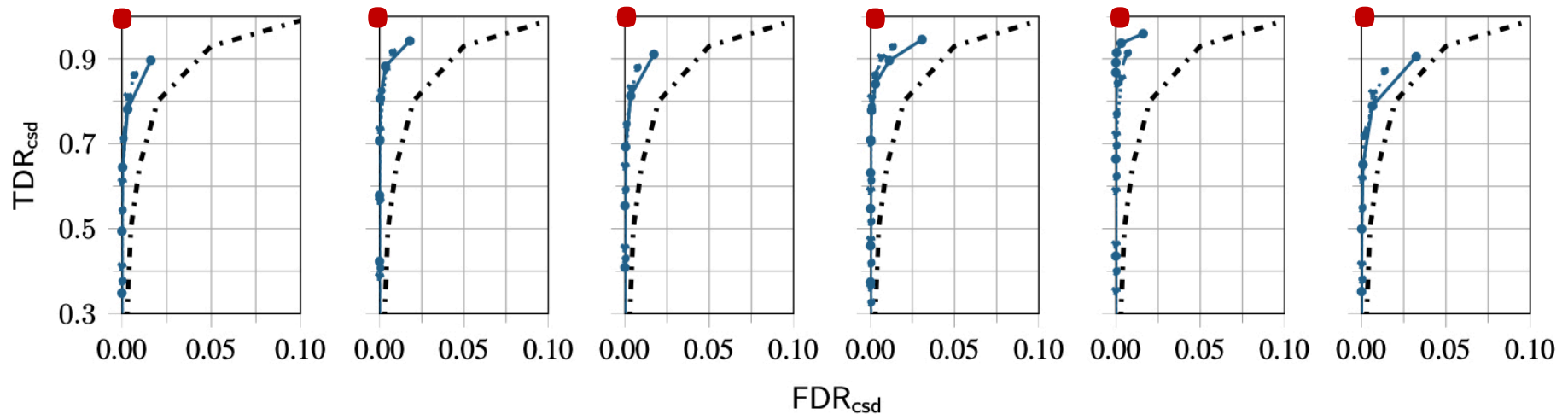


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Conservatively Estimating FDR & TDR

Researching attackers*: valid passwords from same countries of legitimate users.



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reuse

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Higher ADS
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Other features of our framework:



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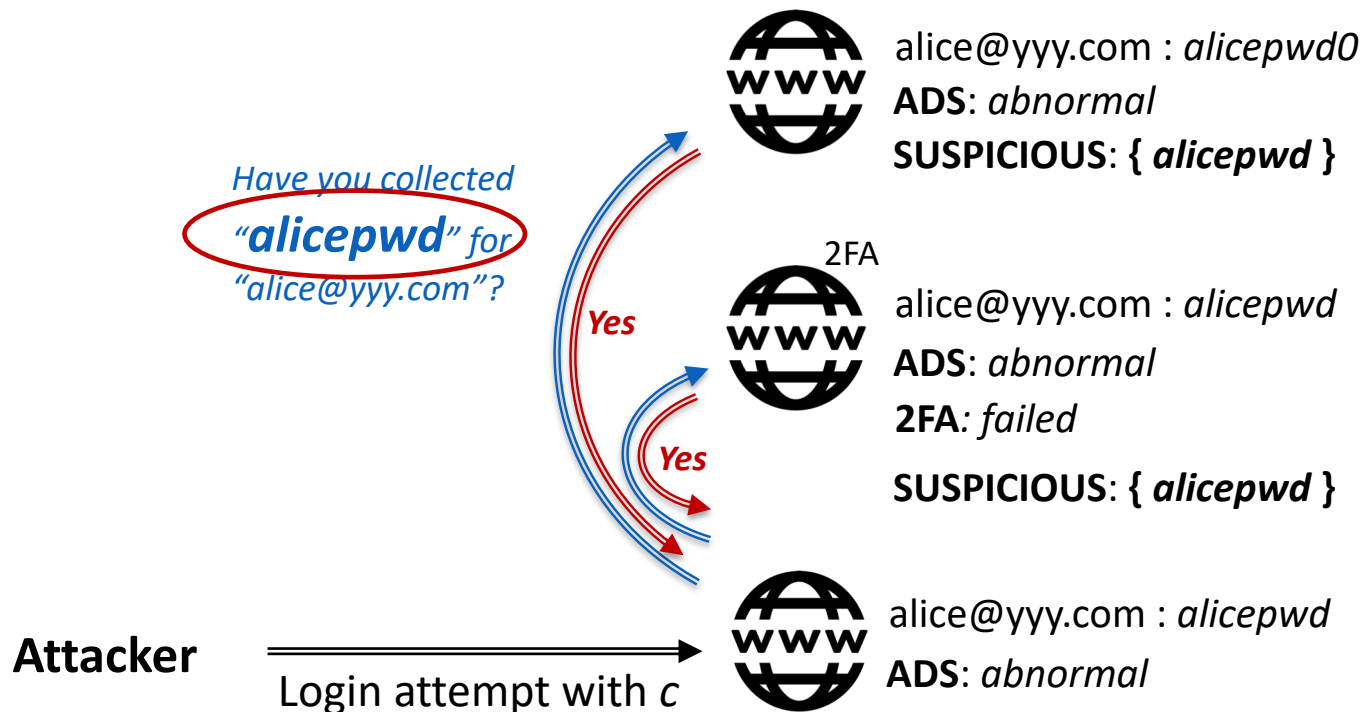
- ***Account security***



Account Security

COUNTING phase

Websites where Alice
has accounts



$c = \text{“alice@yyy.com : alicepwd”}$



Account Security

COUNTING phase

Websites where Alice has accounts

Have you collected
“*alicepwd*” for
“*alice@yyy.com*”?

Private membership
test (PMT) query

Attacker

Login attempt with c



alice@yyy.com : *alicepwd0*
ADS: *abnormal*
SUSPICIOUS: { *alicepwd* }



2FA

alice@yyy.com : *alicepwd*
ADS: *abnormal*
2FA: *failed*
SUSPICIOUS: { *alicepwd* }



alice@yyy.com : *alicepwd*
ADS: *abnormal*

PMT
response

PMT
response



Other features of our framework:

- Account security
 - ***A new one-round two-party private membership test (PMT) protocol***



Other features of our framework:

- Account security
 - A new one-round two-party private membership test (PMT) protocol
- ***Directory***

