

Trust in Crowds

Probabilistic Behaviour in Anonymity Protocols

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(based on joint work with S. Hamadou & E. ElSalamouny)

Introduction

Anonymity in Social Networks



Social Networks: very easy to collect private and sensitive information about individuals.

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Anonymity in Web Transactions



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Anonymity in Web Transactions



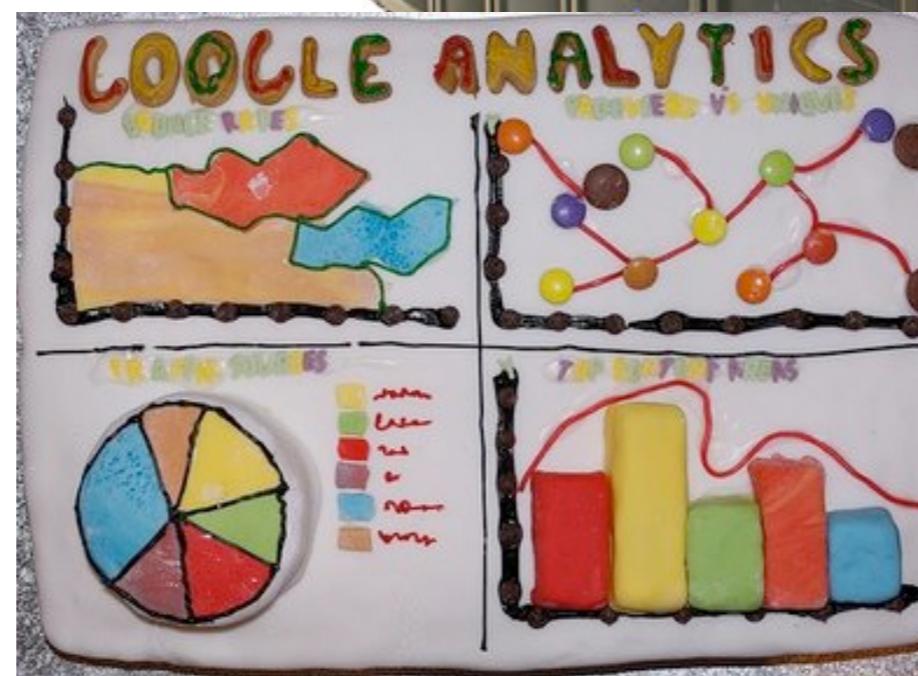
Introduction

Anonymity in Web Transactions



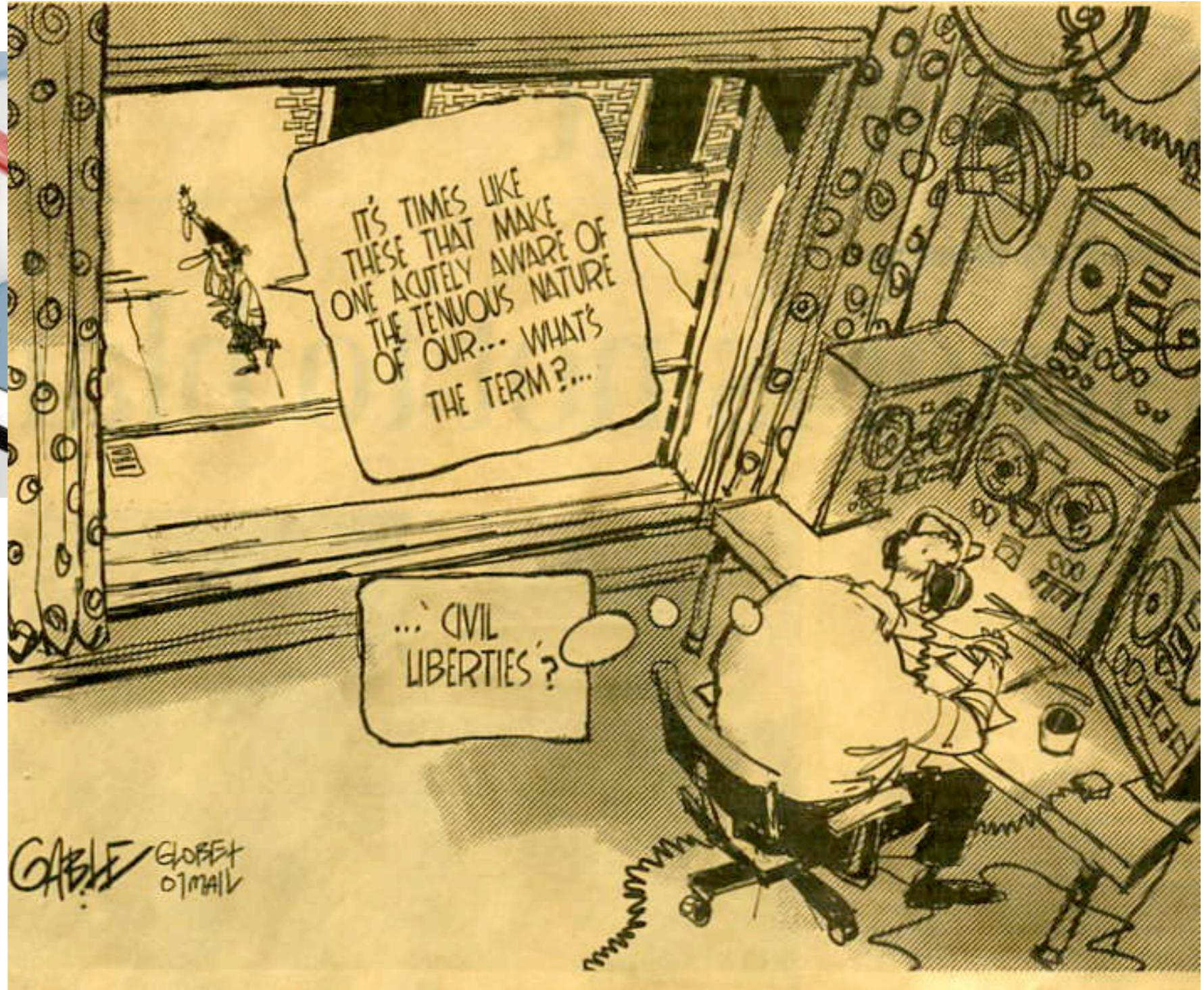
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Anonymity in Web Transactions



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Introduction

Data Confidentiality



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...of course, but also...



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Data Confidentiality



...of course, but also...



deduce high input from low output, in the fashion of information flow

Introduction

Anonymity Protocols (in general)

