

# SocialHEISTing: Understanding Stolen Facebook Accounts

---

**Jeremiah Onaolapo**  
*University of Vermont*

Nektarios Leontiadis  
*Facebook*

Despoina Magka  
*Facebook*

Gianluca Stringhini  
*Boston University*

# Social Accounts

- Often publicly display demographic attributes (age, gender, location, etc.)
- Interesting contents in social accounts!
- Accumulate personal info + sentimental value over time
- Attributes can be abused by malicious parties

# Goal

- Understand the effects of demographic attributes on attacker behavior in stolen social accounts
- Without harming any real users
- Distinct from general characterization of attacker behavior

# How?

---

# Pipeline

Create + populate honey accounts

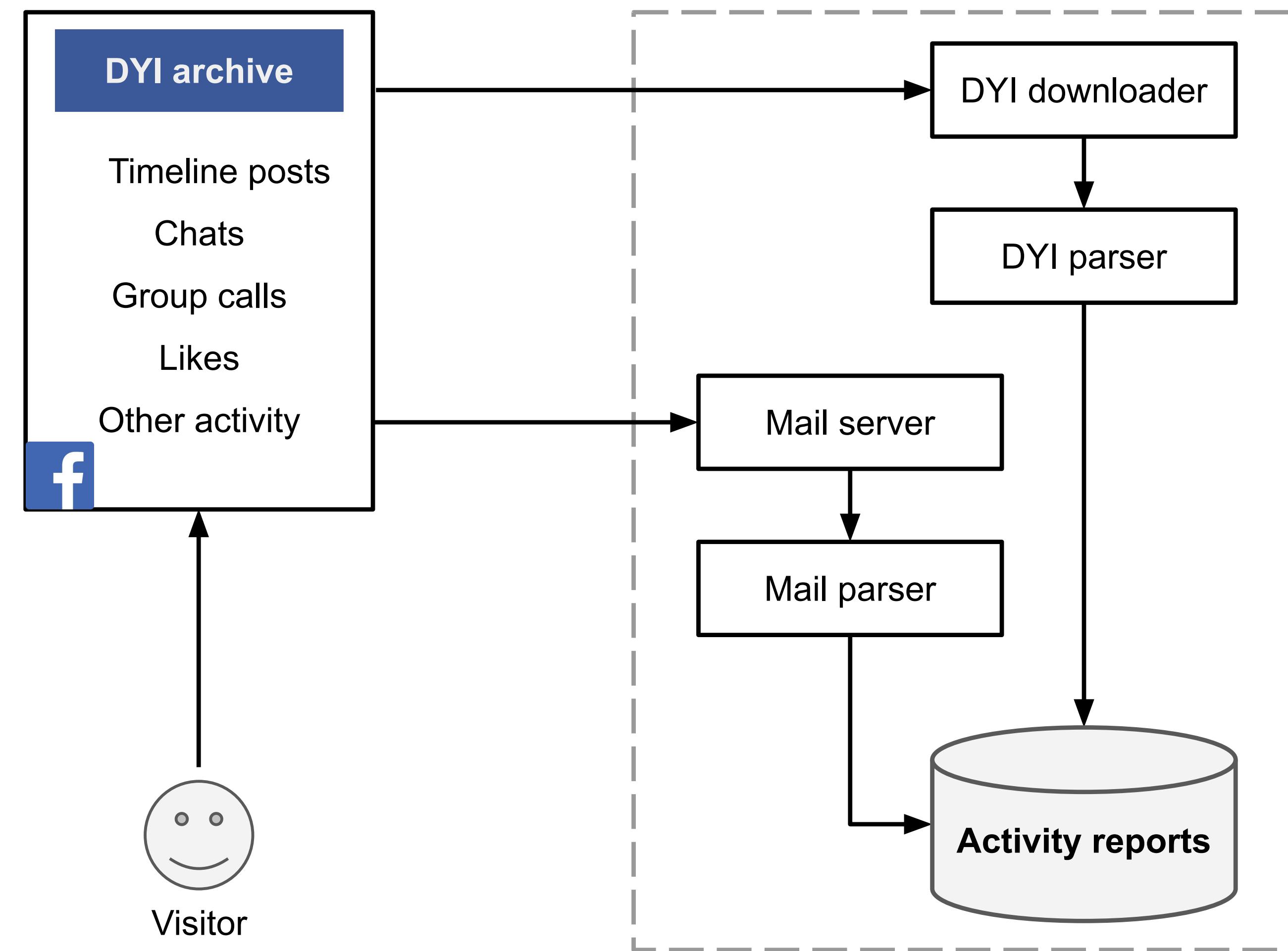
Configure monitor infrastructure

Leak honey credentials

Record + analyze data

Better understand  
cybercriminal behavior

# Data Collection



# Setup

- 1008 realistic Facebook accounts (*age, gender vars*)
- Populated with publicly available data (sanitized)
- Leaked credentials to two-thirds of the accounts
- Via paste sites on Surface Web + Dark Web
- Monitored accounts for 6 months

# Results

---

# Actions

- 322 unique accesses to
- 284 accounts, resulting in
- 1,159 actions
- *Curious, Searcher, and Chatty* activity tops the actions table

# Age of Account

Criminals...