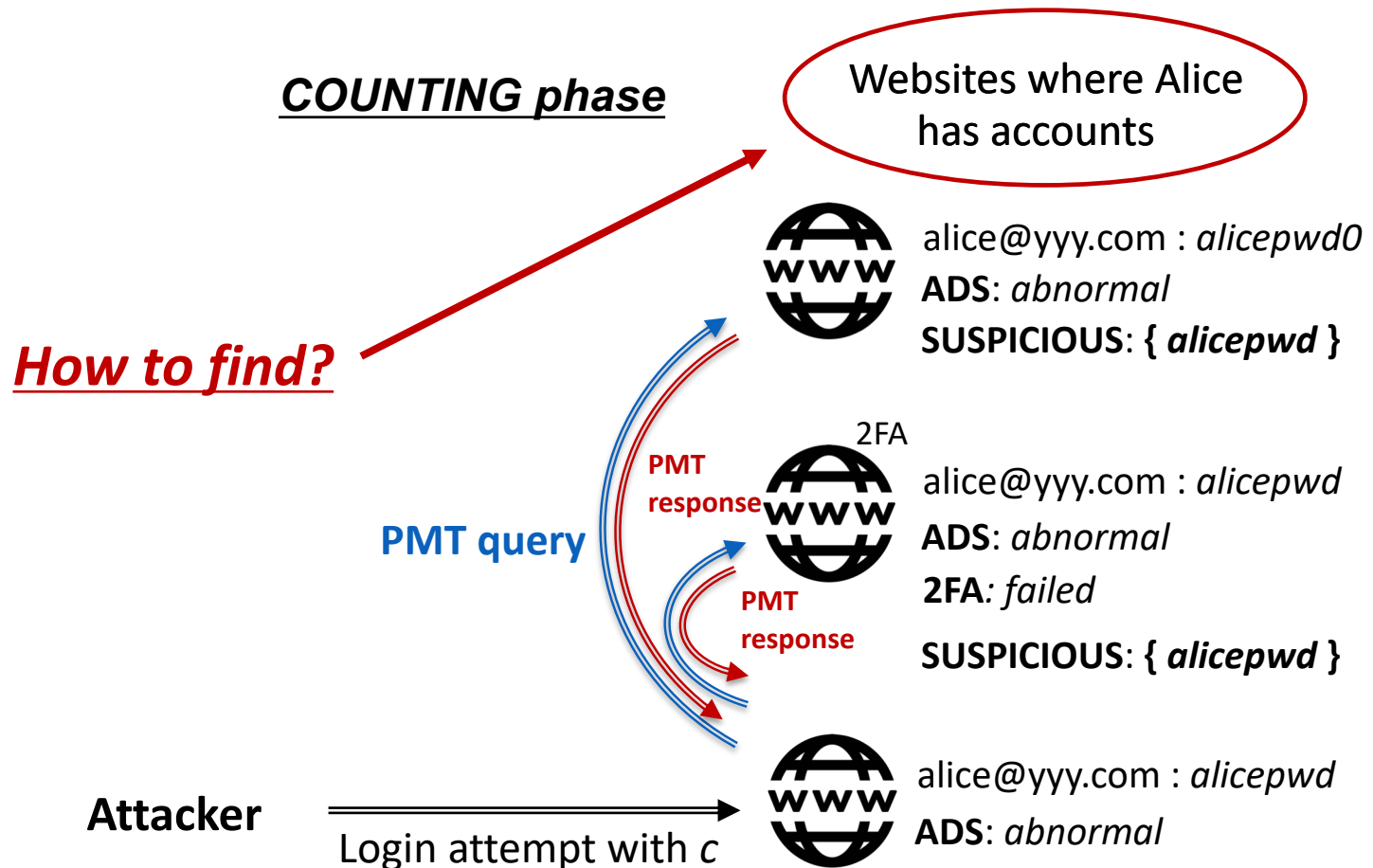


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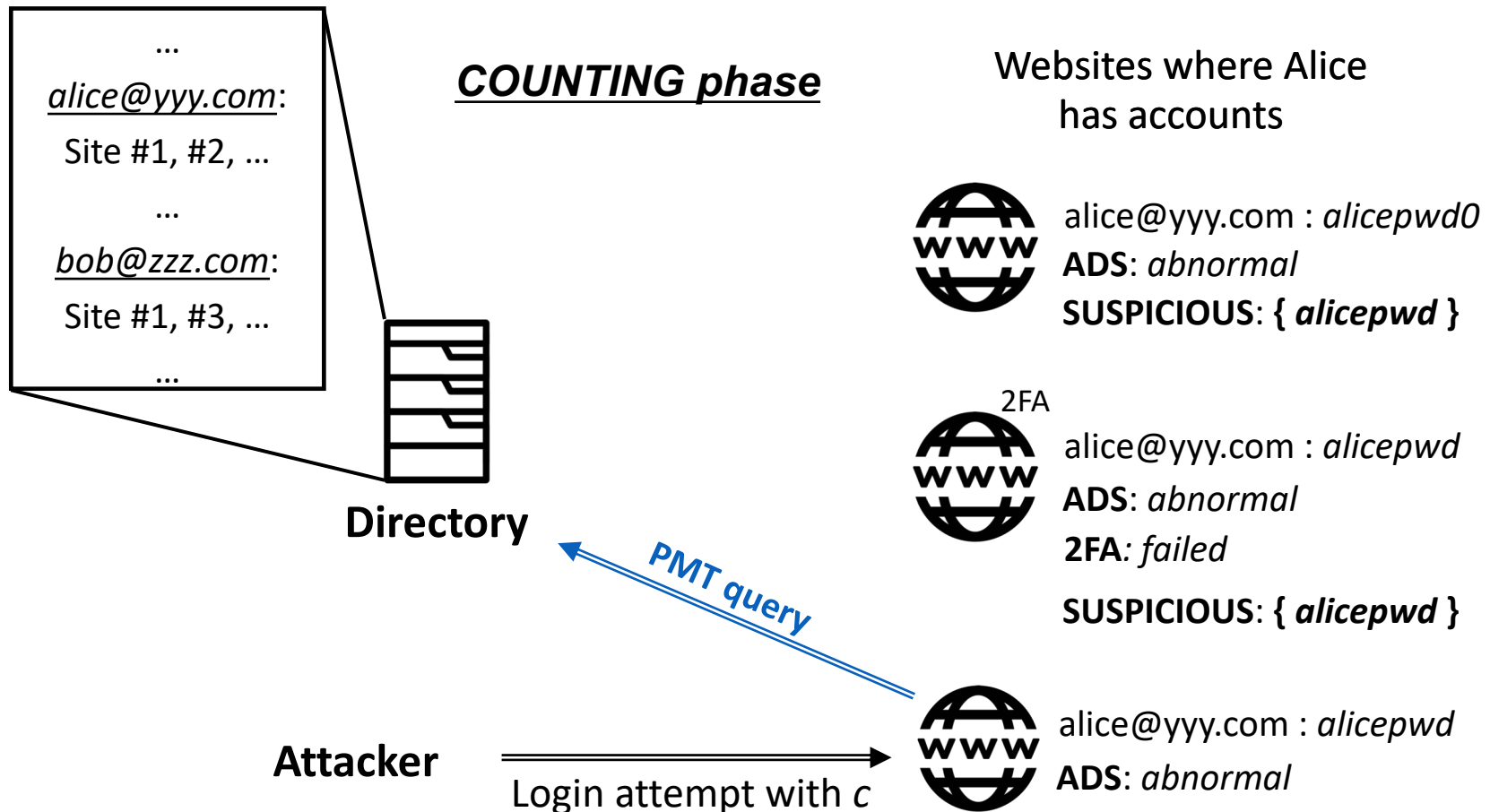
- Account security
 - A new one-round two-party private membership test (PMT) protocol
- Directory
 - *A “look-up table” that maintains where a user has accounts*



