

Biometric Backdoors: A Poisoning Attack Against Unsupervised Template Updating

Giulio Lovisotto, Simon Eberz and Ivan Martinovic

giulio.lovisotto@cs.ox.ac.uk

University of Oxford, UK



UNIVERSITY OF
OXFORD



Biometric Backdoor?

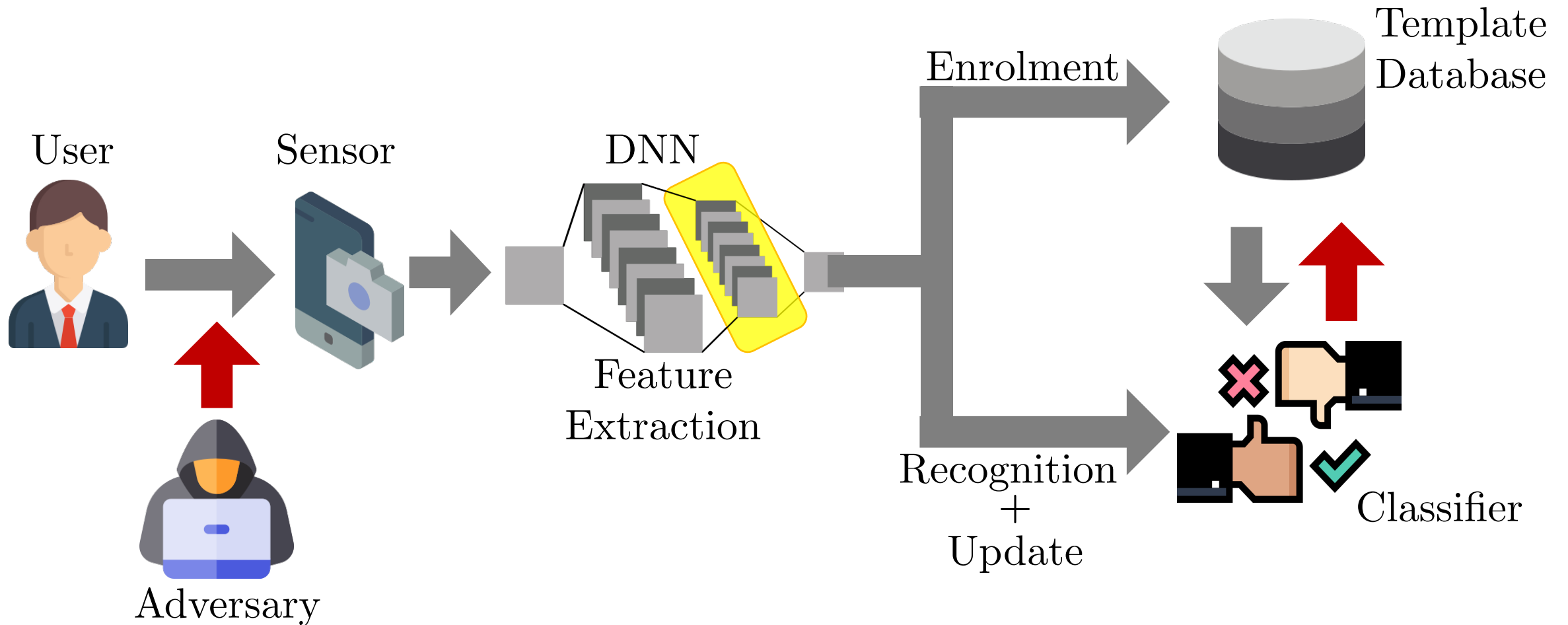
- **Accessories** for impersonation:

- Fashionable ✓
- Physically realizable ✓
- Suspicious ✗
- One-shot ✗

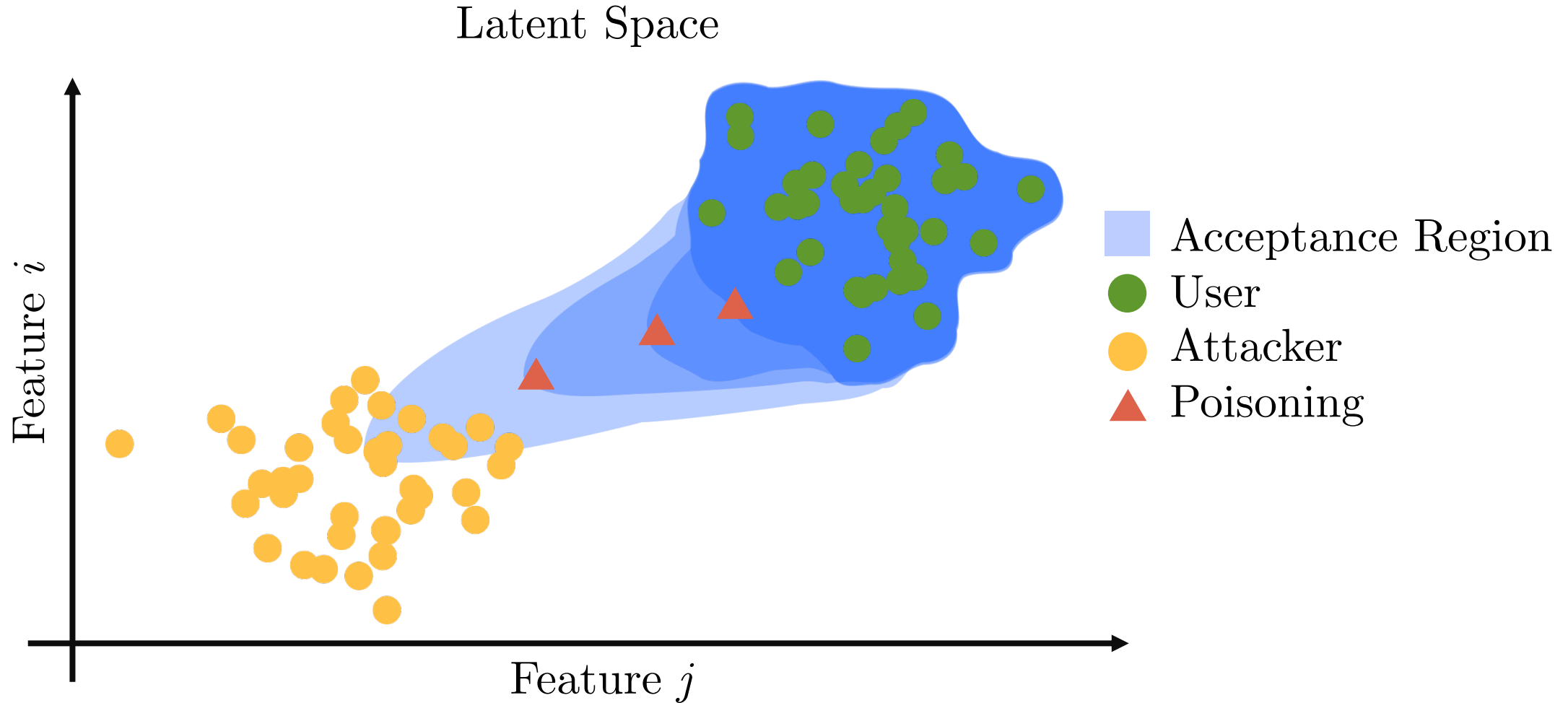


- Can we design an attack that grants (i) **long-term** and (ii) **inconspicuous** impersonation?