- **Add/remove friends:** adult accounts > teen accounts
- **Edit profiles:** adult accounts < teen accounts
- **Create posts, chat:** adult accounts < teen accounts

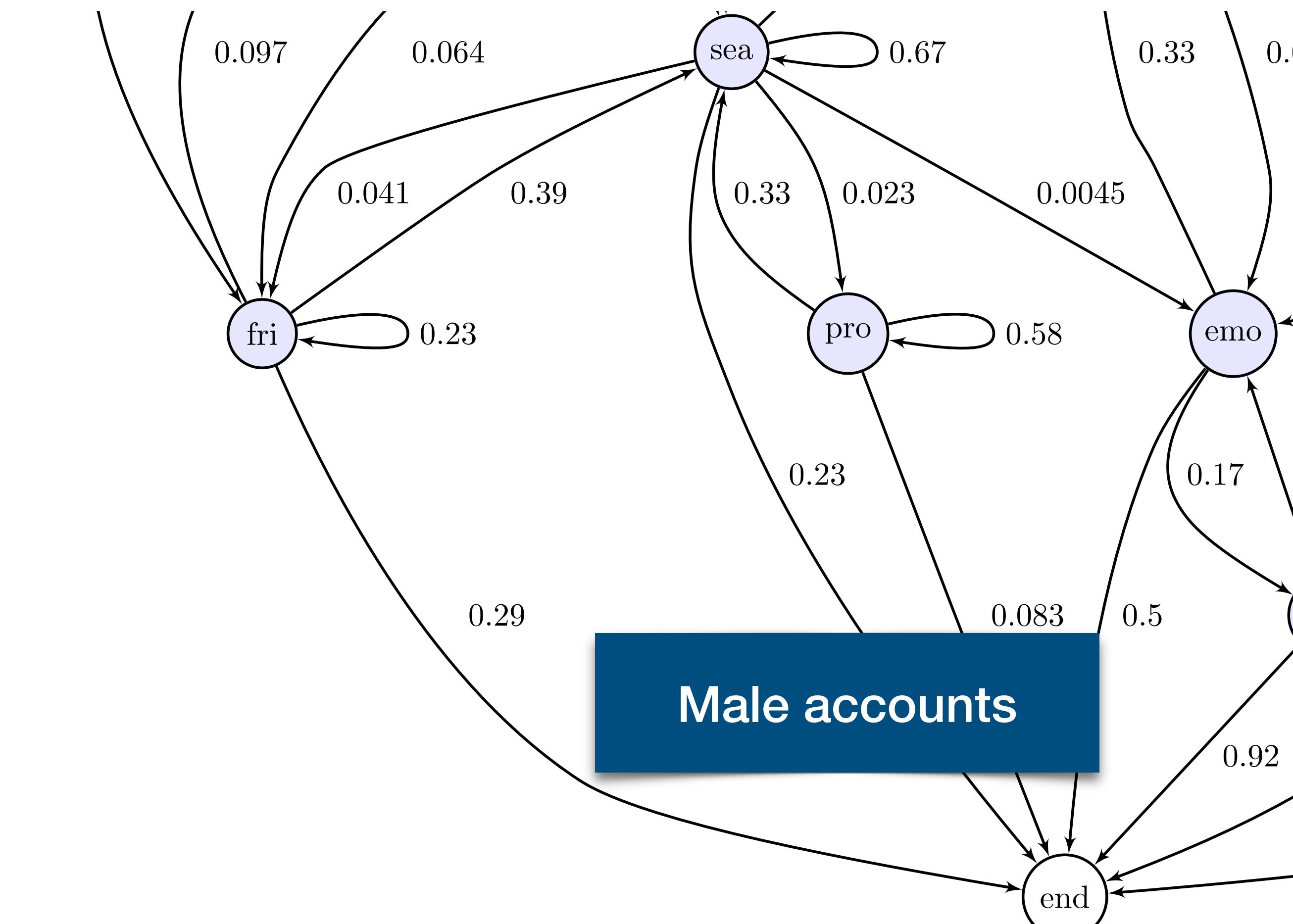
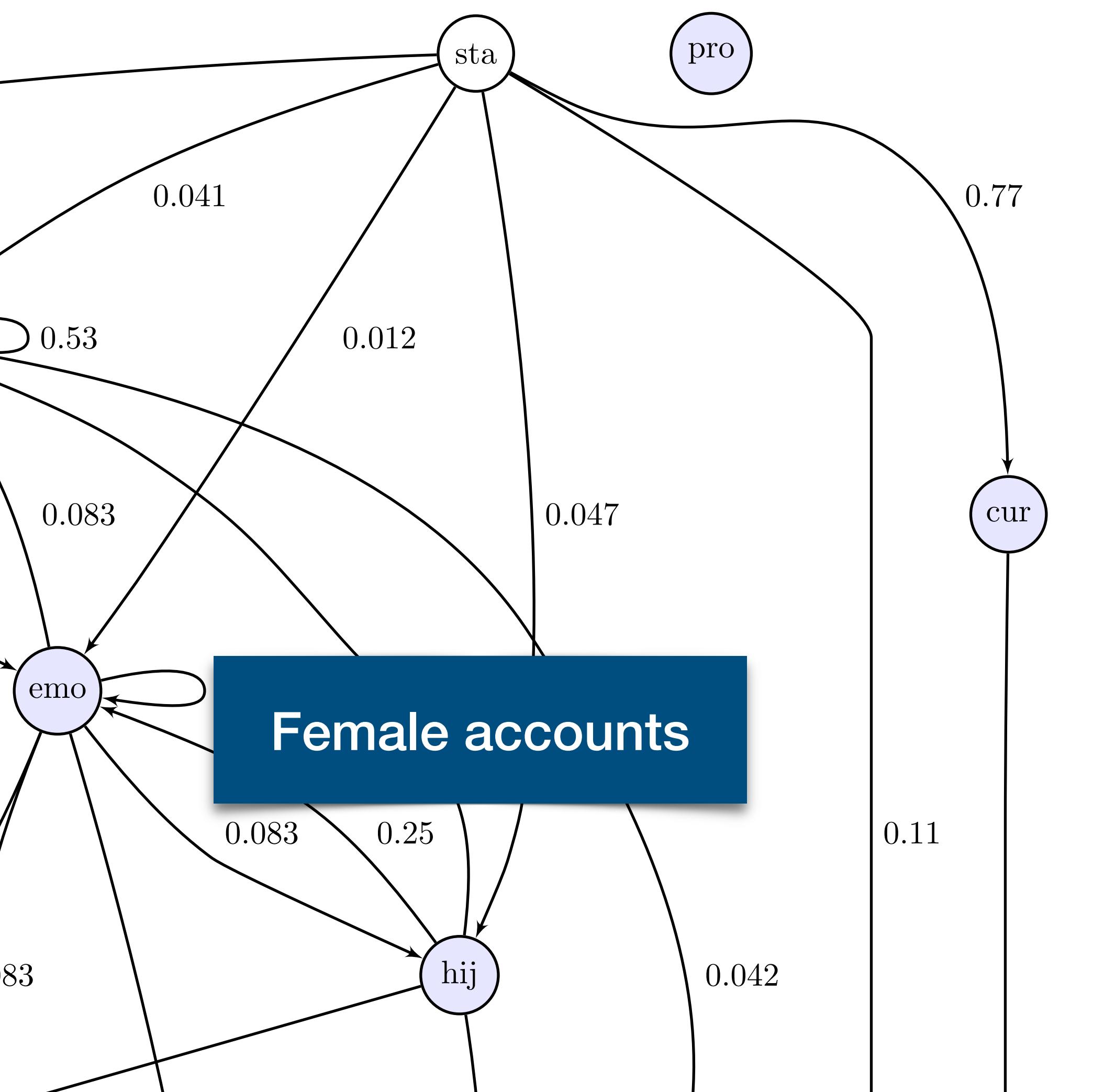
# Gender of Account