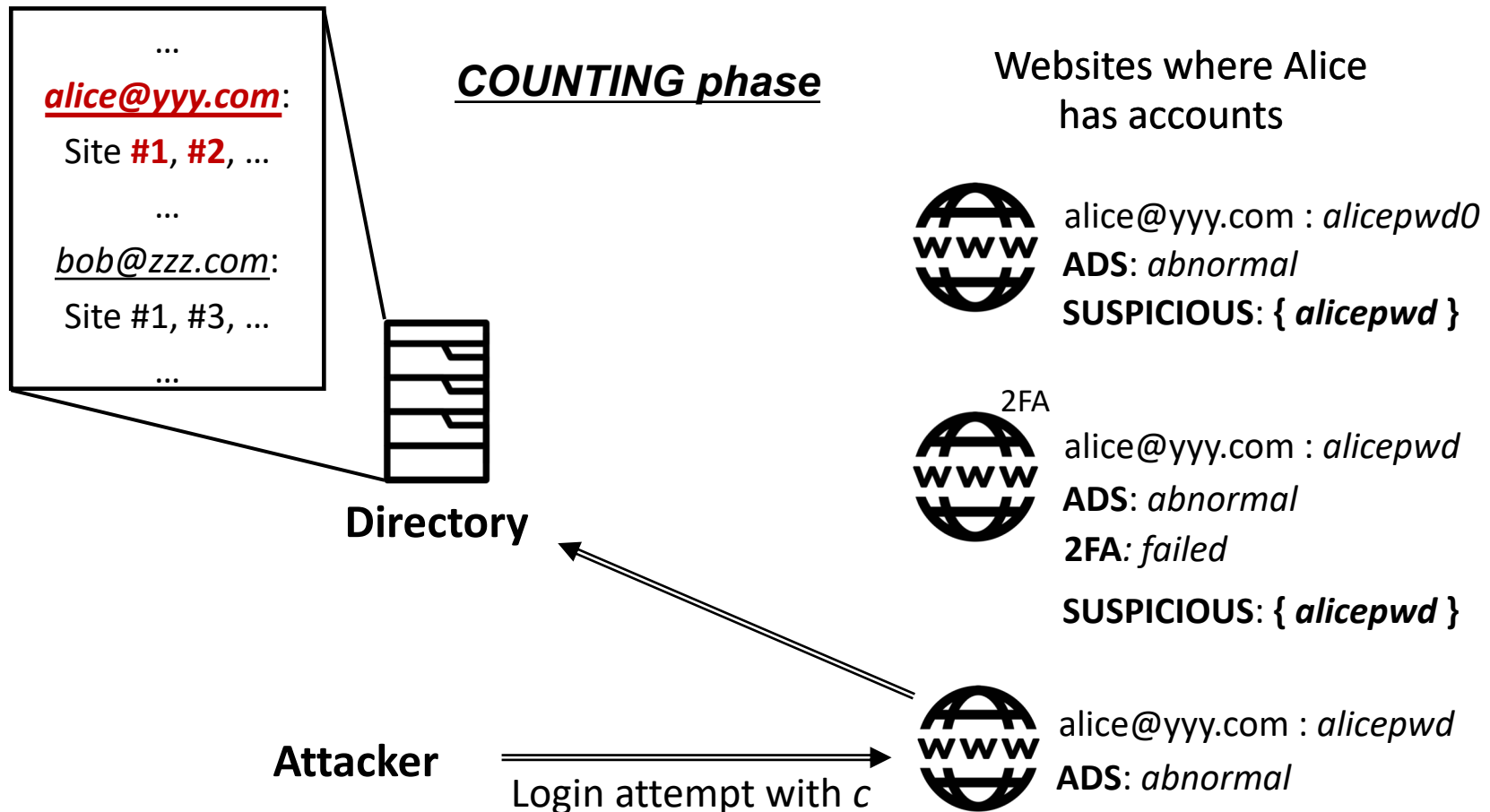
Directory



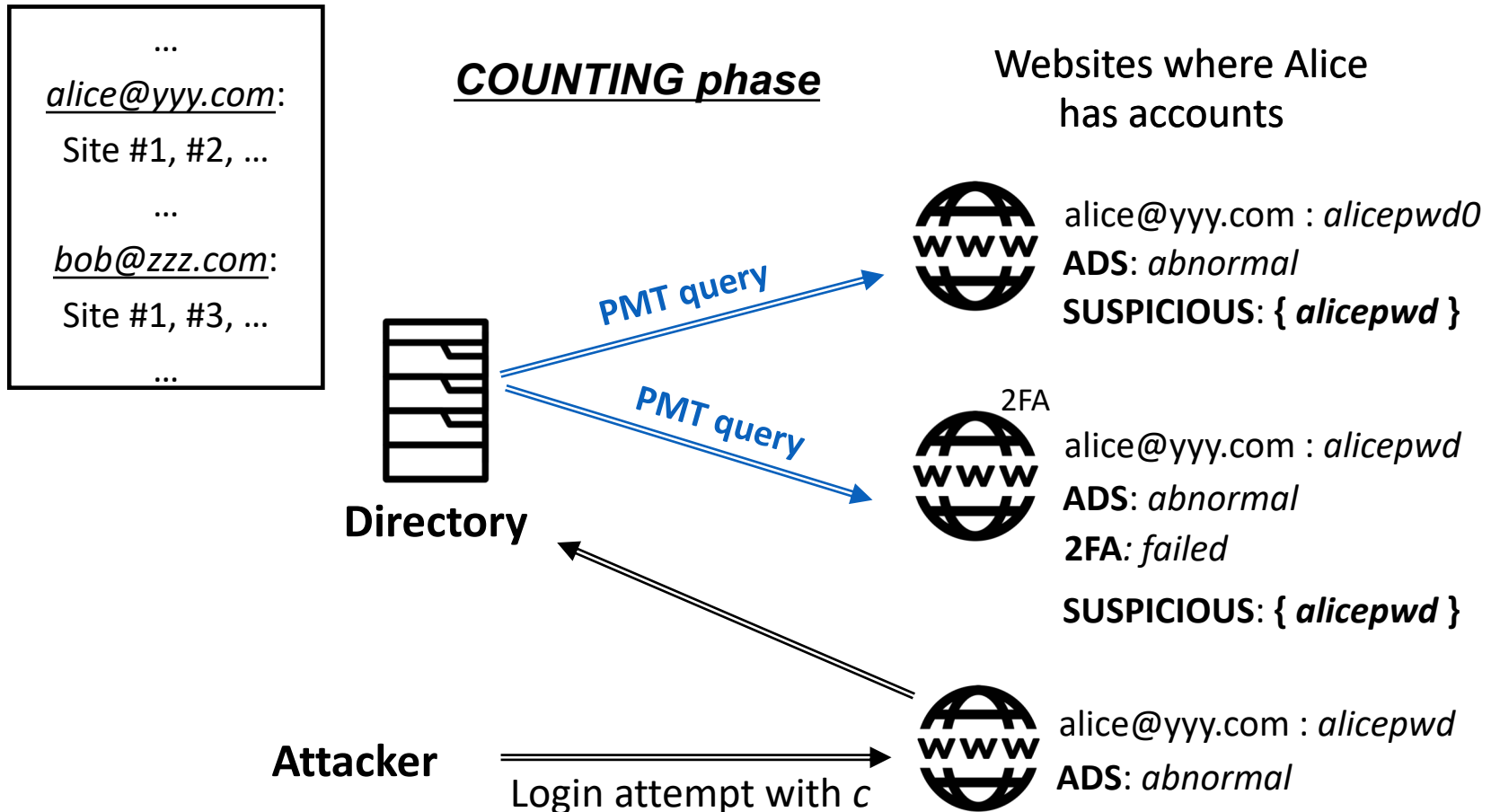
Directory



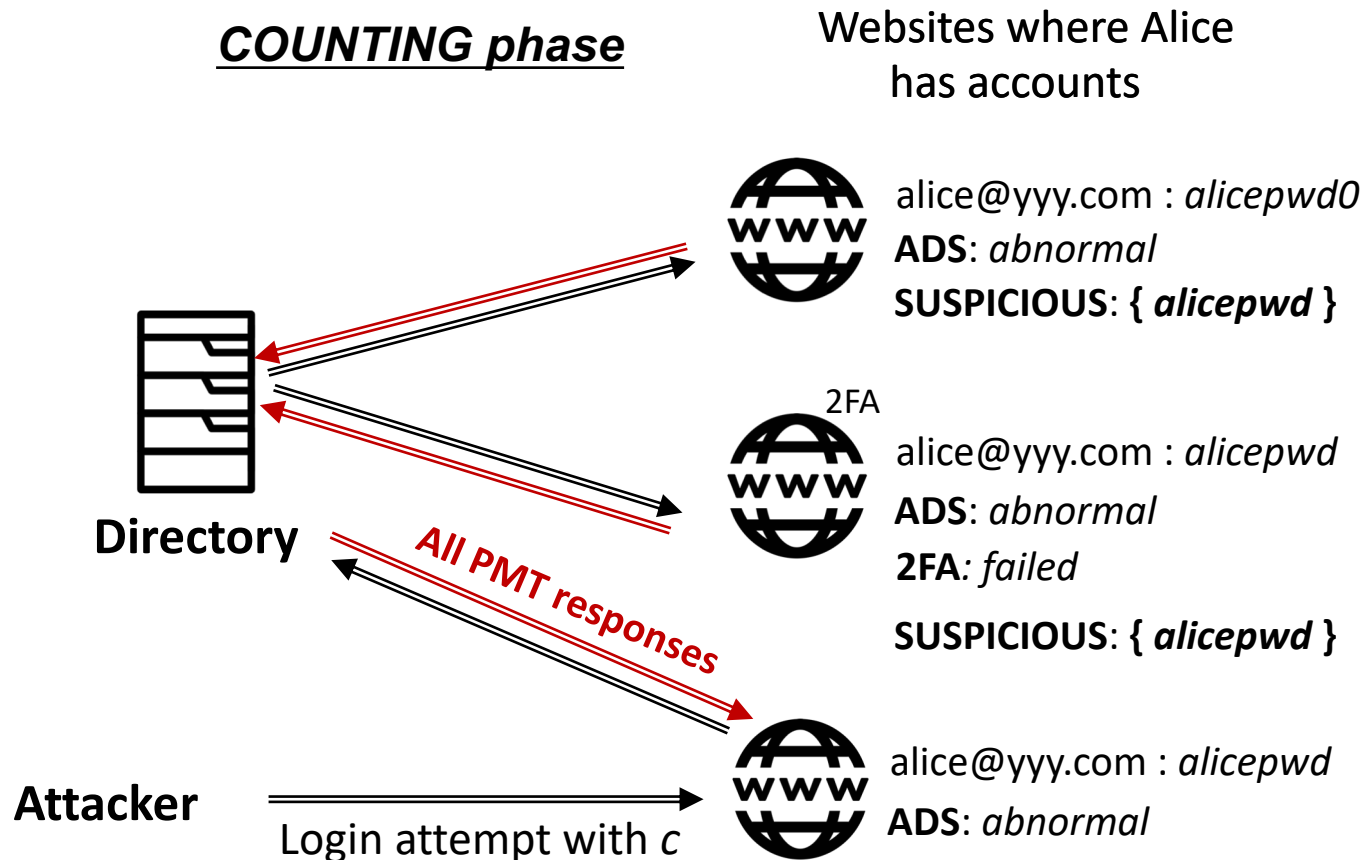
Directory



Directory



Directory



Other features of our framework:

- Account security
 - A new one-round two-party private membership test (PMT) protocol
- Directory