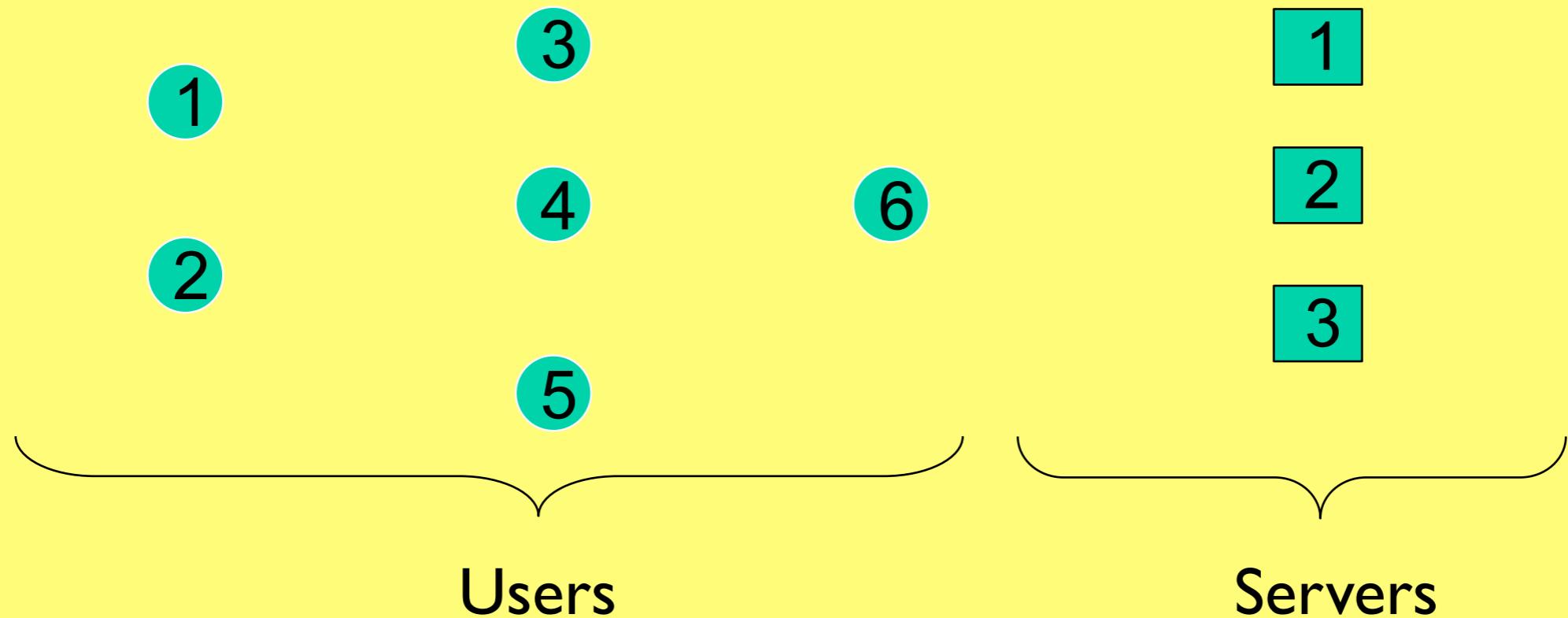
- ❖ Aims at obfuscating the link between private input (anonymous actions) and public (observable) output
- ❖ Attacker tries to infer the hidden info from his observation of the protocol

This presentation

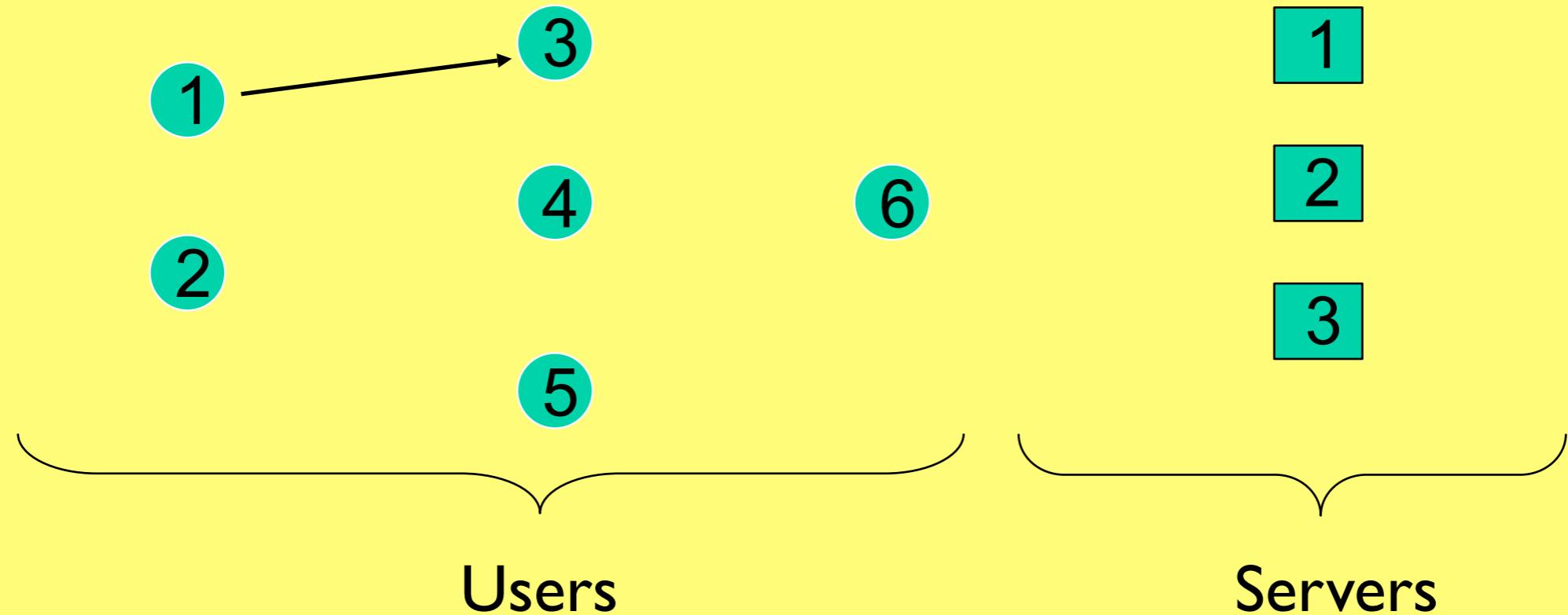
Trust in the Crowds anonymity protocol

- ❖ Extend the Crowds protocol to a scenario where:
 - ❖ Each principal may suddenly become corrupt.
 - ❖ Principal behaviour is influenced by a trust relationship.
- ❖ Work:
 - ❖ Study the impact of these assumptions on the protocol.
 - ❖ Establish necessary and sufficient criteria for choosing a policy able to achieve a desired level of privacy.

- ❖ **Crowds** [Reiter and Rubin 1998]: allows internet users to perform anonymous web transactions.