In Consumer Biometric Recognition



Backdoor (or Poisoning) Outline



Challenges

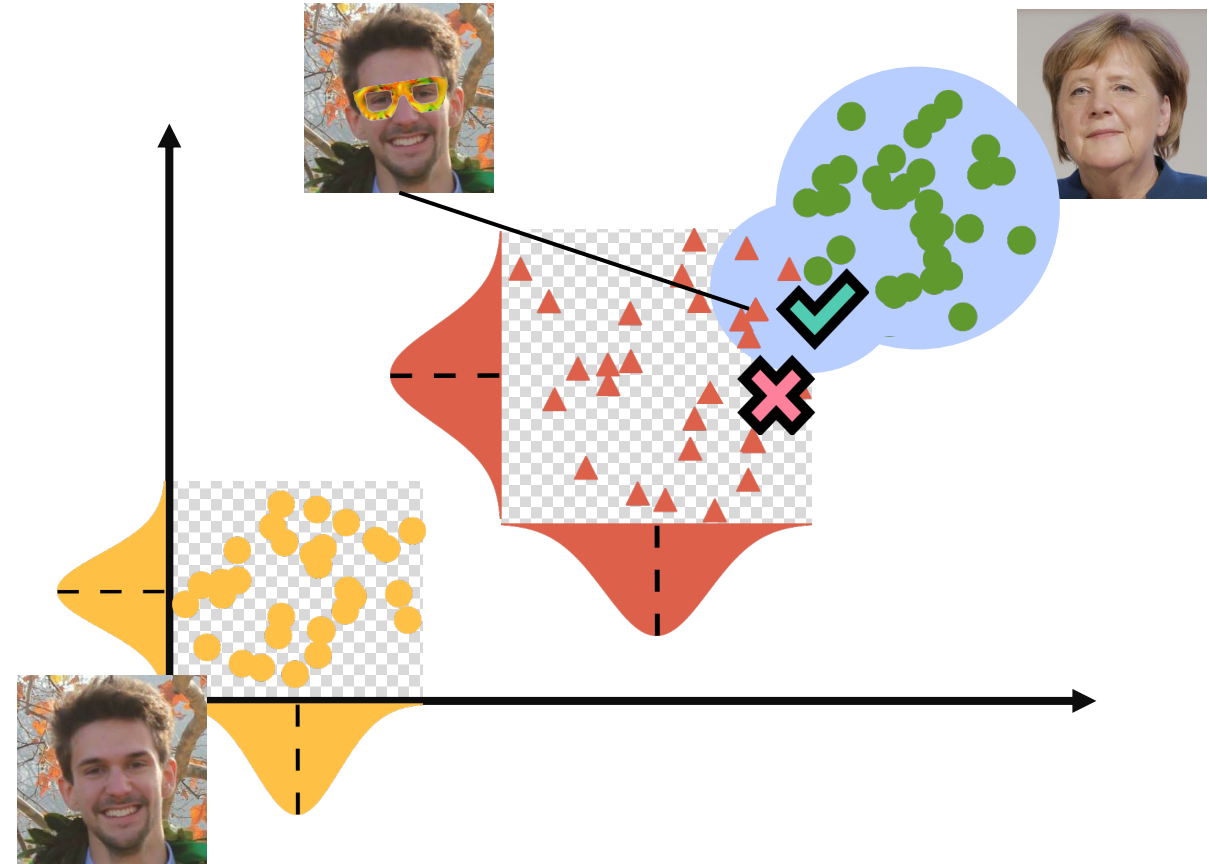
- Crafting malicious inputs



- Control input variability



- Limit # of attempts



Method (briefly)