 - A “look-up table” that maintains where a user has accounts
- ***Login privacy***

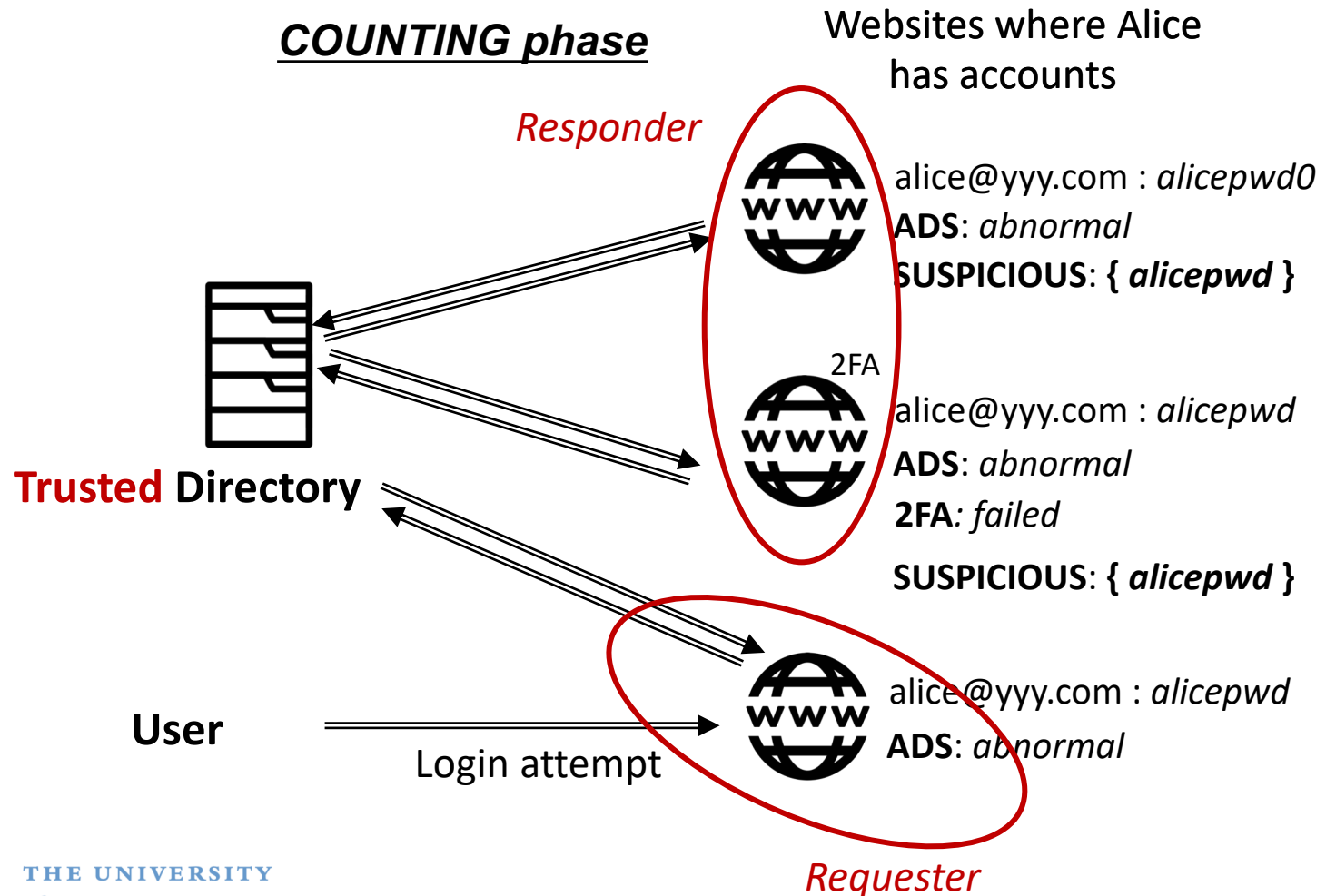


Other features of our framework:

- Account security
 - A new one-round two-party private membership test (PMT) protocol
- Directory
 - A “look-up table” that maintains where a user has accounts
- Login privacy
 - *Trusted directory for login privacy*