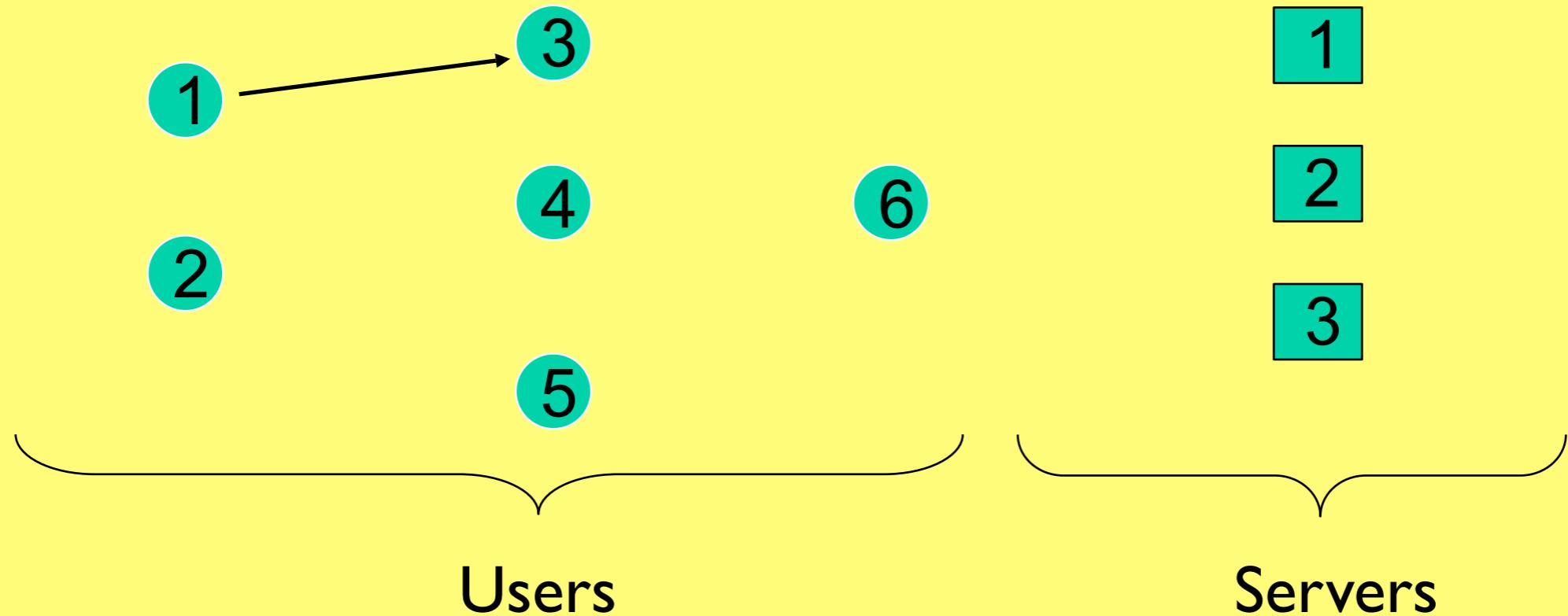


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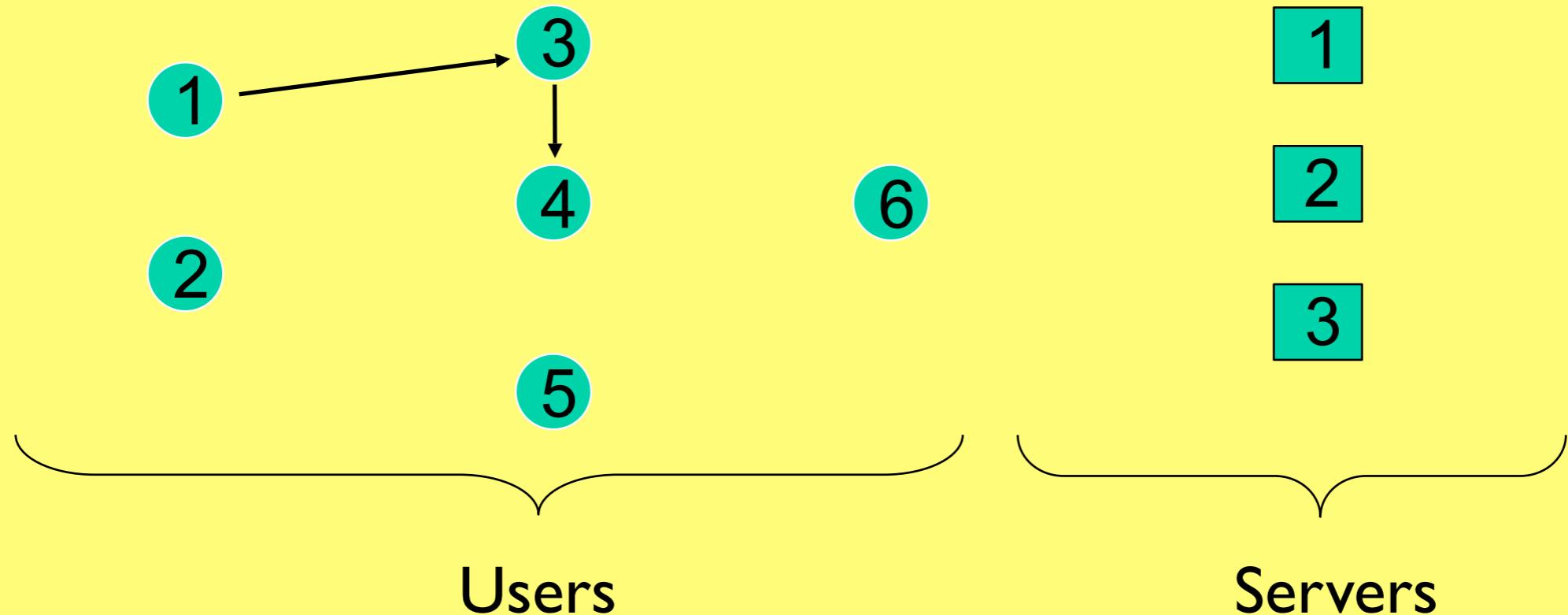
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Flips a biased coin p_f