- Optimize **starting** glasses

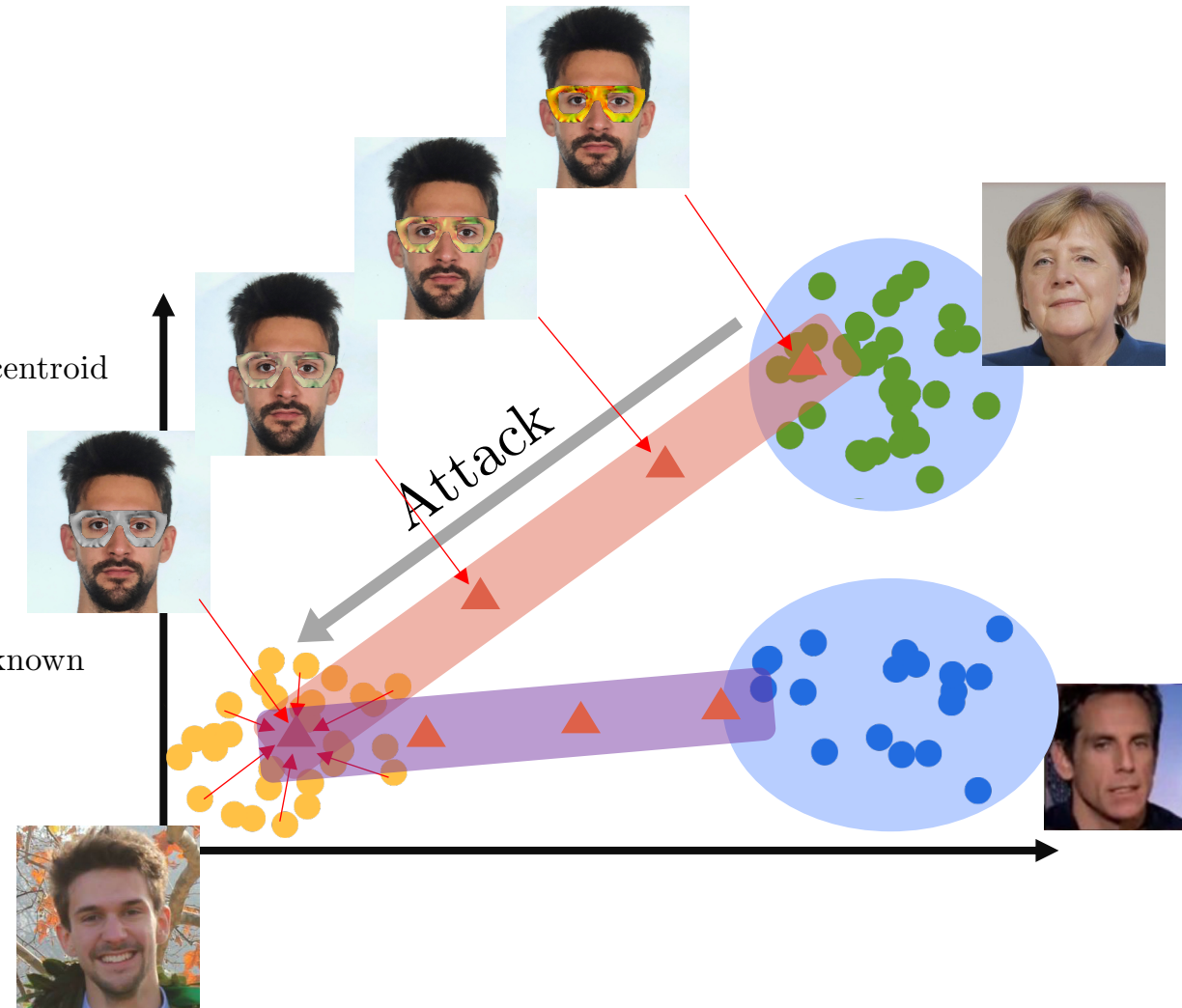
$$\min_{\text{glasses}} \text{ s.t. } \{ \text{img}_1, \text{img}_2, \text{img}_3 \} + \text{glasses} = \text{centroid}$$


- Generate **all** poisoning glasses

$$\min_{\text{glasses}} \text{ s.t. } \{ \text{img}_1, \text{img}_2, \text{img}_3 \} + \text{glasses} = \text{known}$$