Login Privacy

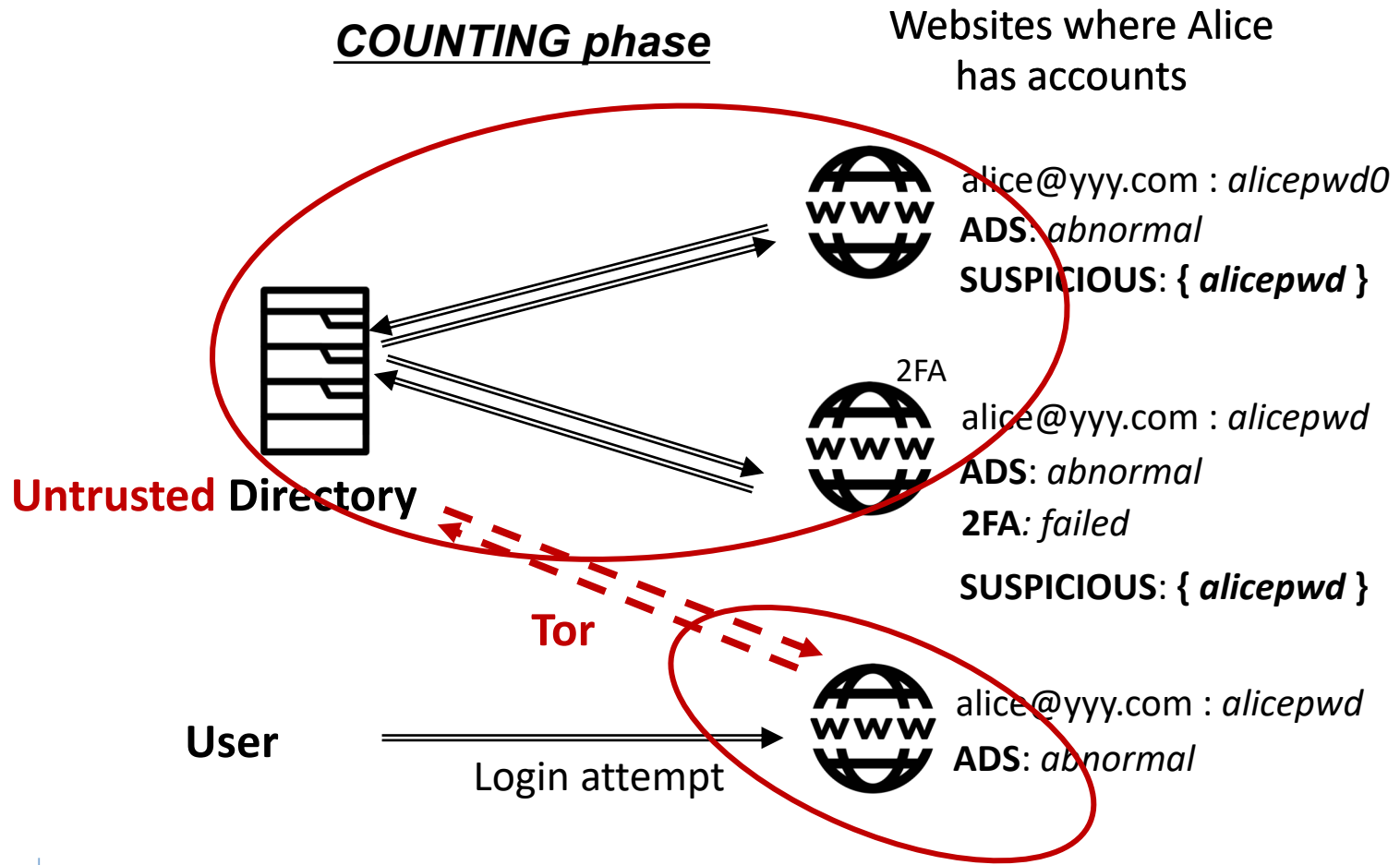


Other features of our framework:

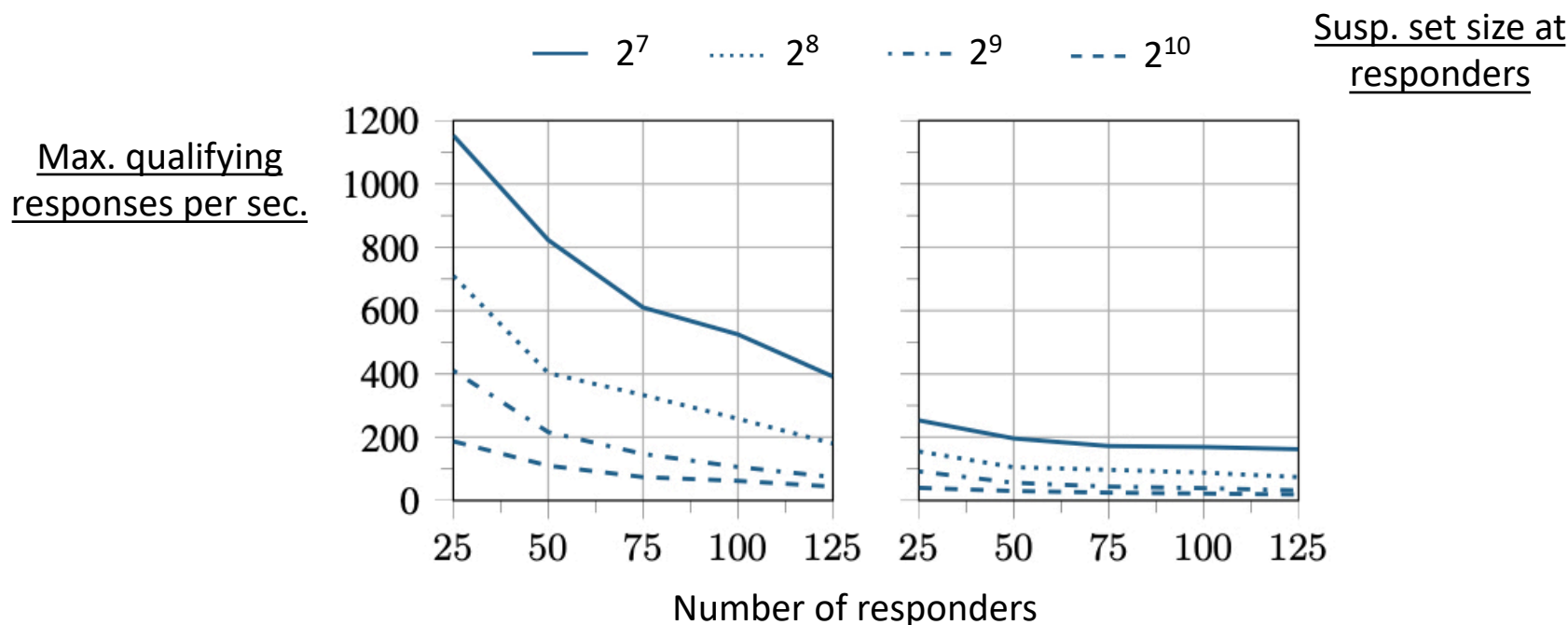
- Account security
 - A new one-round two-party private membership test (PMT) protocol
- Directory
 - A “look-up table” that maintains where a user has accounts
- Login privacy
 - Trusted directory for login privacy
 - ***Untrusted directory for login privacy***