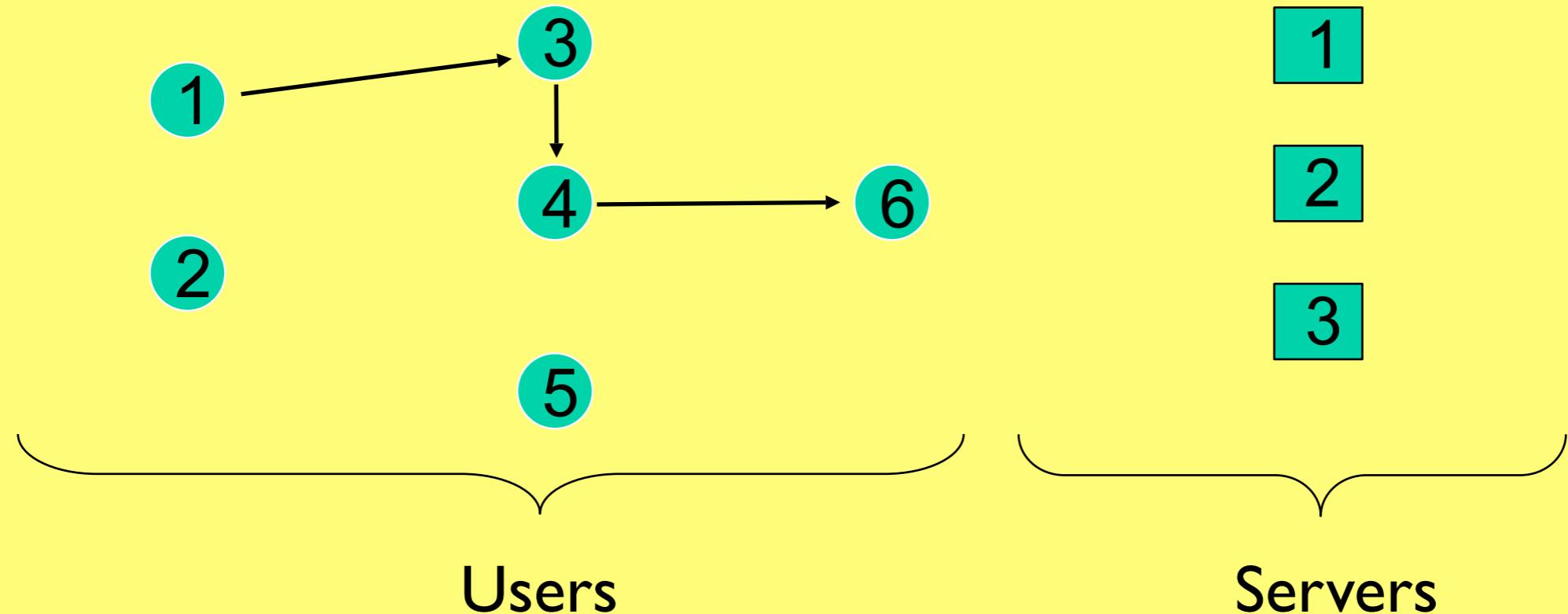
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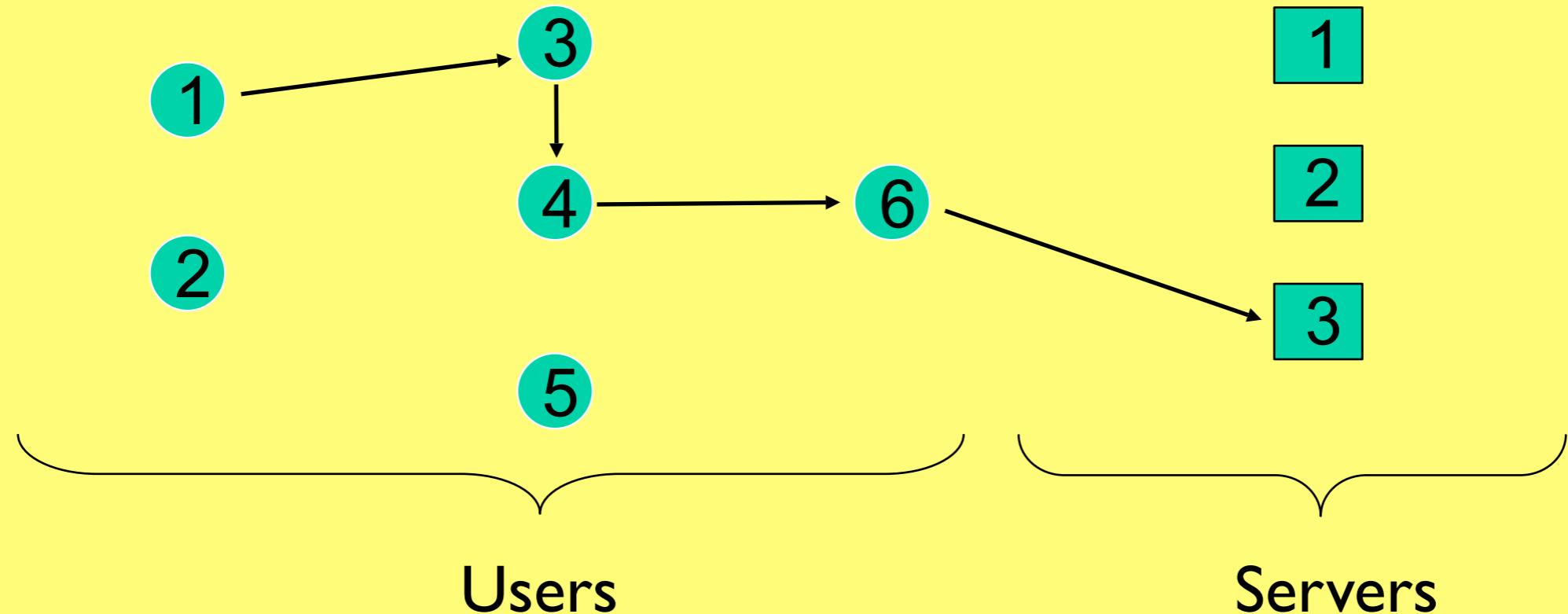
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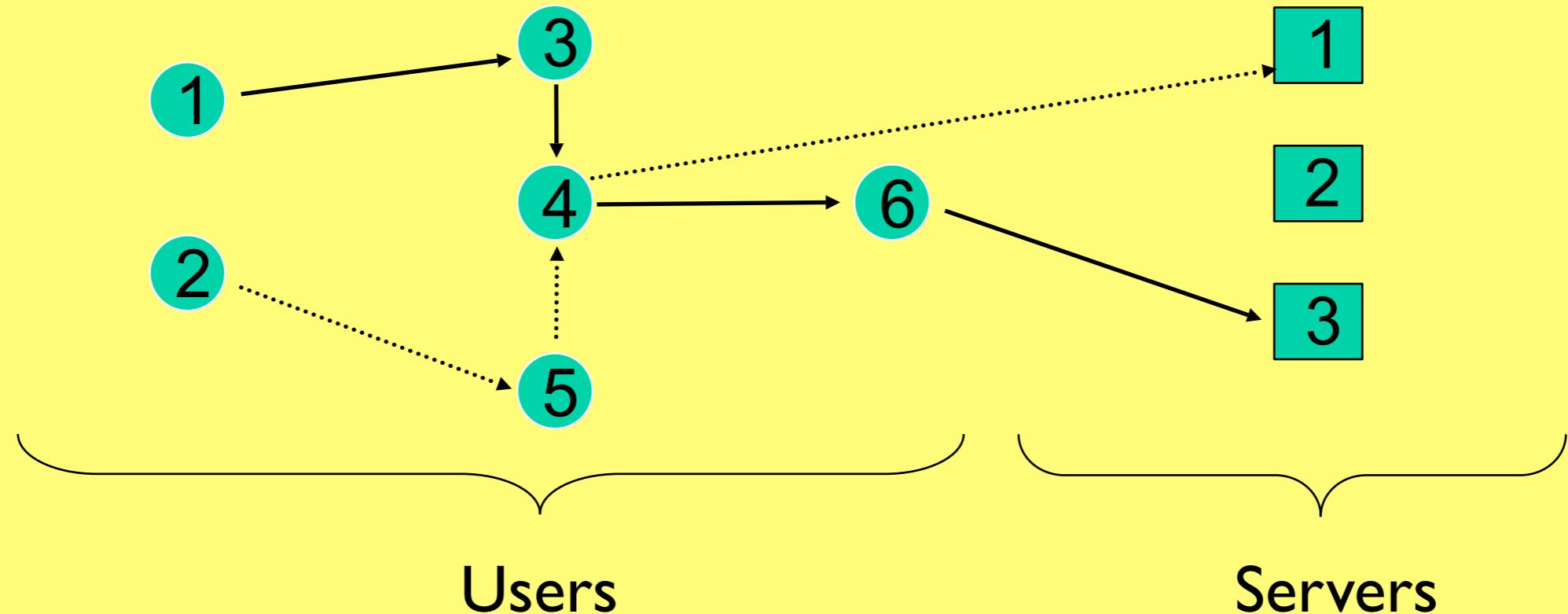
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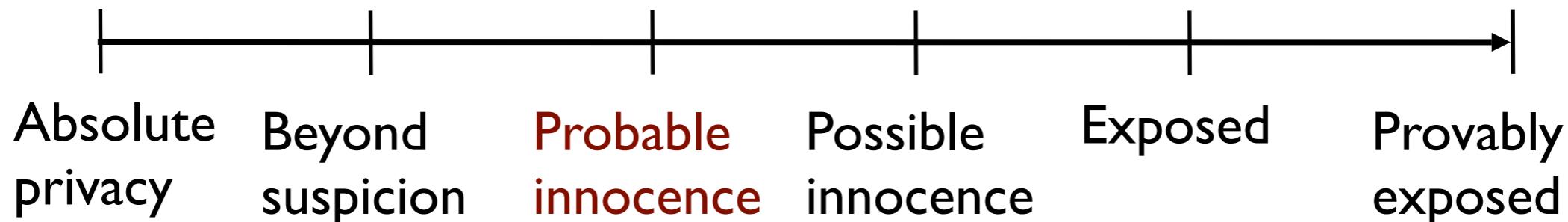
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Probable Innocence

Informal definition



“A sender is probably innocent if, from the attacker's point of view, the sender appears no more likely to be the originator than to not be the originator”

Probable Innocence

Formal definition

- ❖ Members: m members participating in the protocol
 - ❖ n honest members
 - ❖ $c = (m - n)$ corrupt members or collaborating attackers
- ❖ Anonymous events: a random variable A distributed over $\{a_1, a_2, \dots, a_n\}$, where a_i indicates that the honest user i is the initiator of the message.
- ❖ Observable events: a random variable O distributed over $\{o_1, o_2, \dots, o_n\}$, where o_i indicates that user i is honest and forwards the message to a corrupted user. In this case we say that user i is detected.