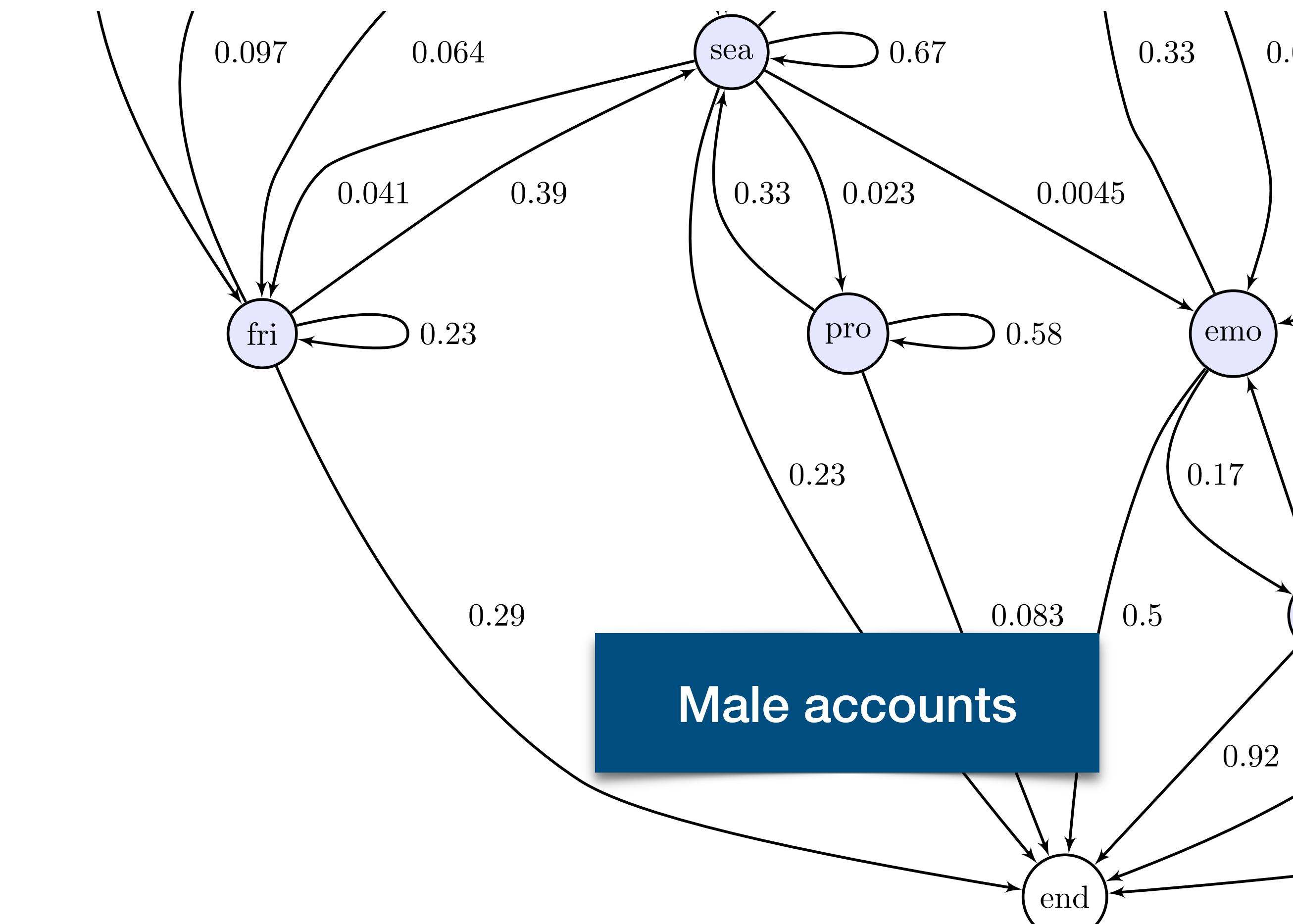
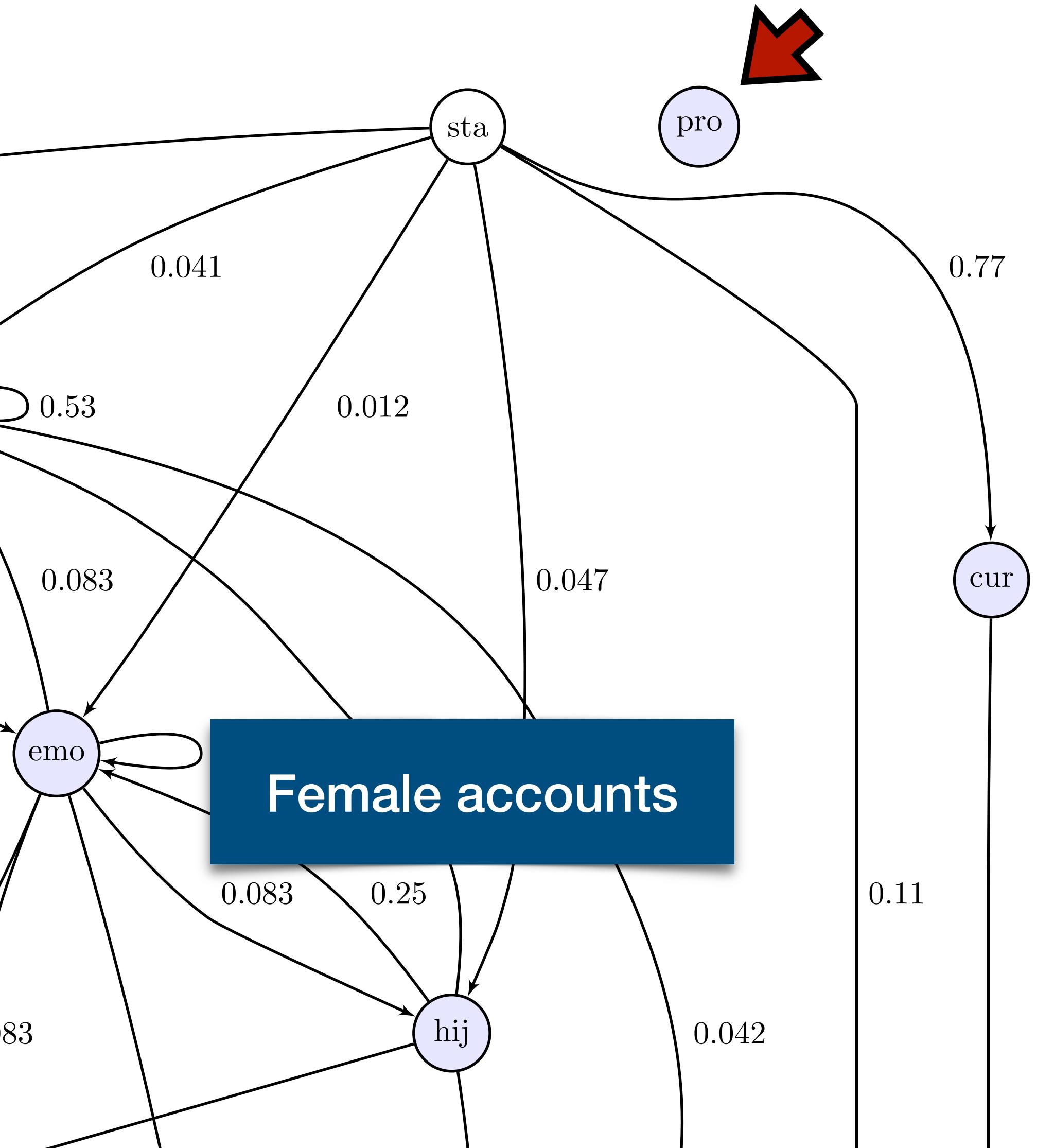
Criminals...

