



# Trust Management for Host-based Collaborative Intrusion Detection

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# Outline

- Introduction
- Motivation
- Trust Model
- Robustness
- Attacks Prevention
- Conclusions

# Introduction

- The worldwide impact of malicious intrusions is estimated to be over \$10 Billion annually
- Intrusions include viruses/worms, spyware, spam, DoS, unauthorized login.
- Traditional isolated IDSes are inefficient in detection unknown intrusions.



# Benefits of IDS Collaboration

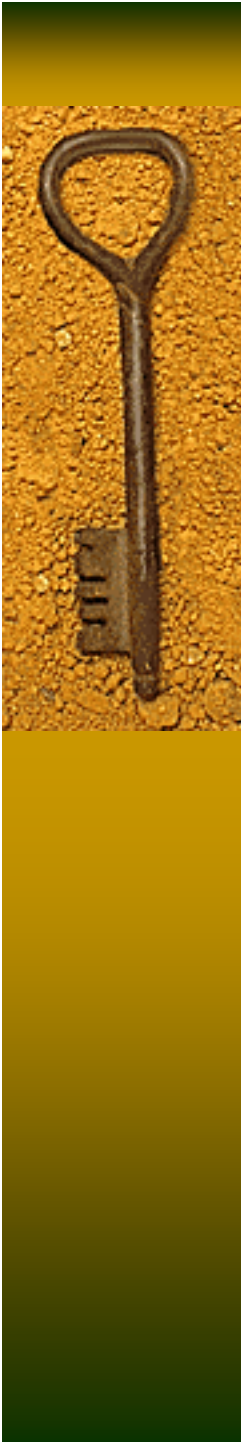
- Accurate alert ranking
- Effective intrusion detection and prevention
- Worm spreading warning

# Related Work

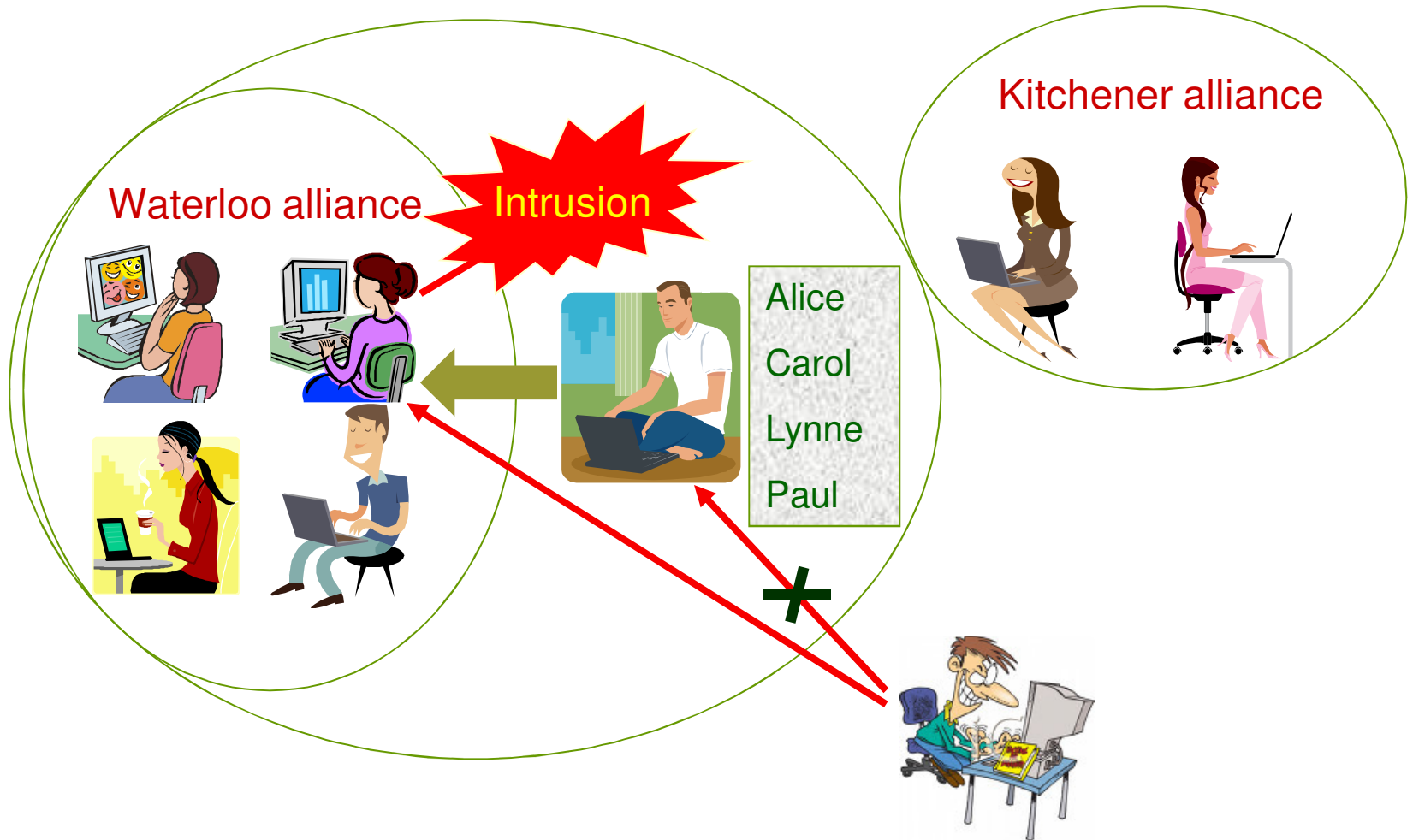
- Current collaborative architectures have strong assumptions on the trustworthiness (all IDSes are trusted and faithfully report intrusion events) [1,2,3,4]
- Used trust models are naïve [5, 6]
- Many efficient trust management models in other areas such as e-market and P2P networks [7,8,9]