Probable Innocence

Formal definition

Definition [Reiter and Ruben, 98]:

a protocol satisfies probable innocence if

$$\forall i \ p(o_i \mid a_i) \leq 1/2$$

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Wrong

Definition [Halpern and O'Neill, 05]:

$$\forall i \ p(a_i | o_i) \leq 1/2$$

Right

Probable Innocence

Formal definition

Proposition: if the a priori distribution is uniform then

$$\forall i \ p(o_i | a_i) = p(a_i | o_i)$$

Proof: by Bayes theorem we have

$$p(o_j | a_i)p(a_i) = p(a_i | o_j)p(o_j)$$

If A is uniformly distributed then (in Crowds) O is uniformly distributed too. Hence $p(a_i) = p(o_j) = 1/n$

Probable Innocence

Extended

Definition:

a protocol satisfies α -probable innocence ($0 \leq \alpha \leq 1$) if

$$\forall i \ p(a_i \mid o_i) \leq \alpha$$

Proposition:

a protocol satisfies α -probable innocence if and only if

$$l + n(l-\alpha)/p_f \leq m$$

Overview

Trust in Crowds

- ❖ Extend the Crowds protocol to a more realistic scenario:
 - ❖ Associate to each principal i a probability $1 - t_i \in [0, 1]$ to become corrupt.
 - ❖ The forwarding process is governed by a policy $q_i \in [0, 1]$ which together with the forwarding factor p_f determines the probability that each member i is chosen as a forwarder.
- ❖ Results:
 - ❖ Analyse the impact of such probabilistic behaviour of principals.
 - ❖ Establish necessary and sufficient criteria for choosing an appropriate forwarding policy to achieve required privacy level.

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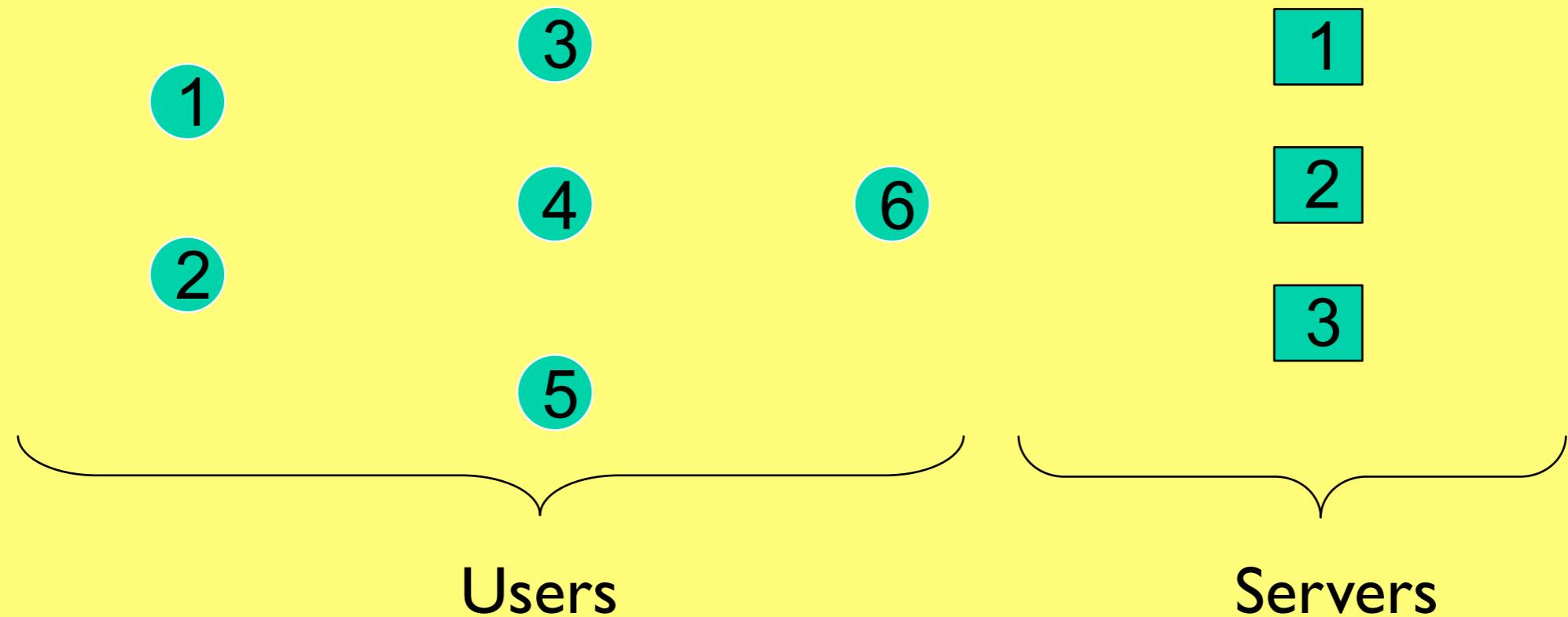
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Can be established experimentally, eg by the “*blender*” using Bayesian method, eg the Beta trust model