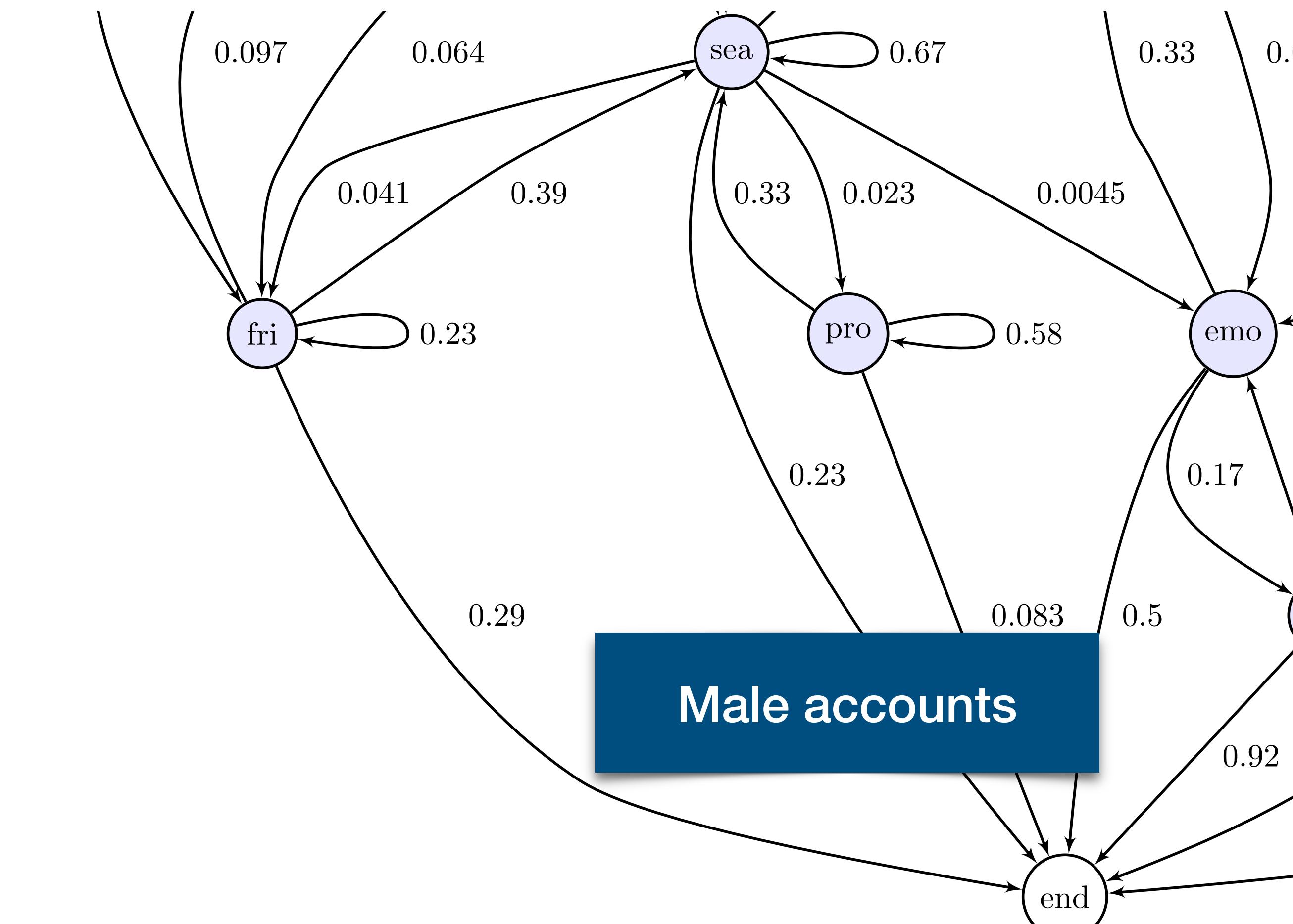
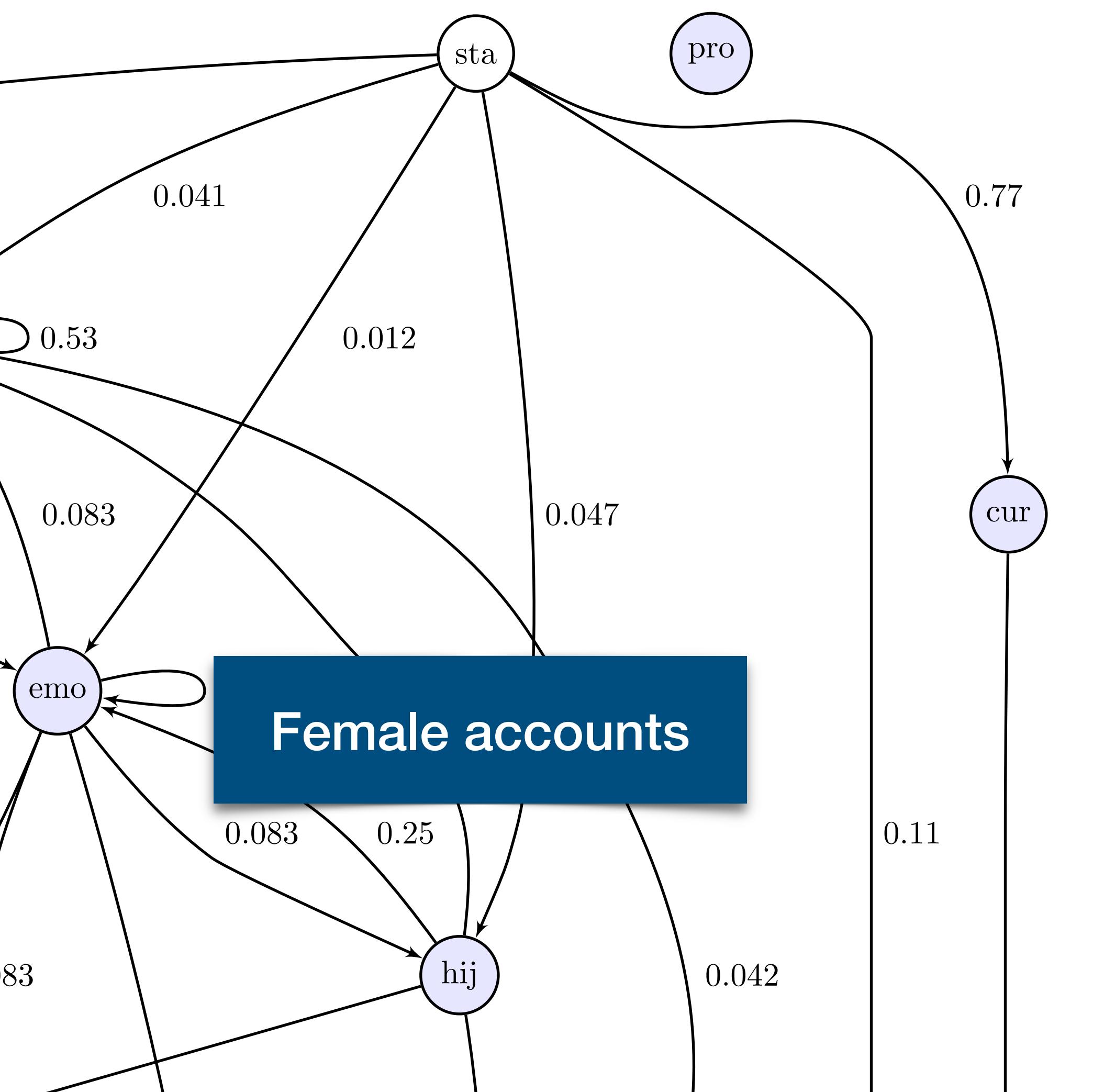
- **Add/remove friends:** female accounts > male accounts
- **Edit profiles:** female accounts (none) < male accounts
- **Search:** female accounts < male accounts