# Contribution

1. Build a trust management model for IDS collaboration
2. Propose a framework for efficient collaboration of IDSes



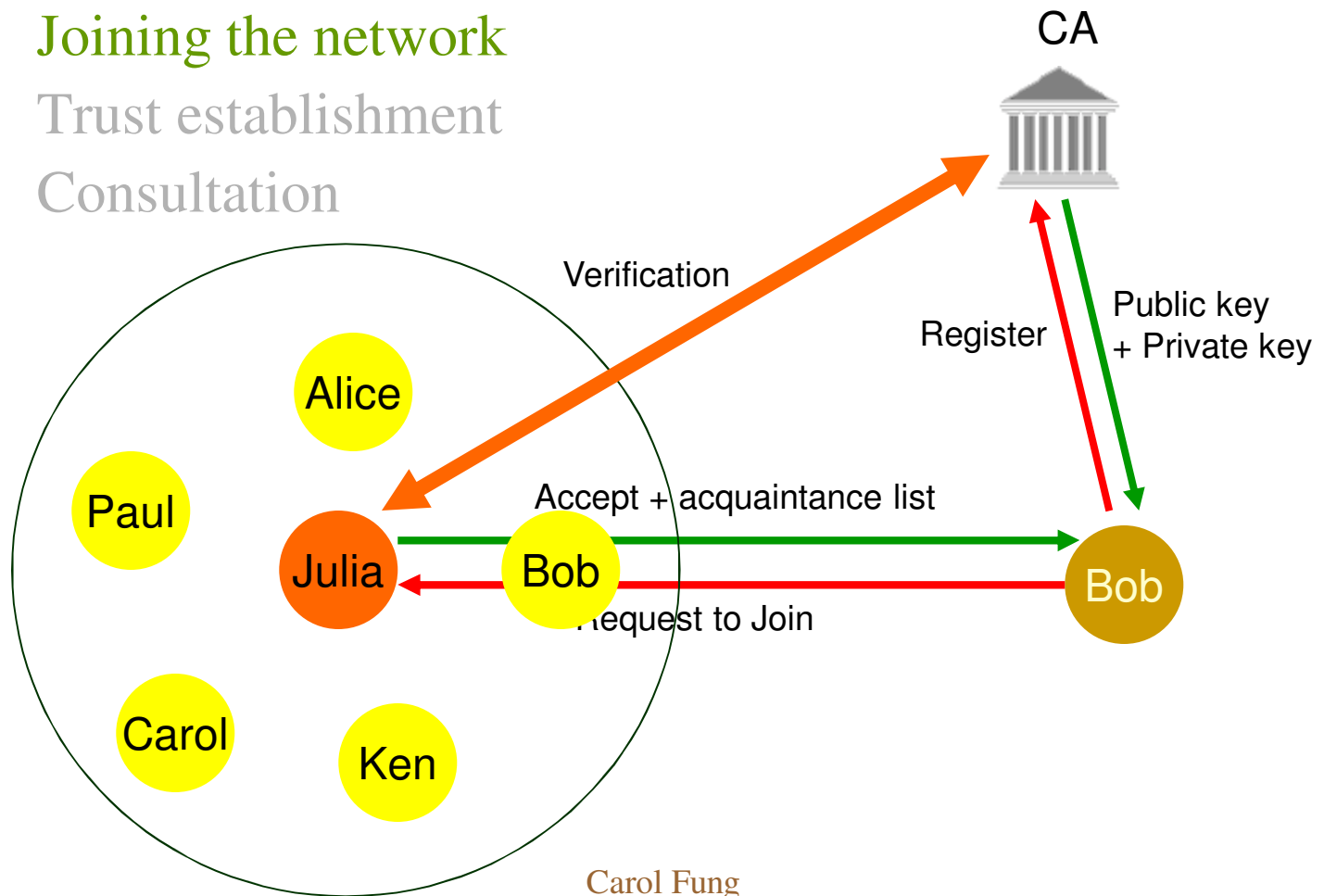
# Scenario



# Framework

Three components:

- Joining the network
- Trust establishment
- Consultation