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The extended protocol

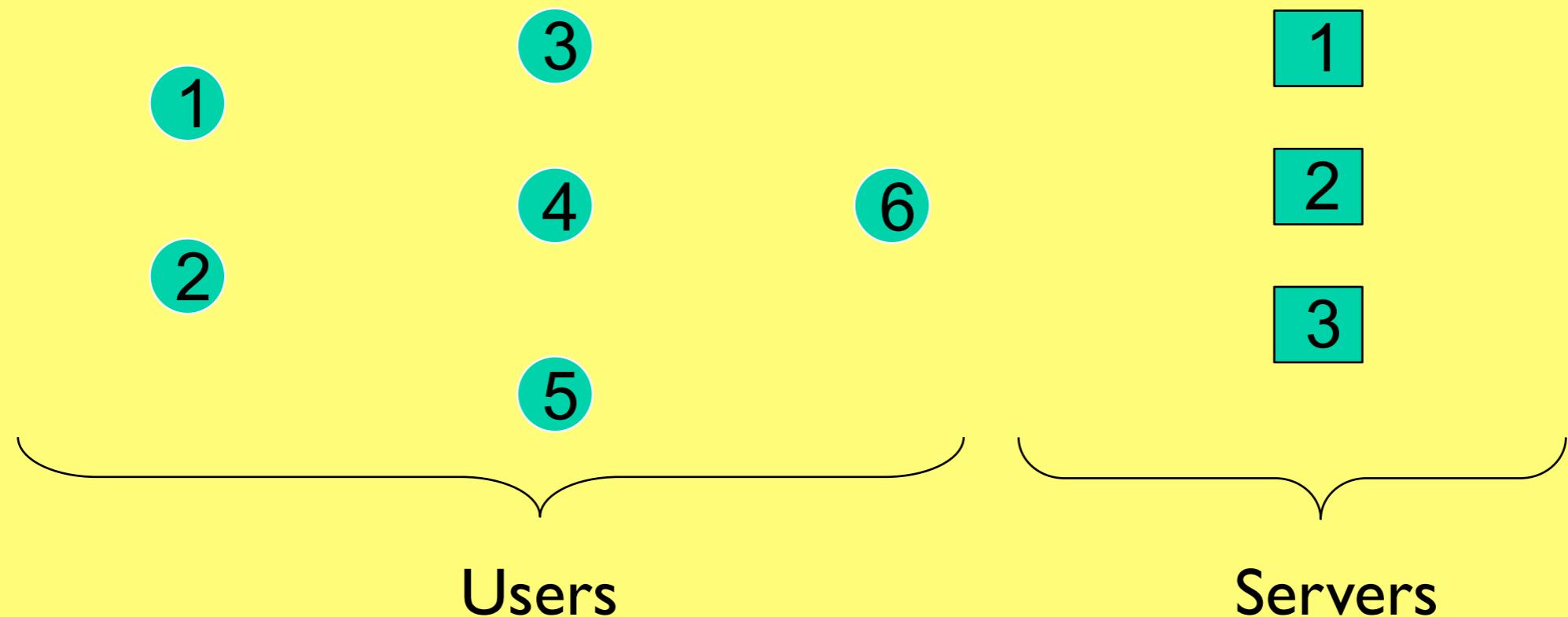
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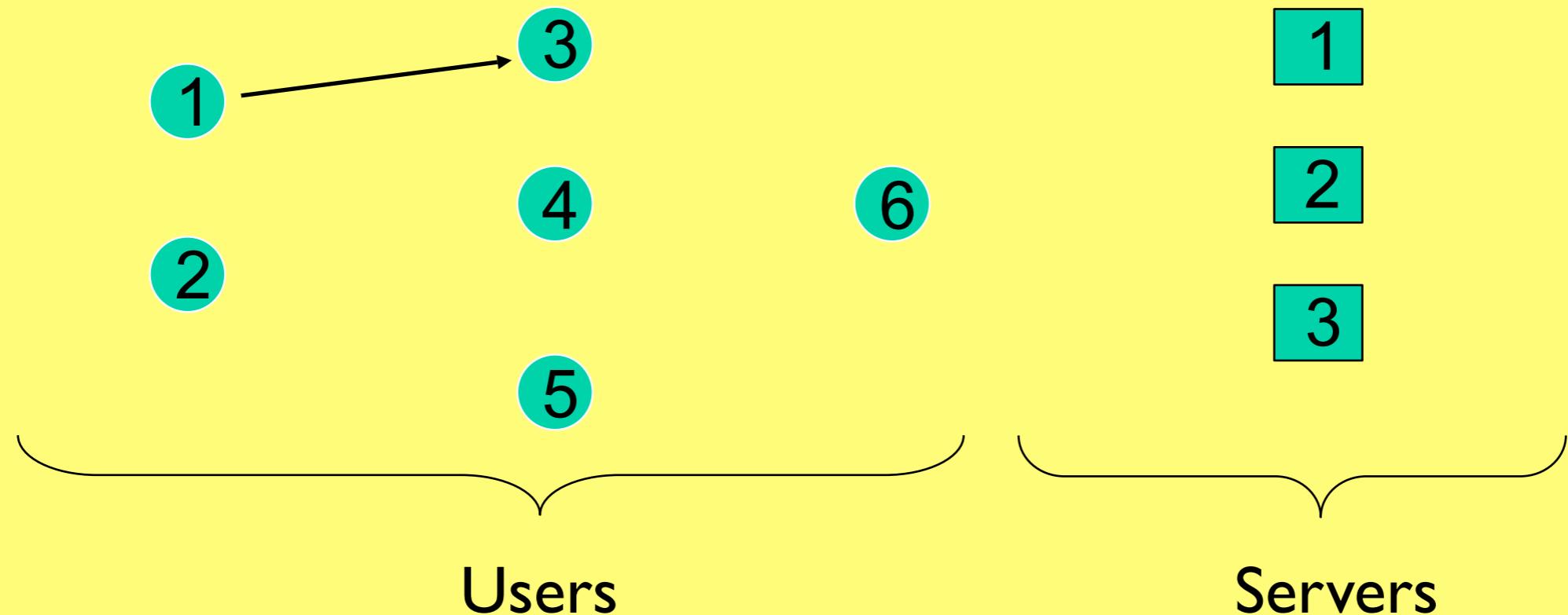
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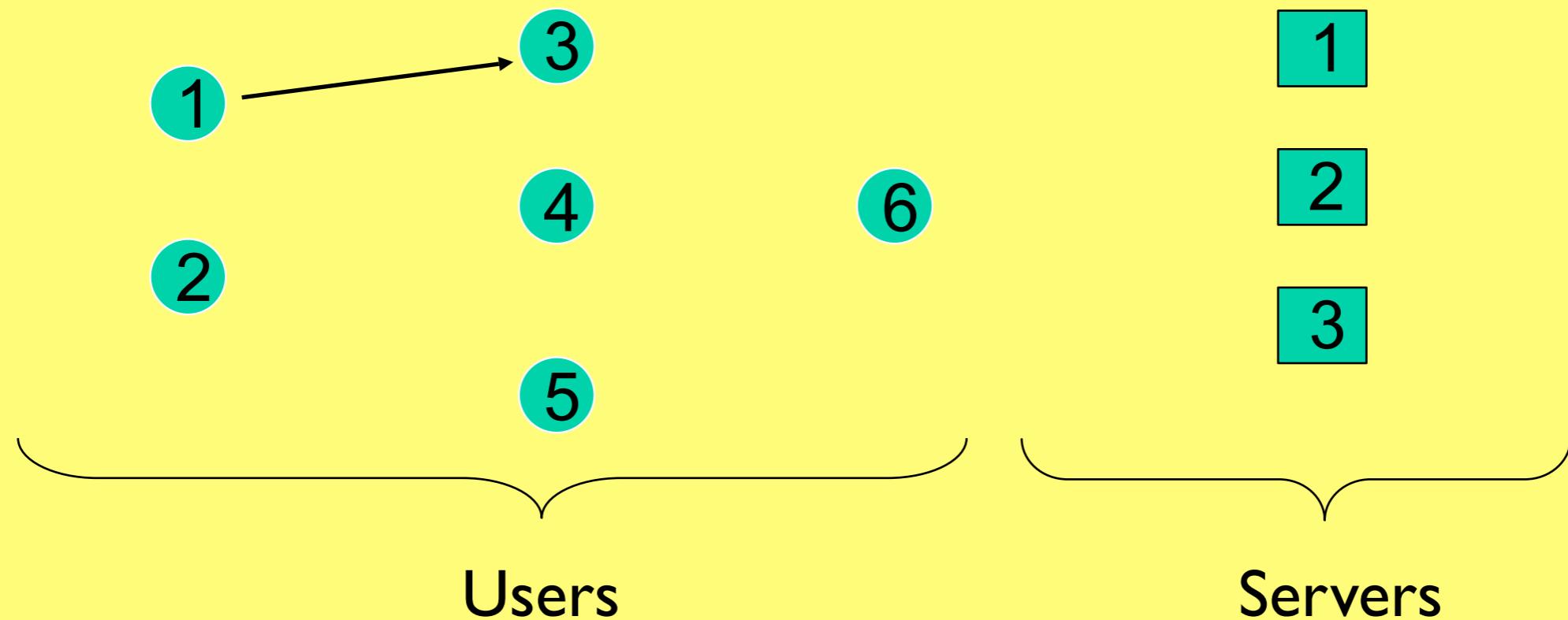


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Delivers to server with prob $1 - p_f$
Forwards to j with prob $p_f \cdot q_j$