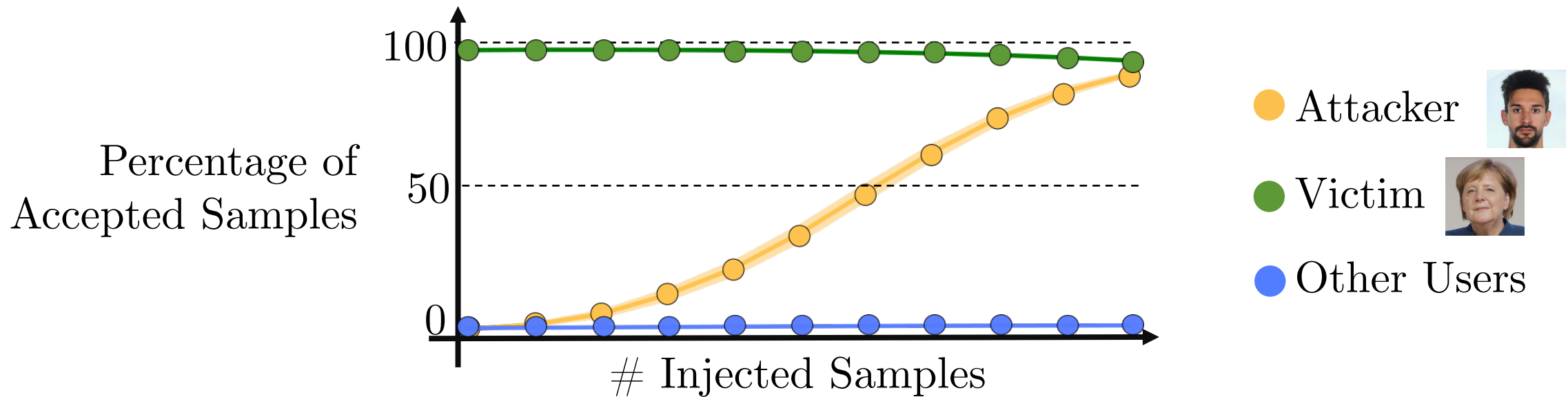

- Estimate sample to inject with **population** data

$$\text{orange bar} \approx \text{purple bar}$$

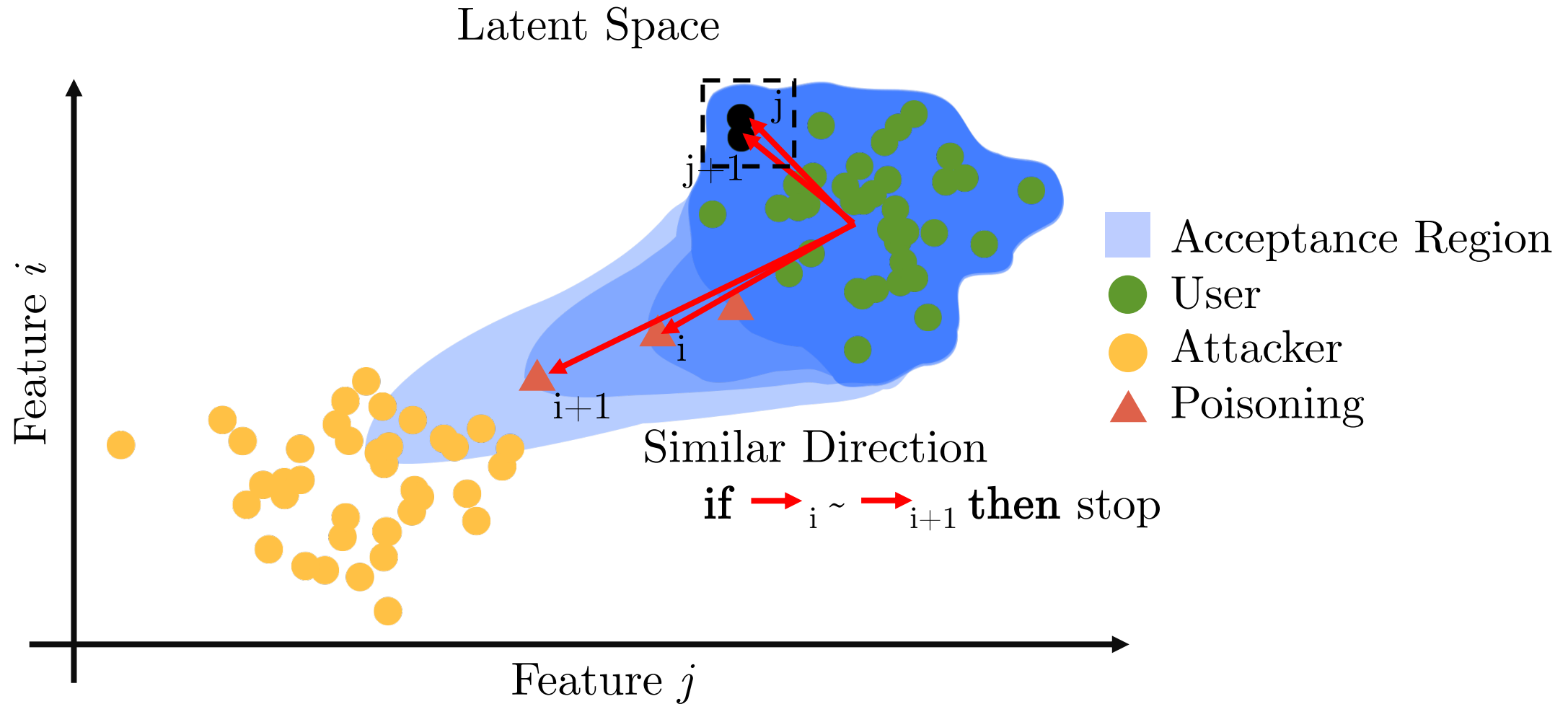


Results Takeaways

- Few injected samples suffice for the adversary to impersonate
- Victim can still authenticate with barely any performance degradation
- The system can still reject non-legitimate “other” users

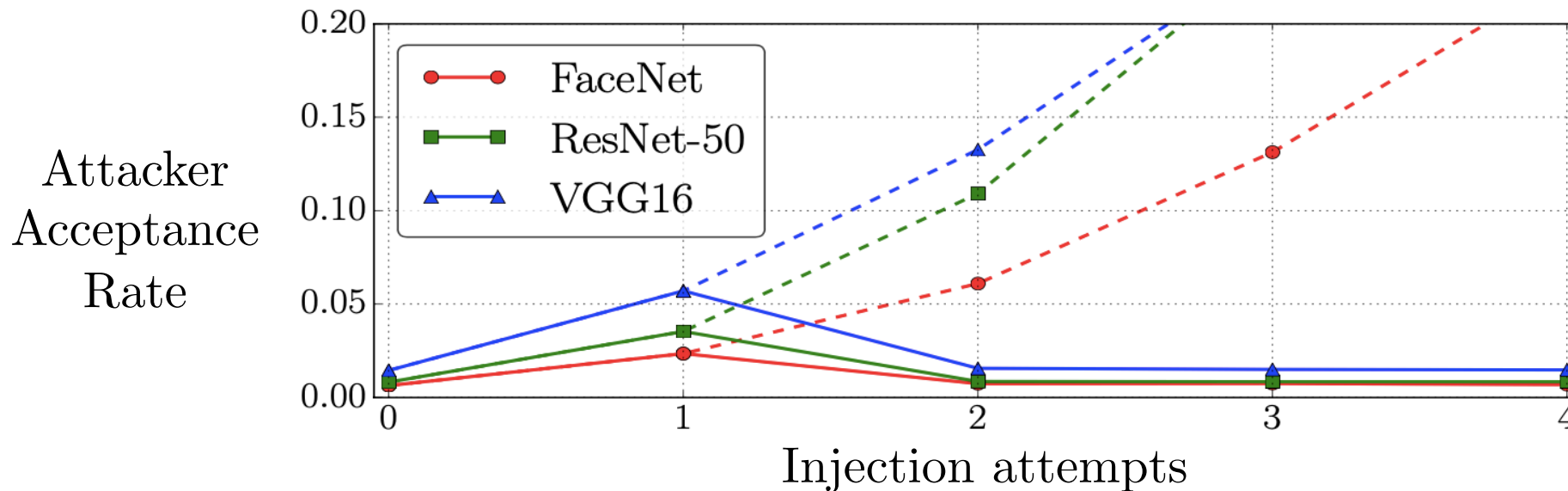


Countermeasure



Countermeasure

- **Insight:** intra-user variation factors may generate consistent directional updates
- **Evaluation:**
 1. Select legitimate sequences of updates which generate directional shifts
 2. Test the detection with a binary threshold on the cosine similarity
- **Result:** 93% detection rate (@EER) on whether a pair of updates is malicious.



Conclusion

- Introduced a **backdoor attack** by exploiting the unsupervised **template update** procedure:
 - The attack copes with *limited knowledge* about the victim and *limited capabilities* of injection
 - A successful attack leads to *inconspicuous* and *long-term* impersonation
 - Some classifiers are particularly vulnerable, with only *one* injected sample sufficient to allow impersonation.
- We proposed a **countermeasure** and we evaluated its detection trade-offs with legitimate template updates:
 - Our countermeasure can detect poisoning samples *93%* of times