Login Privacy



Scalability

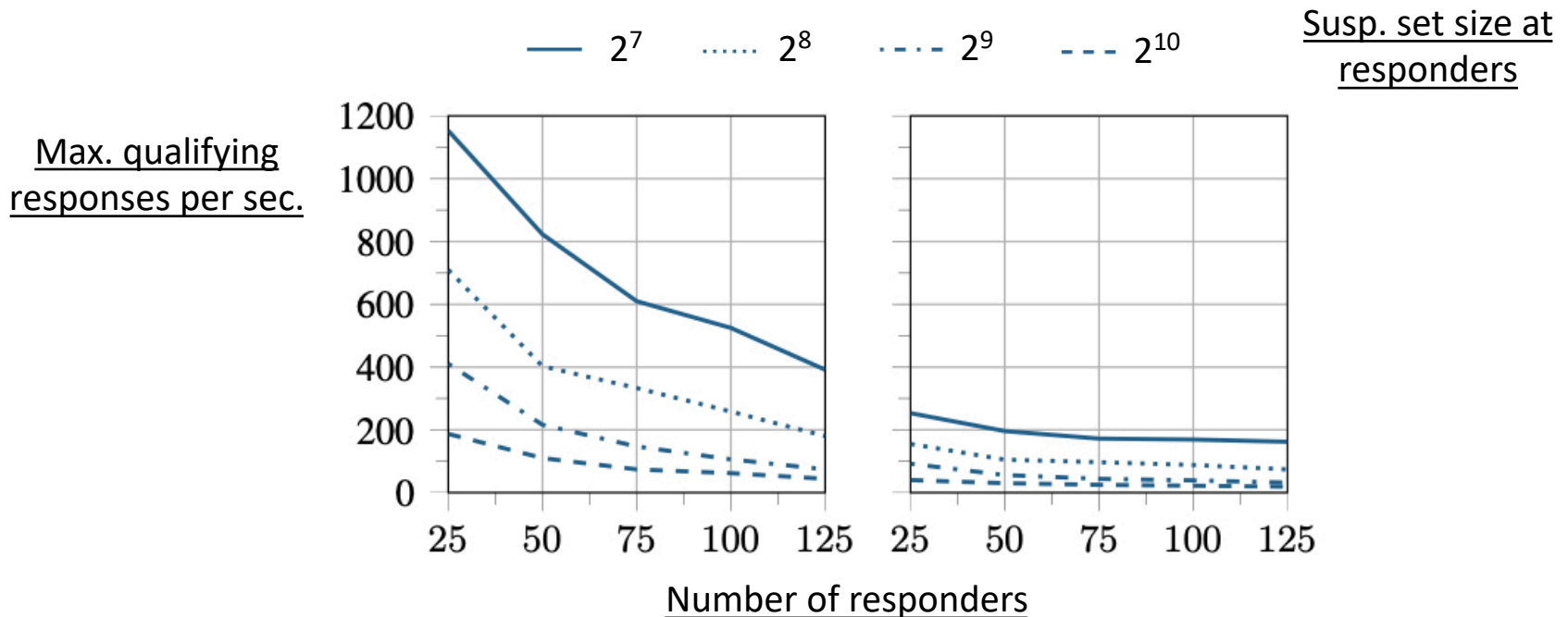


Trusted directory
for *login privacy*
(Qualifying response: $\leq 5s$)

Untrusted directory
for *login privacy*
(Qualifying response: $\leq 8s$)



Scalability



Trusted directory
for login privacy

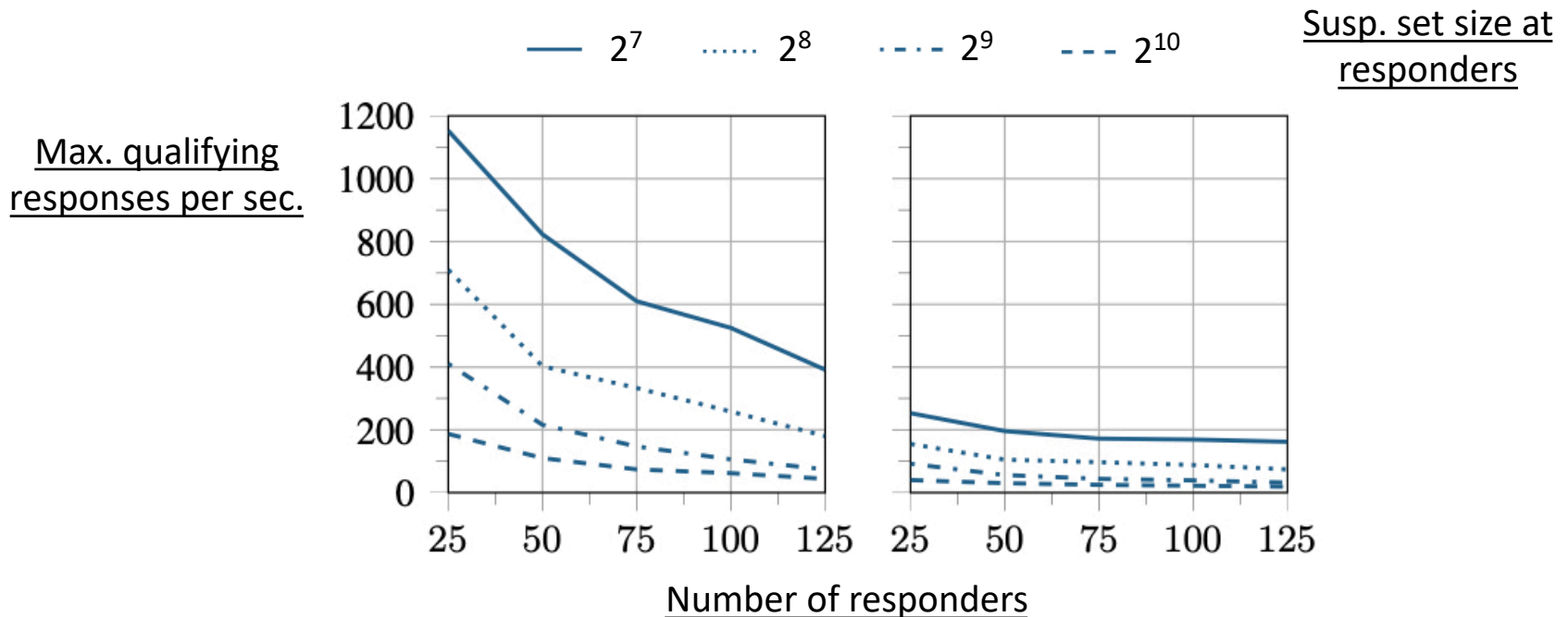
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Untrusted directory
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Scalability

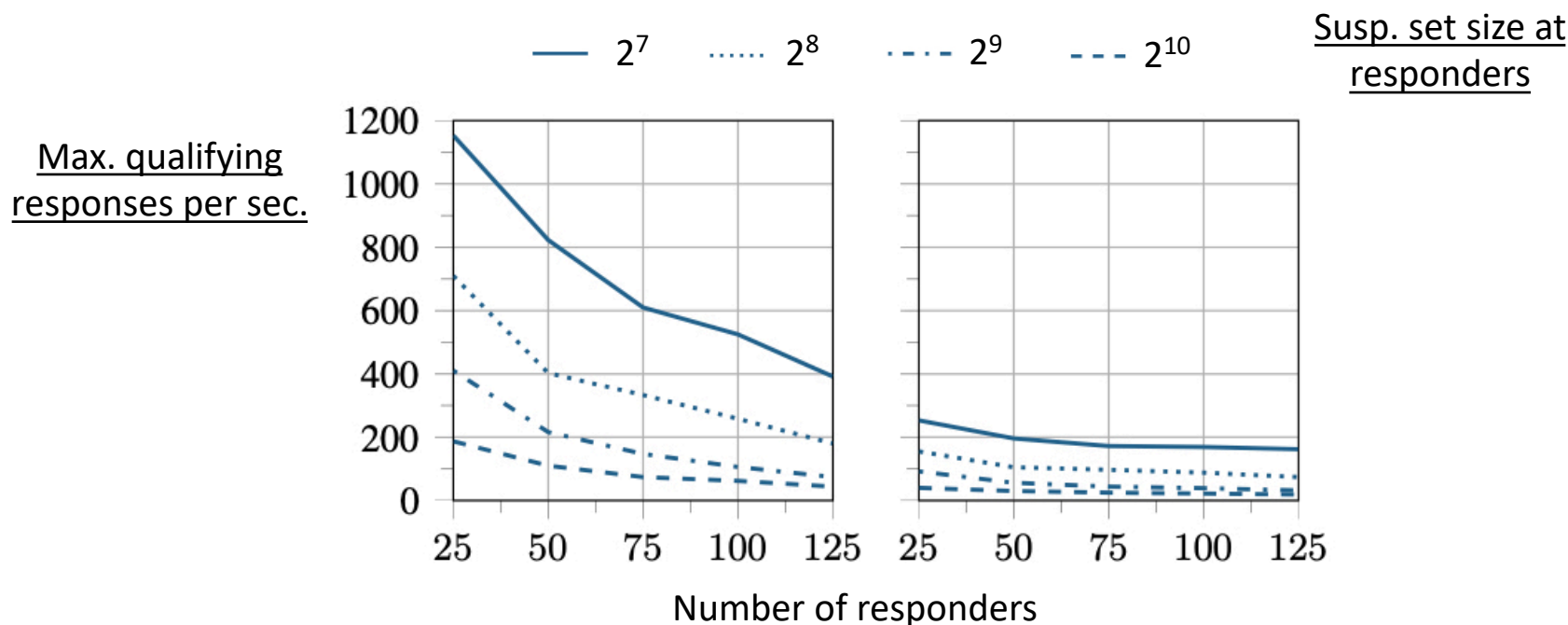


Trusted directory
for *login privacy*
(Qualifying response: $\leq 5s$)