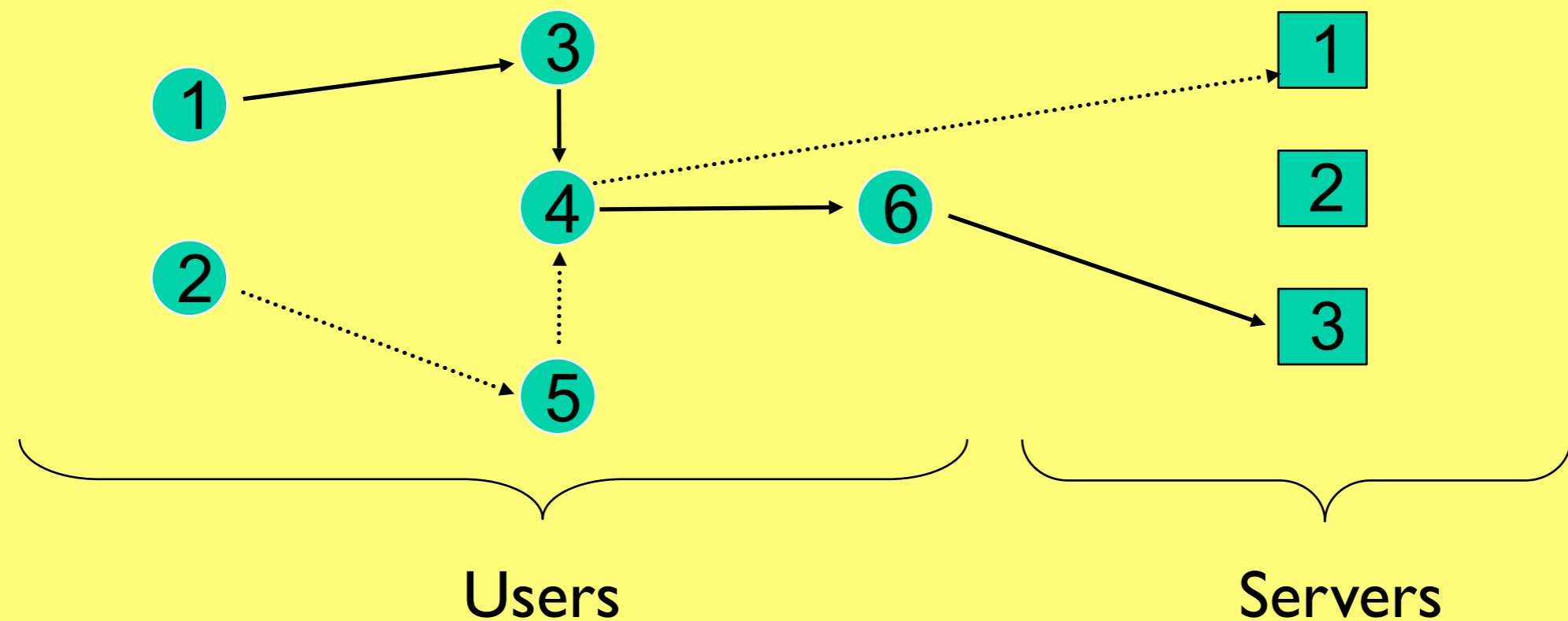


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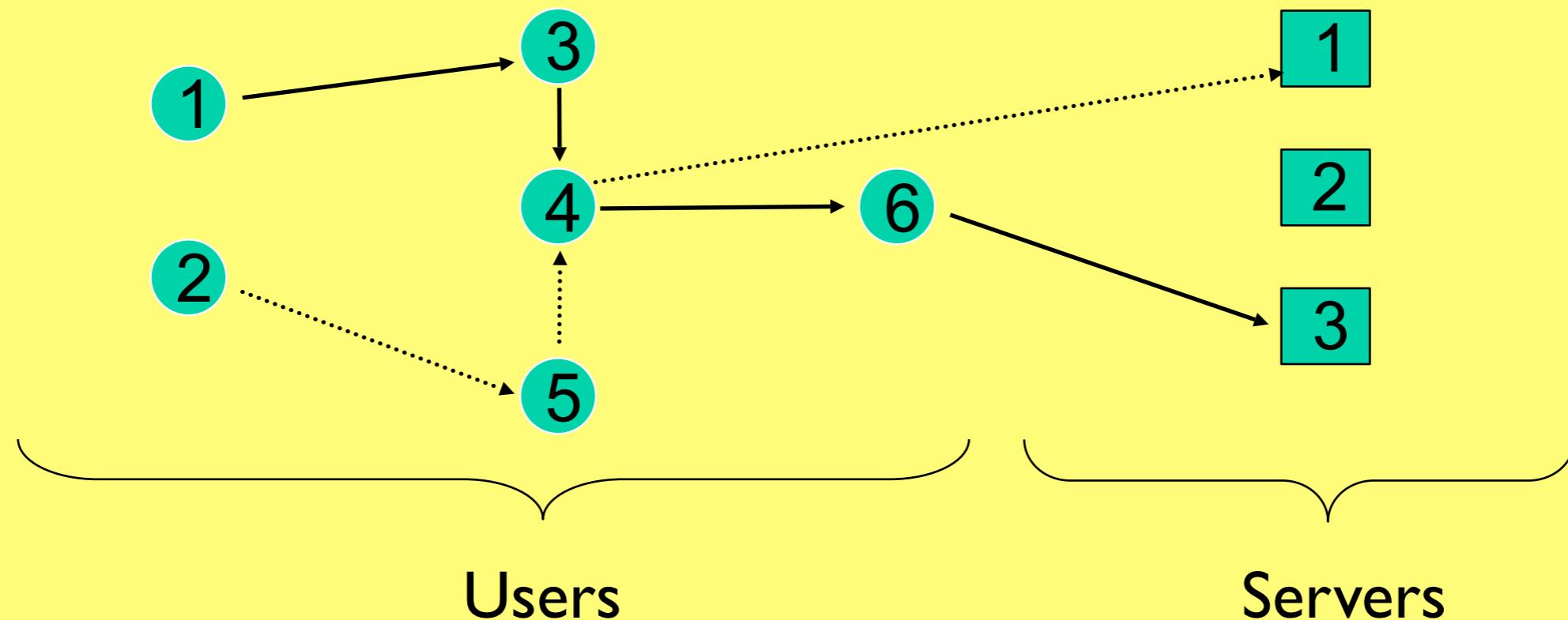


observe we assume transactions are short,
otherwise users could become corrupt
whilst answer from server travels back.

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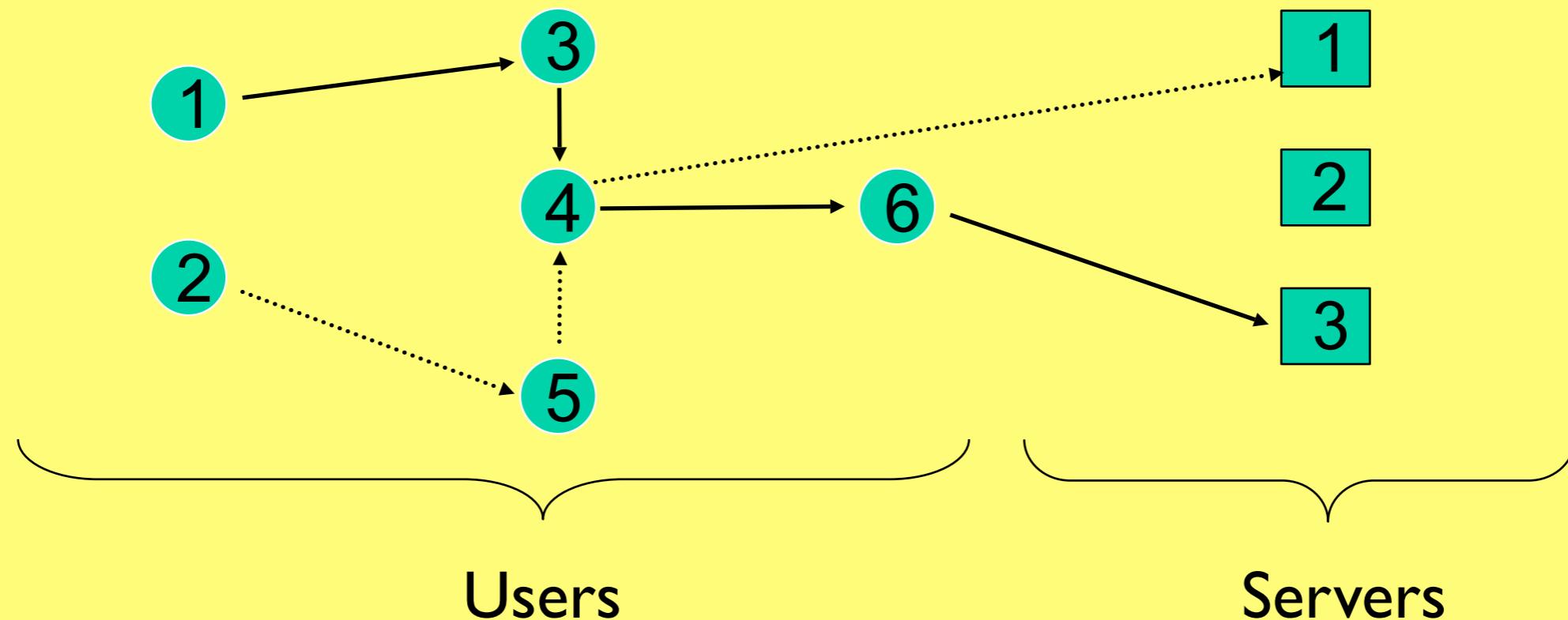


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transactions in the presence of α and β behaviours.