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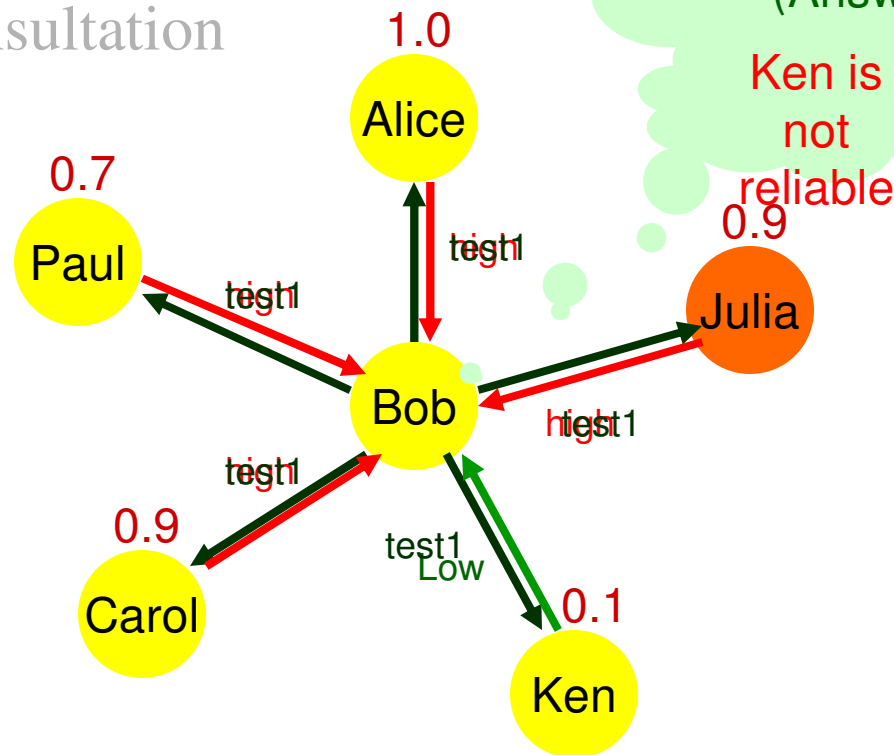




# Framework

Three components:

- Join the network
- Trust establishment
  - Test phase
- Consultation



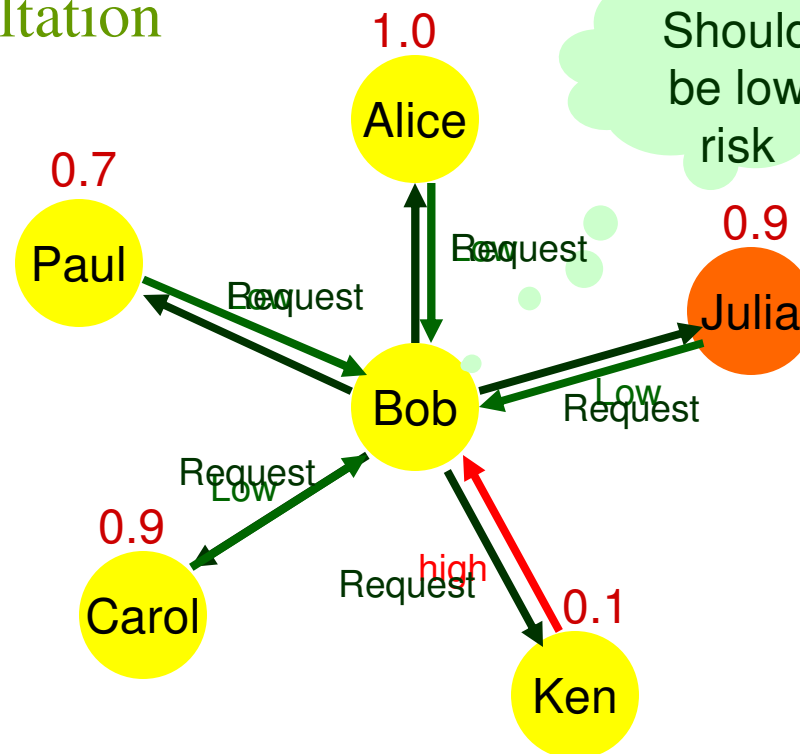
What is the risk of this alert?  
<Alert description >  
(Answer: High)