Untrusted directory
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(Qualifying response: $\leq 8s$)



Scalability



Trusted directory
for login privacy

(Qualifying response: $\leq 5s$)

Untrusted directory
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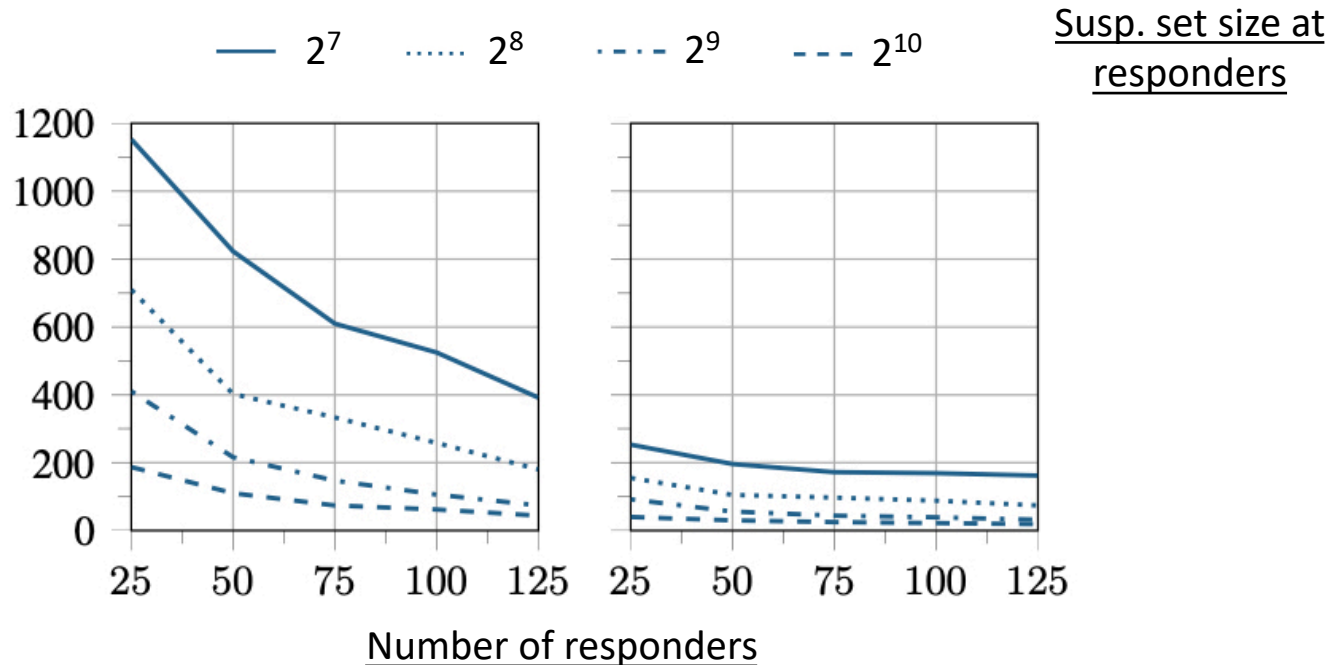
(Qualifying response: $\leq 8s$)

Response time measured at the requester



Scalability

Max. qualifying
responses per sec.

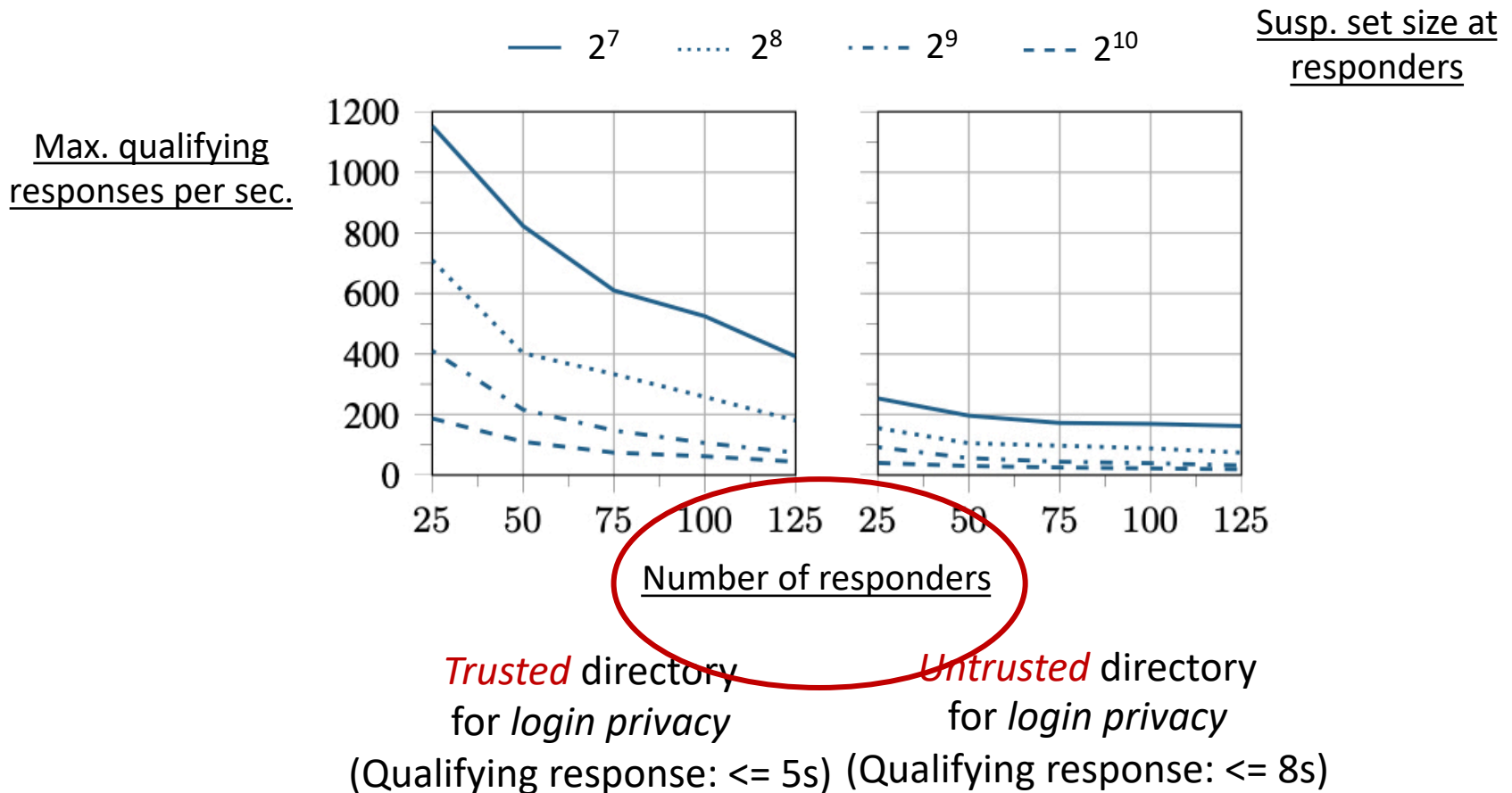


Trusted directory
for *login privacy*
(Qualifying response: $\leq 5s$)

Untrusted directory
for *login privacy*
(Qualifying response: $\leq 8s$)