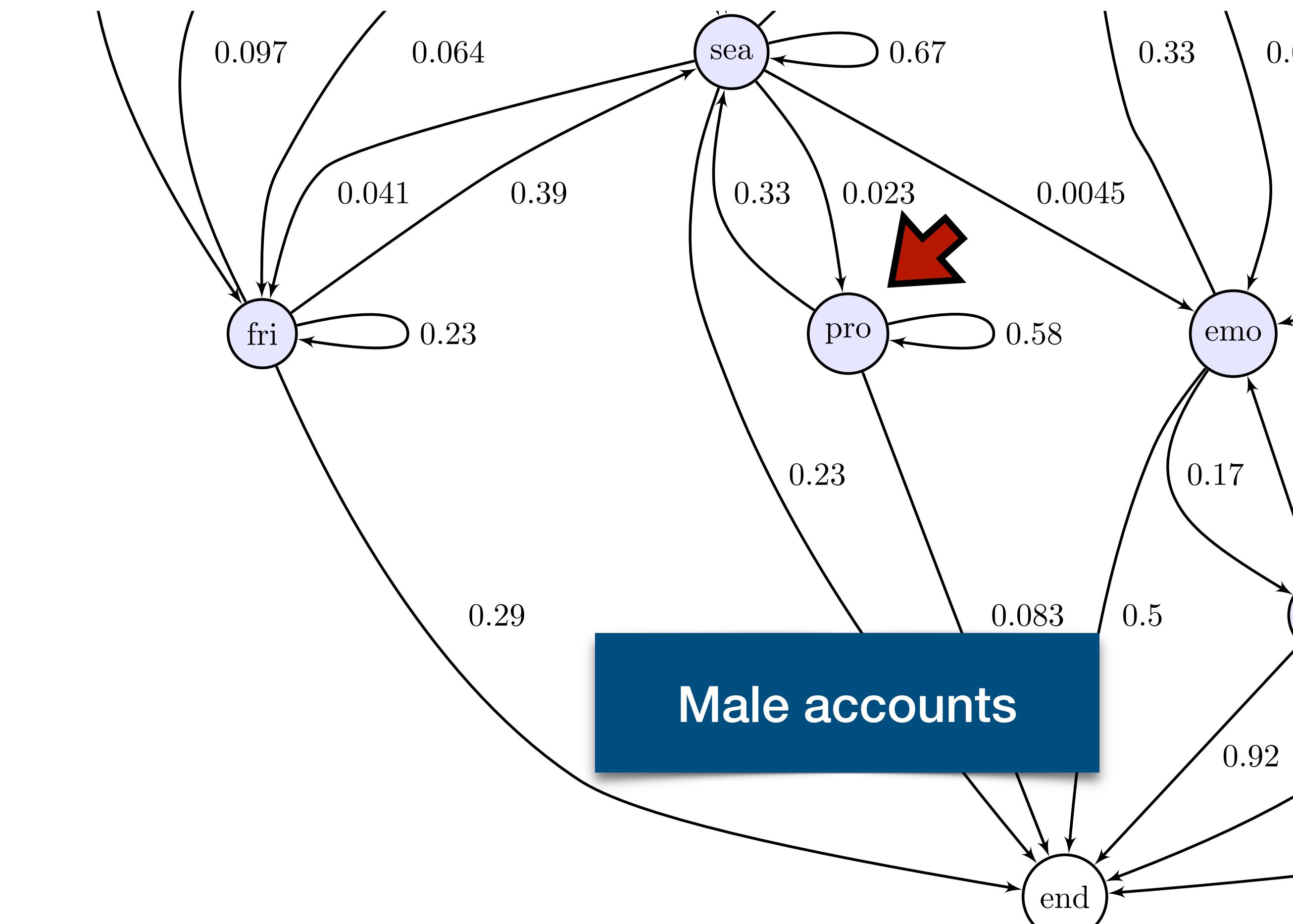
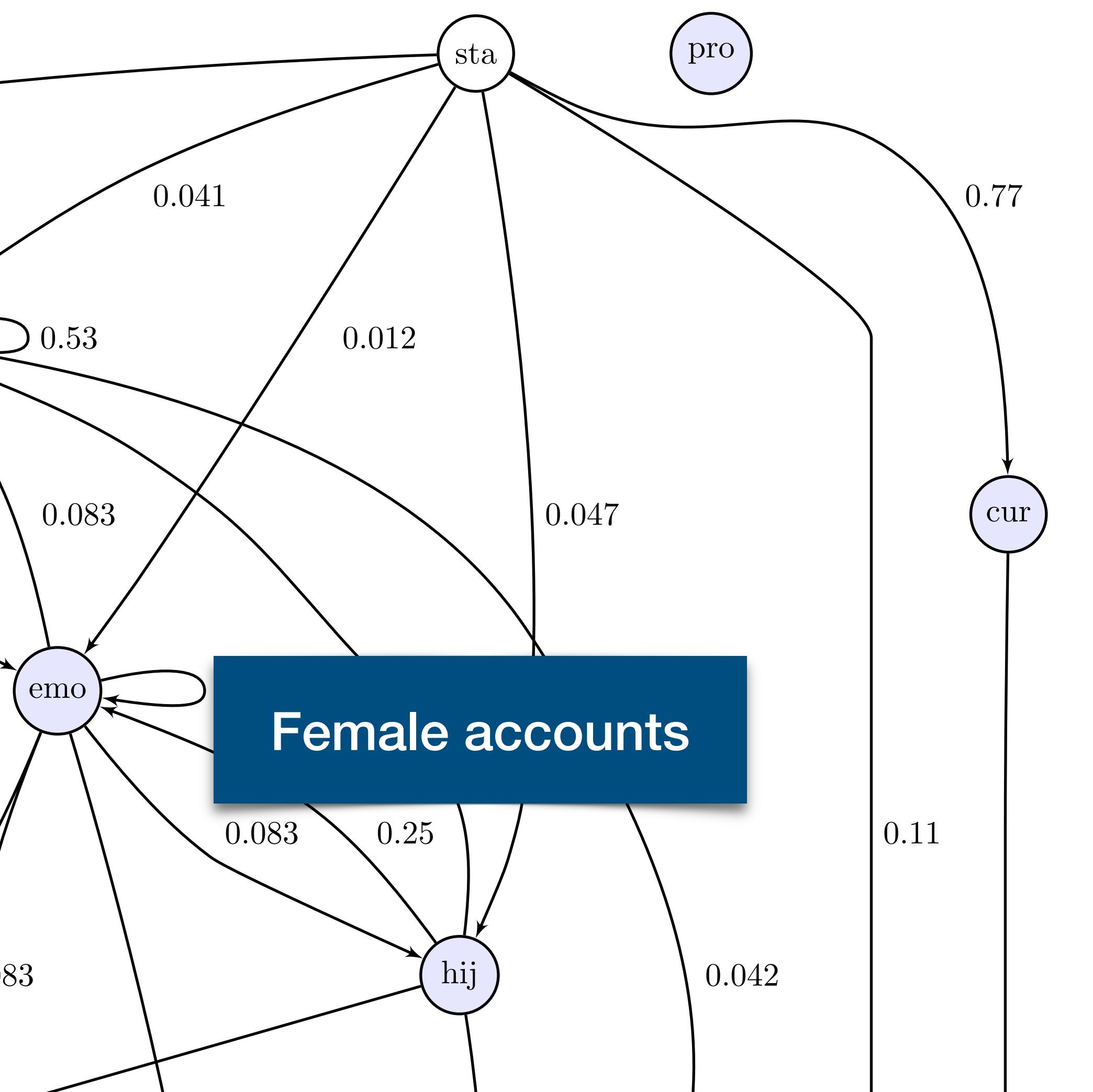
# Action Sequences