Ken is not reliable

# Framework

Three components:

- Join the network
- Trust establishment
- Consultation



What is the risk of this alert?  
<Alert description...>  
(Alert Case)  
Should be low risk

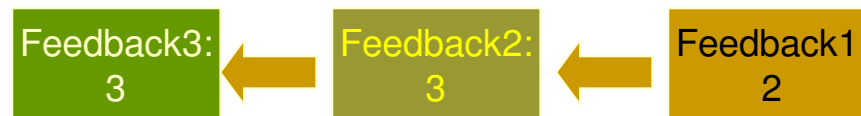
# Trust Establishment

Trust level is built on history

- Satisfactory level of past feedbacks
- Helpfulness

Feedback	Expected Answer	Satisfaction Level
High	High	3
Medium	High	1
Low	High	0

Paul's History



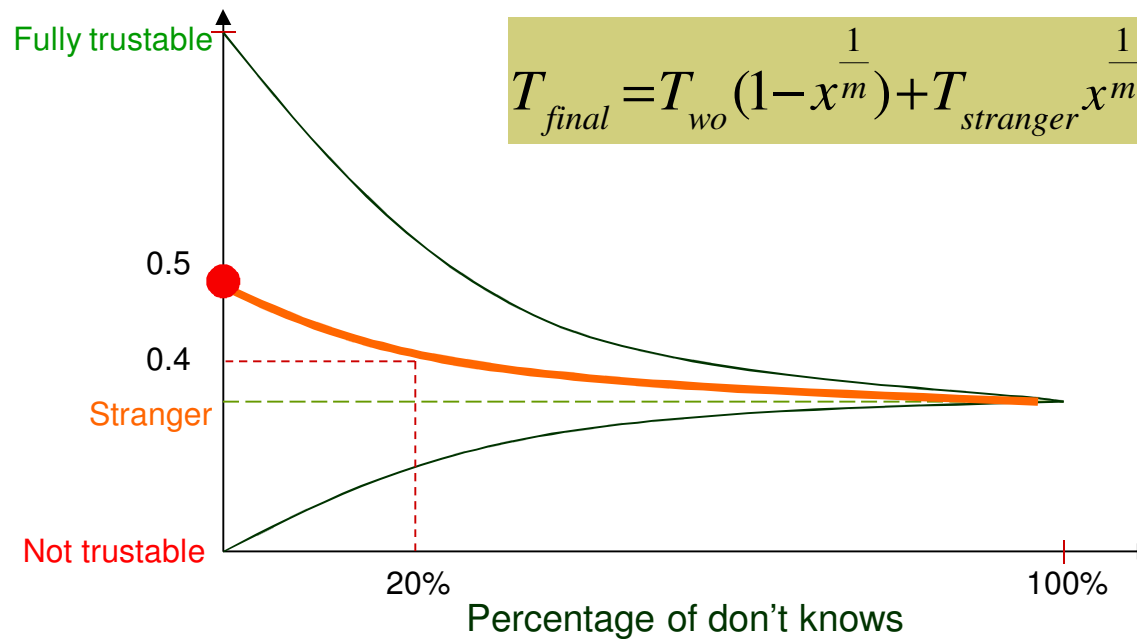
~~$TrustLevel = \frac{3 + 3 + 2}{3 + 3 + 3}$~~   $\rightarrow$   $TrustLevel = \frac{3 + 3\lambda + 2\lambda^2}{3 + 3\lambda + 3\lambda^2}$



# Integration of Don't Knows

Reply “don't know” is allowed

- Trust value will approach to the level of stranger



# Feedback Aggregation

Depends on:

- Peers' trust values
  - Trust weight
- Peer's location
  - Proximity weight



# Feedback Aggregation

- Weighted average
- Threshold

Name	Trust	Proximity	Ranking
Alice	1	1 (Waterloo)	High(3)
Carol	0.9	1 (Waterloo)	High(3)
Julia	0.9	0.9 (Toronto)	High(3)
[REDACTED]			
Paul	0.7	0.7 (US)	Medium(2)

High Risk!