Initiator selects
j with prob q_j

Delivery
case is work in progress
Forwards to j with prob $p_f \cdot q_j$



Probable Innocence, again

→ Need to compute

$$P(a_i | o_i) = \frac{P(a_i, o_i)}{P(o_i)}$$

→ Start with:

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$$P(o_i, H_k) = \begin{cases} \frac{1}{n}(1 - t_i) & k = 0 \\ \frac{1}{n}t_i(1 - T) & k = 1 \\ \frac{1}{n}S T^{k-2} q_i t_i (1 - T) \cdot p_f^{k-1} & k \geq 2 \end{cases}$$

$$\text{with } S = \sum_{j=1}^n t_j \quad T = \sum_{j=1}^n q_j t_j$$

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prob to pick a
honest principal

Probable Innocence, again

→ Need to compute

$$P(a_i | o_i) = \frac{P(a_i, o_i)}{P(o_i)}$$

→ Continue with:

$$\begin{aligned} P(o_i) &= \sum_{k=0}^{\infty} P(o_i, H_k) \\ &= \frac{1}{n}(1 - t_i) + \frac{1}{n}t_i(1 - T) \\ &\quad + \sum_{k=2}^{\infty} \frac{1}{n} S T^{k-2} \cdot q_i t_i (1 - T) p_f^{k-1} \\ &= \frac{1}{n} \left(1 - t_i T + S p_f q_i t_i \left(\frac{1 - T}{1 - p_f T} \right) \right) \end{aligned}$$

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observe this is 0
iff $T=1$ and $t_i=1$
 i is undetectable

Probable Innocence, again

→ Need to compute

$$P(a_i | o_i) = \frac{P(a_i, o_i)}{P(o_i)}$$

→ Similarly:

$$\begin{aligned} P(a_i, o_i) &= \sum_{k=0}^{\infty} P(a_i, H_k, o_i) \\ &= \frac{1}{n}(1 - t_i) + \frac{1}{n}t_i(1 - T) \\ &\quad + \sum_{k=2}^{\infty} \frac{1}{n}t_iT^{k-2} \cdot q_i t_i (1 - T) p_f^{k-1} \\ &= \frac{1}{n} \left(1 - t_i T + p_f q_i t_i^2 \left(\frac{1 - T}{1 - p_f T} \right) \right) \end{aligned}$$

Probable Innocence, again

→ Need to compute

$$P(a_i | o_i) = \frac{P(a_i, o_i)}{P(o_i)}$$

→ And therefore:

$$P(a_i | o_i) = \frac{1 - t_i T + p_f q_i t_i^2 \left(\frac{1-T}{1-p_f T} \right)}{1 - t_i T + S p_f q_i t_i \left(\frac{1-T}{1-p_f T} \right)}$$

→ Observe that if i is detectable, this quantity is positive: ie, it can always be caught when i is the initiator: Crowds never achieves “*absolute privacy*”

Probable Innocence, again

→ Need to compute

$$P(a_i | o_i) = \frac{P(a_i, o_i)}{P(o_i)}$$

also observe that when $T = 1 - c/n$ and $S = n - c$, which characterise the (standard) Crowds, then this formula simplifies to the standard one.

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Proposition: (*Provably Exposed Principals*)

For all users s.t. $p(o_i) \neq 0$, we have $p(a_i | o_i) = 1$
iff one of the following holds.

1. $p_f = 0$
2. $t_i = 0$
3. $q_i = 0$
4. $T = 1$
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all participants
are honest!

all but i are
corrupt!

Theorem: *(Monotonicity in forwarding)*

$p(a_i | o_i)$ is a decreasing function of p_f

Corollary: *(Anonymity range)*

$$\forall i. P(a_i | o_i) \geq 1 - \frac{q_i t_i \sum_{j \neq i}^n t_j}{1 - t_i \sum_{j \neq i}^n q_j t_j + q_i t_i \sum_{j \neq i}^n t_j}$$

Theorem: (Monotonicity in forwarding)

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On Forwarding

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tells us that $p_f = 1$ minimises $p(a_i | o_i)$. But then the message never reaches...

Theorem: (α -Probable Innocence)

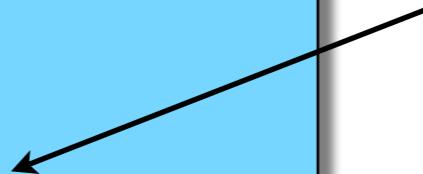
For all $\alpha \in [0, 1]$, the extended protocol guarantees α -probable innocence to all its participants if

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observe that this provides a system of linear inequalities that can be solved in q_i to try and achieve α -probable innocence

Achieving α -Probable Innocence

Maintain the lower bound on $p(a_i | o_i) = l$ below α by manipulating the forwarding distribution (*social policy*), or by excluding untrustworthy participants (*rational policy*).

Example: Suppose $t_1 = 0.70$, $t_2 = 0.97$, $t_3 = 0.99$

For $\alpha = l/2$ the system admits two solutions, eg

$$q_1 = 0.4575, \quad q_2 = 0.2620, \quad q_3 = 0.2805 .$$

Observe how user **1** is helped (at the others' risk!) to offset its higher tendency to corruption. Indeed, probable innocence in (standard) Crowds cannot be achieved.

The alternative, is for **2** and **3** to exclude **1** and yield higher overall security.

Conclusion & Further Work

- We have extended *Crowds* to take into account that principals are not usually either honest or malicious, but are liable to become *corrupt* (and again *uncorrupt*). Ours is the first attempt to cope with such probabilistic behaviour.
- Our forwarding policies can be used to make the protocol more secure (either *socially* or *rationally*) once an estimation of trust is available. A lot more work on integrating trust estimation is to be done.
- A deeper analysis of trust is likely to be possible on advanced anonymity protocols such as *Tarzan* and *ToR*.
- We are in the process of complete this analysis by *dropping* the hypothesis of short transactions.

Related Work

Crowds & External knowledge

- ❖ Real world: attackers usually gather additional information correlated to the anonymous agents before attacking the protocol.
- ❖ Example: two agents voting by “yes” or “no” and the result of the vote is {yes, no}
 - ❖ Agents used different colours but the adversary does not know the correlation between the colors and the agents:
 $\{yes, no\} \equiv \{yes, no\}$
 - ❖ The adversary knows the correlation: $\{yes, no\} \neq \{yes, no\}$

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analysis of the impact of attackers' extra knowledge on the security of information hiding protocols.

in FAST 2009
with C. Palamidessi

Related Work

Crowds & Beliefs & Vulnerability

- ❖ Open problem: measure and account for the **accuracy** of the adversary extra knowledge.
- ❖ Integrate the notion of adversary's beliefs:
 - ❖ Assume both actual a priori distribution of the hidden input and its correlation to the extra information unknown to adversary.
 - ❖ Generalise the approach to information flow systems.
- ❖ Results:
 - ❖ New metric for quantitative information flow based on the concept of vulnerability that takes into account the adversary's beliefs.
 - ❖ Model allows to identify the levels of accuracy for the adversary's beliefs which are compatible with the security of a given program or protocol.

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 - ❖ Model allows to identify the levels of accuracy for the adversary's beliefs which are compatible with the security of a given program or protocol.

in *IEEE Symp on Security & Privacy 2010*
with C. Palamidessi