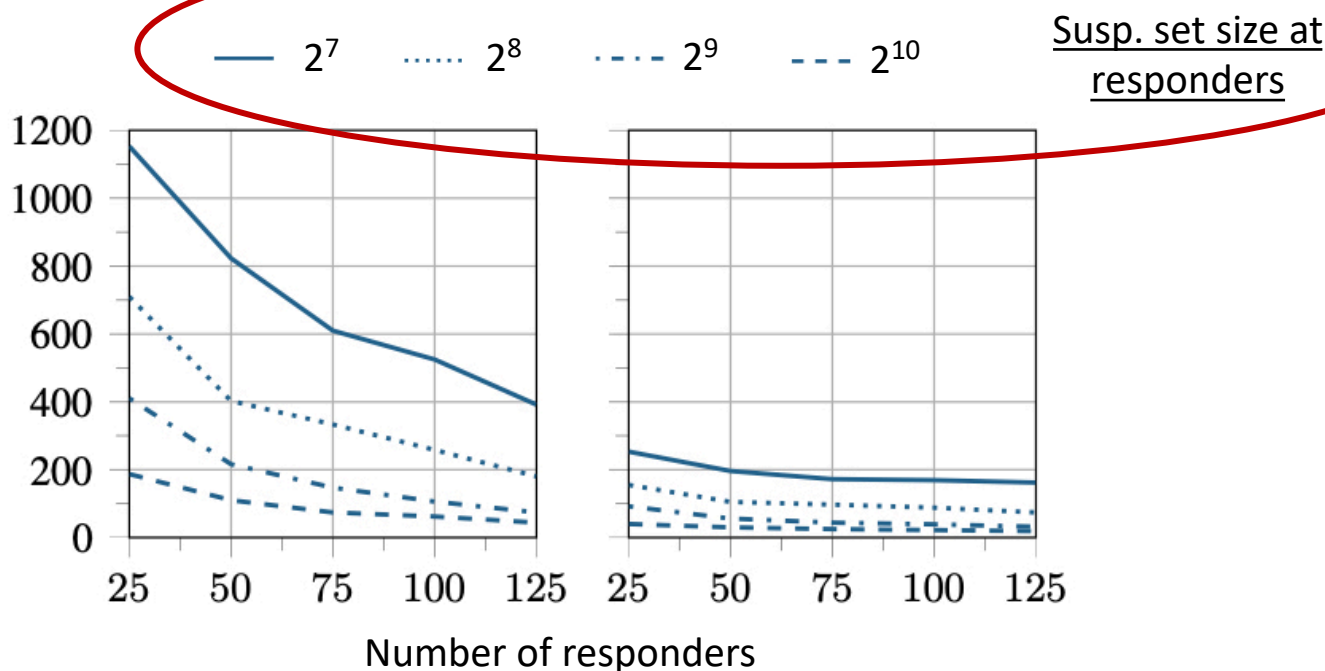


Scalability



Scalability

Max. qualifying
responses per sec.



Trusted directory
for *login privacy*
(Qualifying response: $\leq 5s$)

Untrusted directory
for *login privacy*
(Qualifying response: $\leq 8s$)



Scalability

| | Credential-stuffing login attempts per day | Proportion that succeed | Proportion of all login attempts |
|------------------|--|-------------------------|----------------------------------|
| Airline | 1.4 Million | 1.00% | 60% |
| Hotel | 4.3 Million | 1.00% | 44% |
| Retail | 131.5 Million | 0.50% | 91% |
| Consumer banking | 232.2 Million | 0.05% | 58% |

Table: Credential stuffing estimates for four major U.S. industries*

Total number of PMT queries per second:

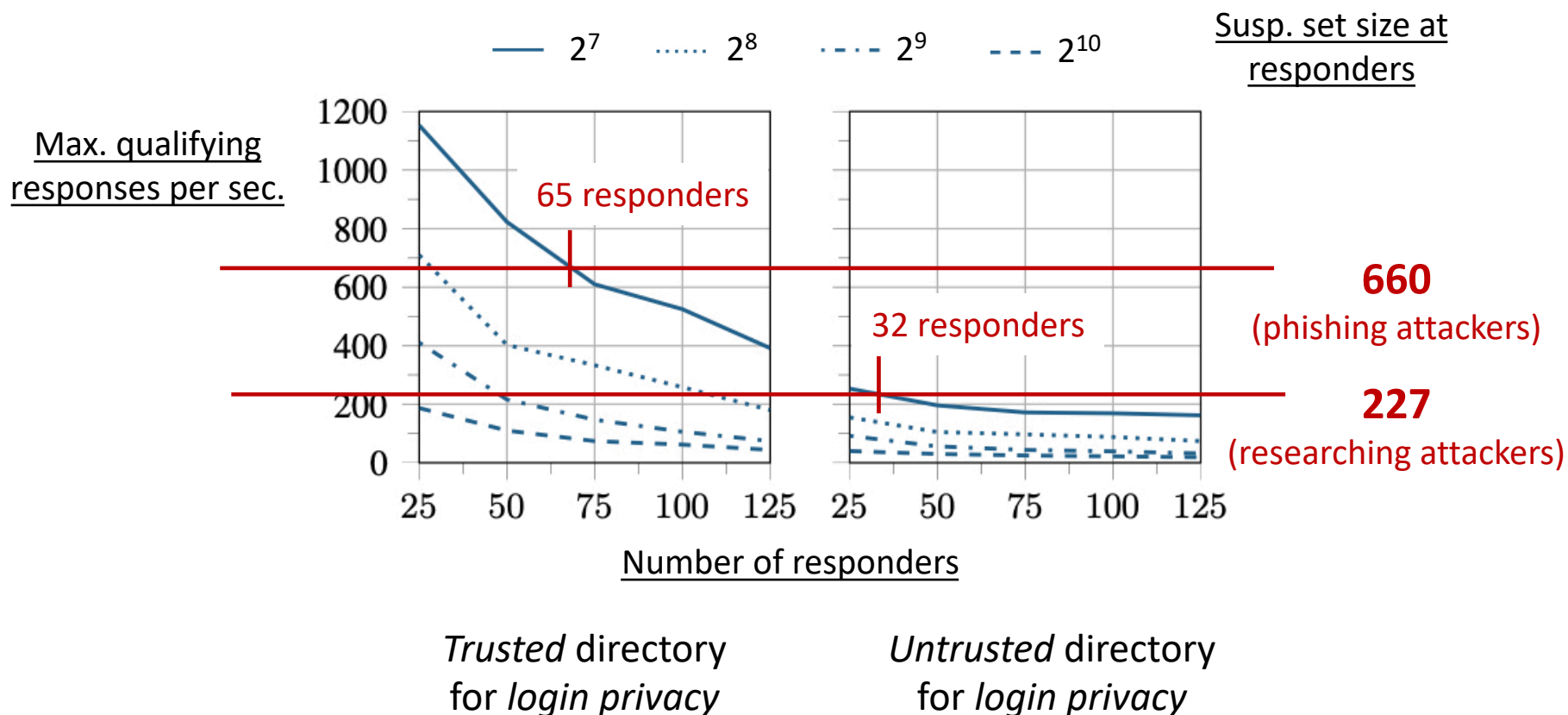
- If ADS false & true detection rates are 0.30 & 0.95 (against *phishing* attackers): **660**
- If ADS false & true detection rates are 0.10 & 0.99 (against *researching* attackers): **227**



THE UNIVERSITY
of NORTH CAROLINA
at CHAPEL HILL

* Shape Security, "2018 Credential spill report"

Scalability



Summary

- ***A framework to detect credential stuffing***



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 - *Leverages ADS and evidence trail left by credential stuffing*



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- ***First to detect active credential stuffing attacks across multiple websites***



Summary

- A framework to detect credential stuffing
 - Leverages ADS and evidence trail left by credential stuffing
 - Account security achieved by a novel PMT protocol
 - Login privacy enforced by the directory or by Tor
- First to detect active credential stuffing attacks across multiple websites
- ***Even a minimal-infrastructure deployment of our framework should support the combined login load experienced by four major sectors of the U.S economy***



Thank you!

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