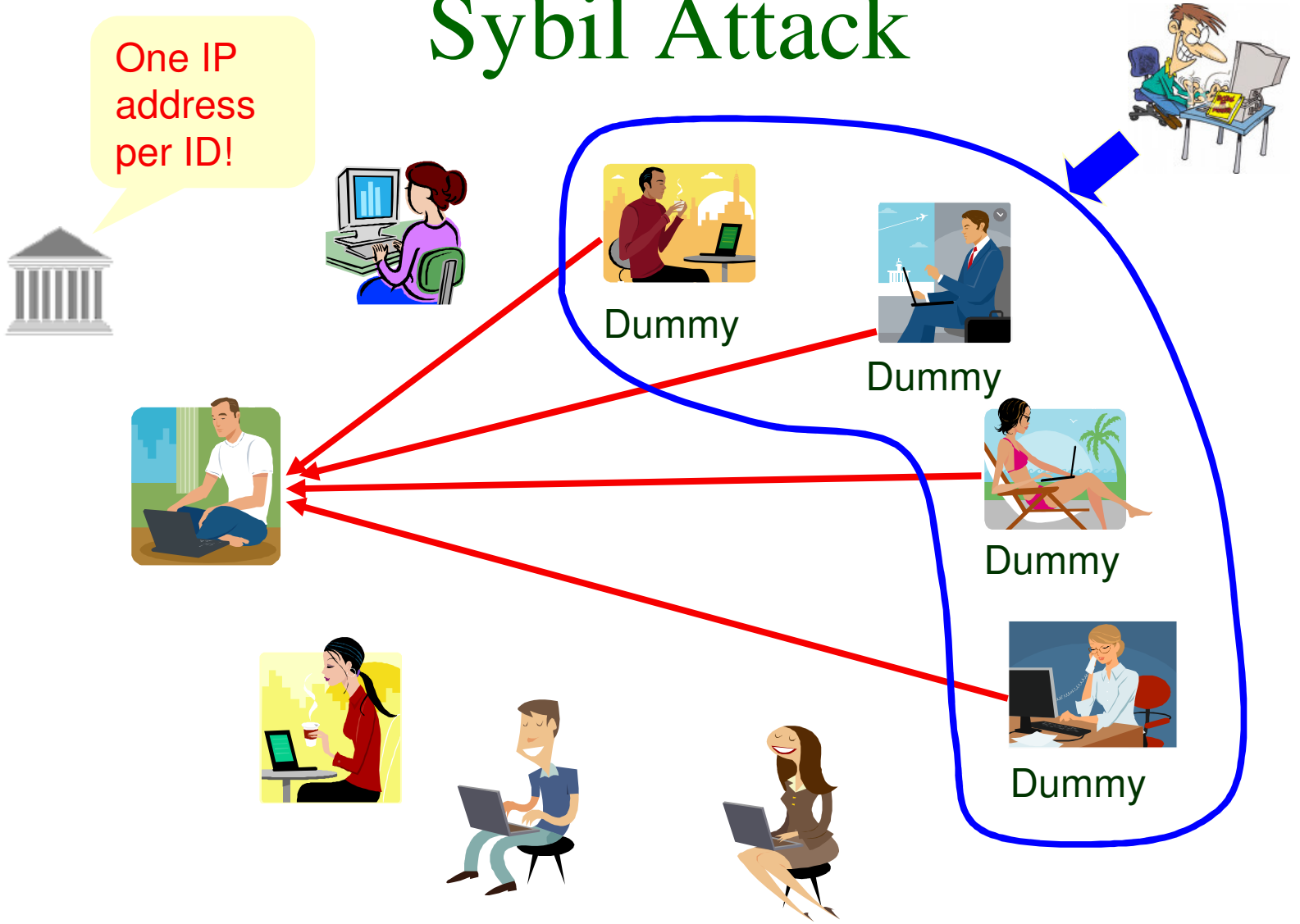
$$\begin{aligned} \text{finalRanking} &= \frac{3 \cdot 1 \cdot 1 + 3 \cdot 0.9 \cdot 1 + 3 \cdot 0.9 \cdot 0.9 + 2 \cdot 0.7 \cdot 0.7}{1 \cdot 1 + 0.9 \cdot 1 + 0.9 \cdot 0.9 + 0.7 \cdot 0.7} \\ &= 2.85 \end{aligned}$$

# Threat Model

- Sybil Attack
- New Comer Attack
- Identity Cloning Attack
- Betrayal Attack
- Collusion Attack

# Sybil Attack

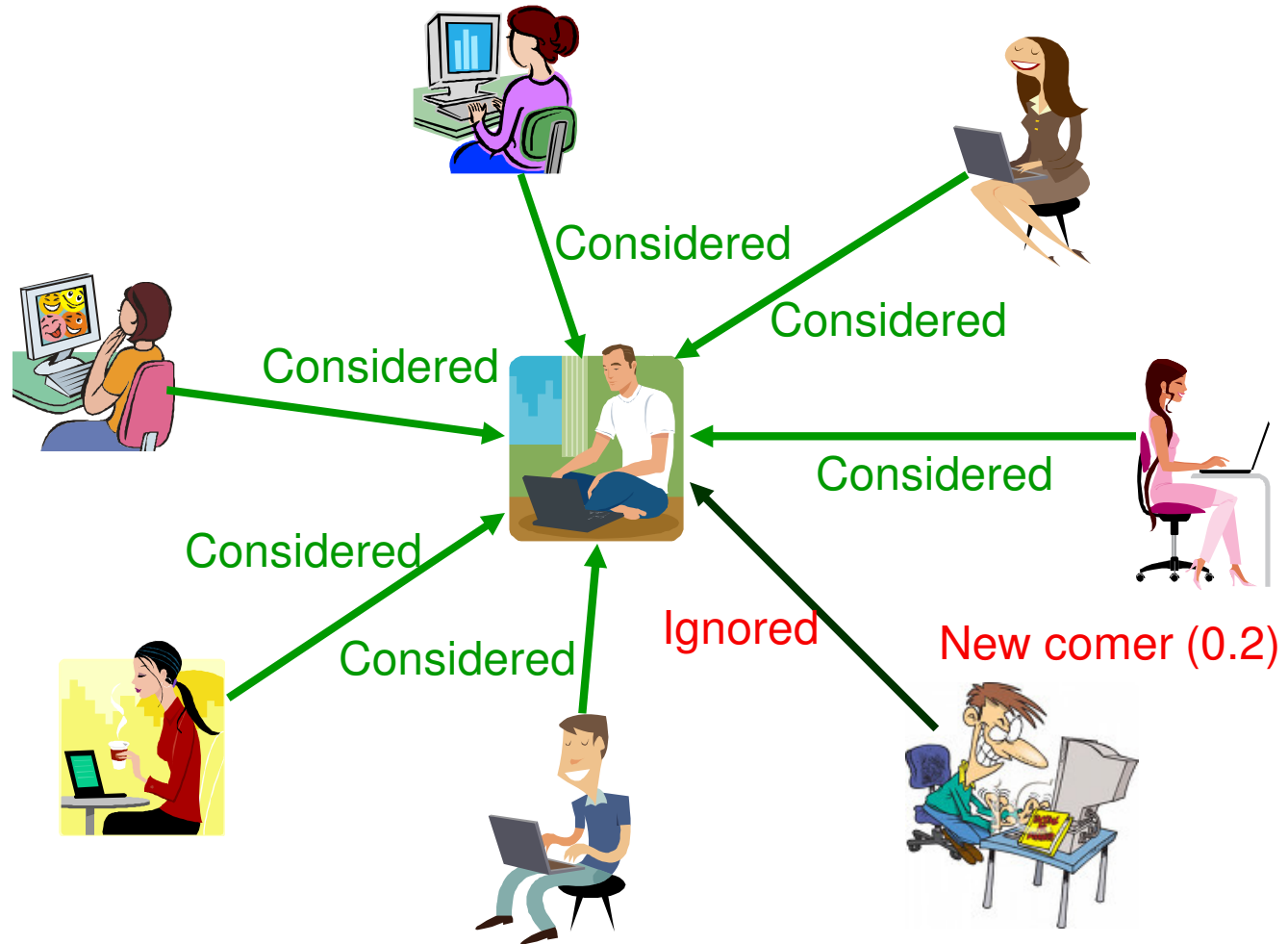
One IP address per ID!



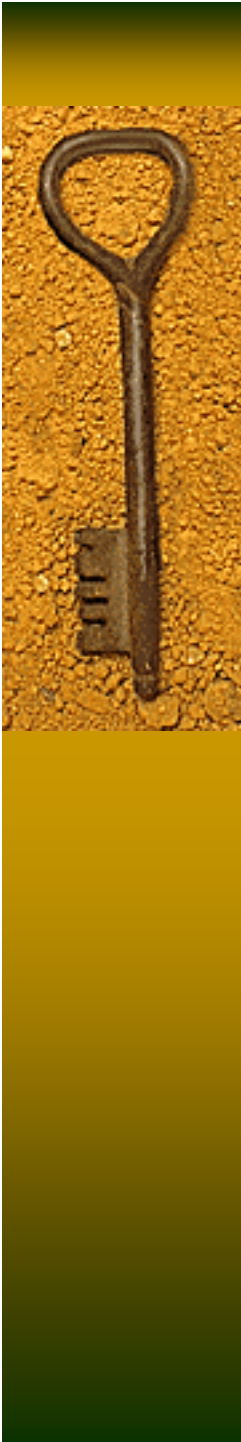




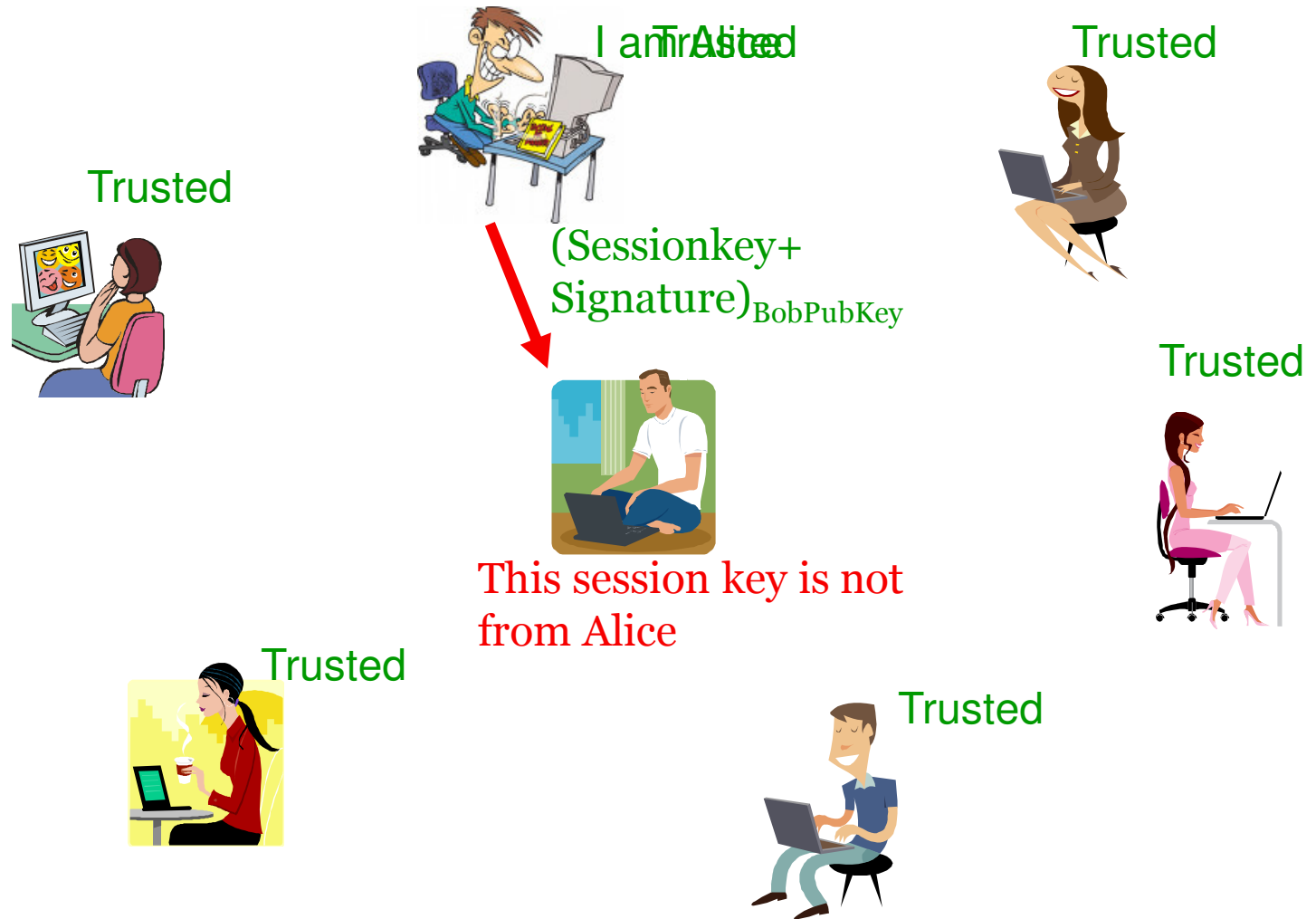
# New Comer Attack

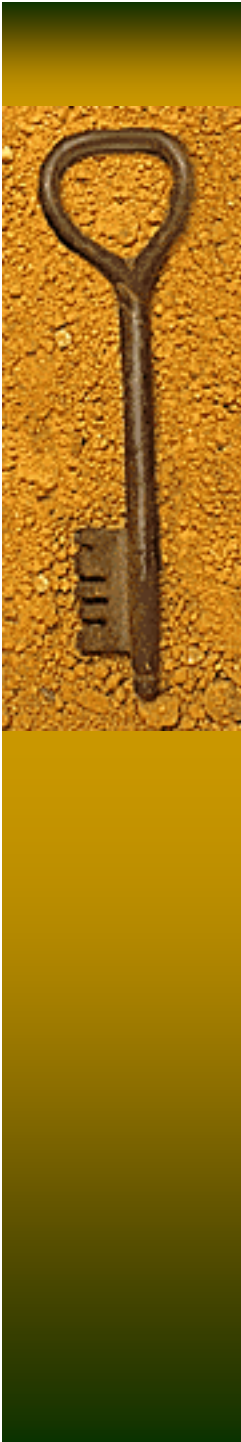


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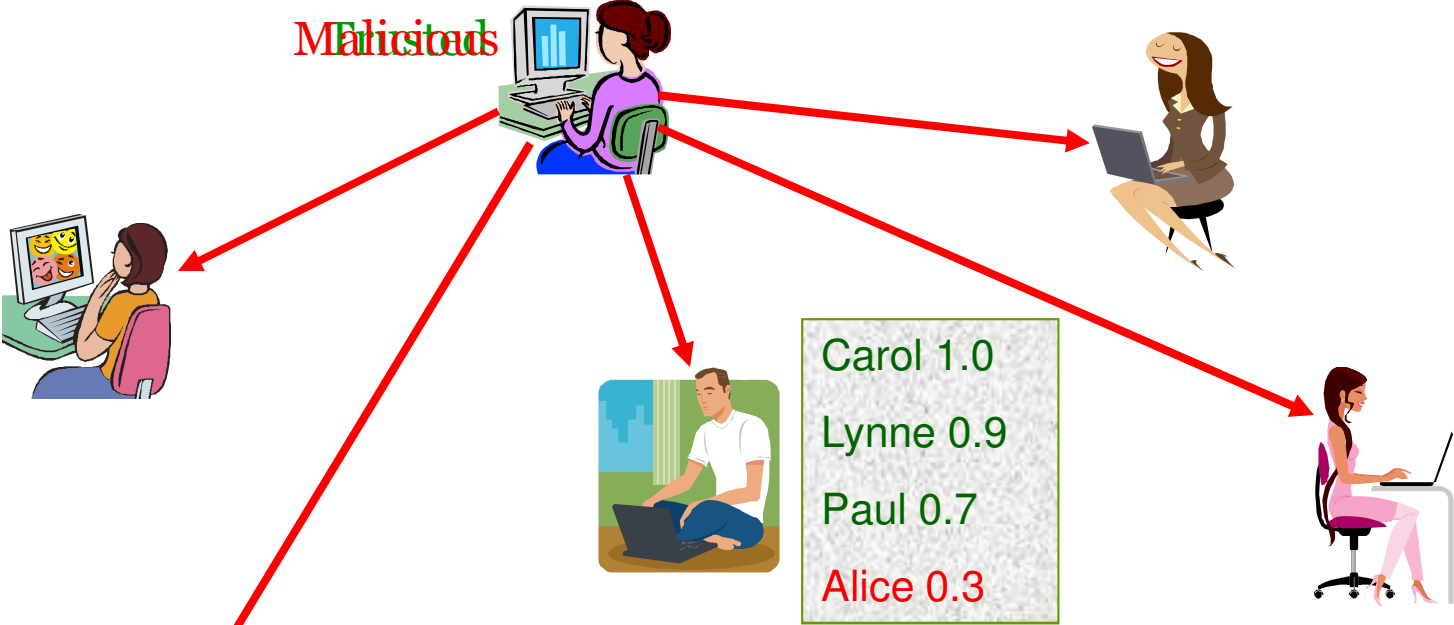


# Identity Cloning Attack





# Betrayal Attack



Trust is easy to lose and hard to gain



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# What's Next?

- Simulation design and implementation
- Design more sophisticated trust management model
  - Alert categorization
  - Expertise in intrusion detection

# Conclusion

- Proposed a trust-based IDS collaboration model
  - More accurate intrusion detection
  - Robust to several attacks
- Novel ideas
  - Use of test messages in trust establishment
  - Integration of “don’t knows” into trust value
  - Introduction of proximity
  - Aggregation threshold