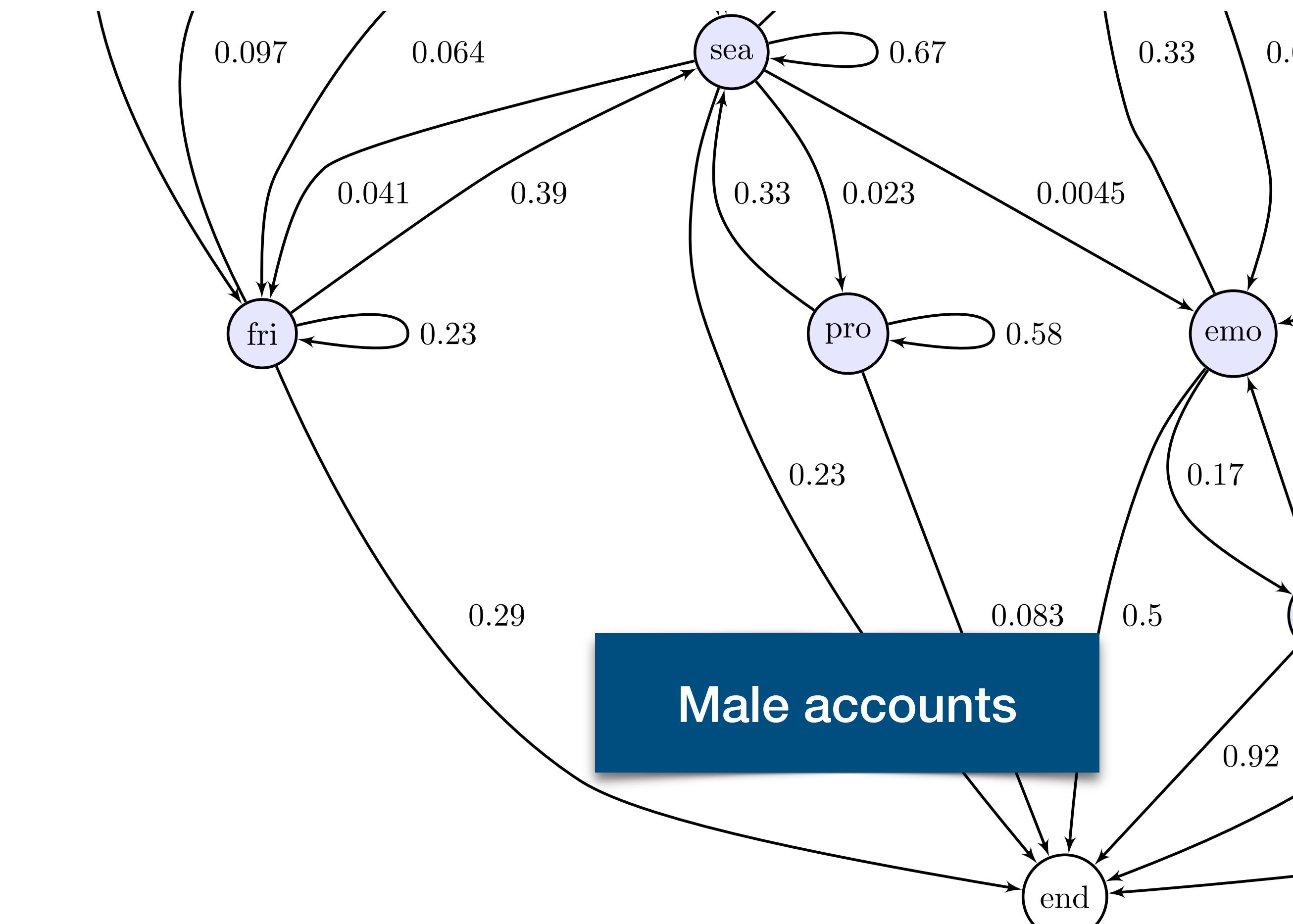
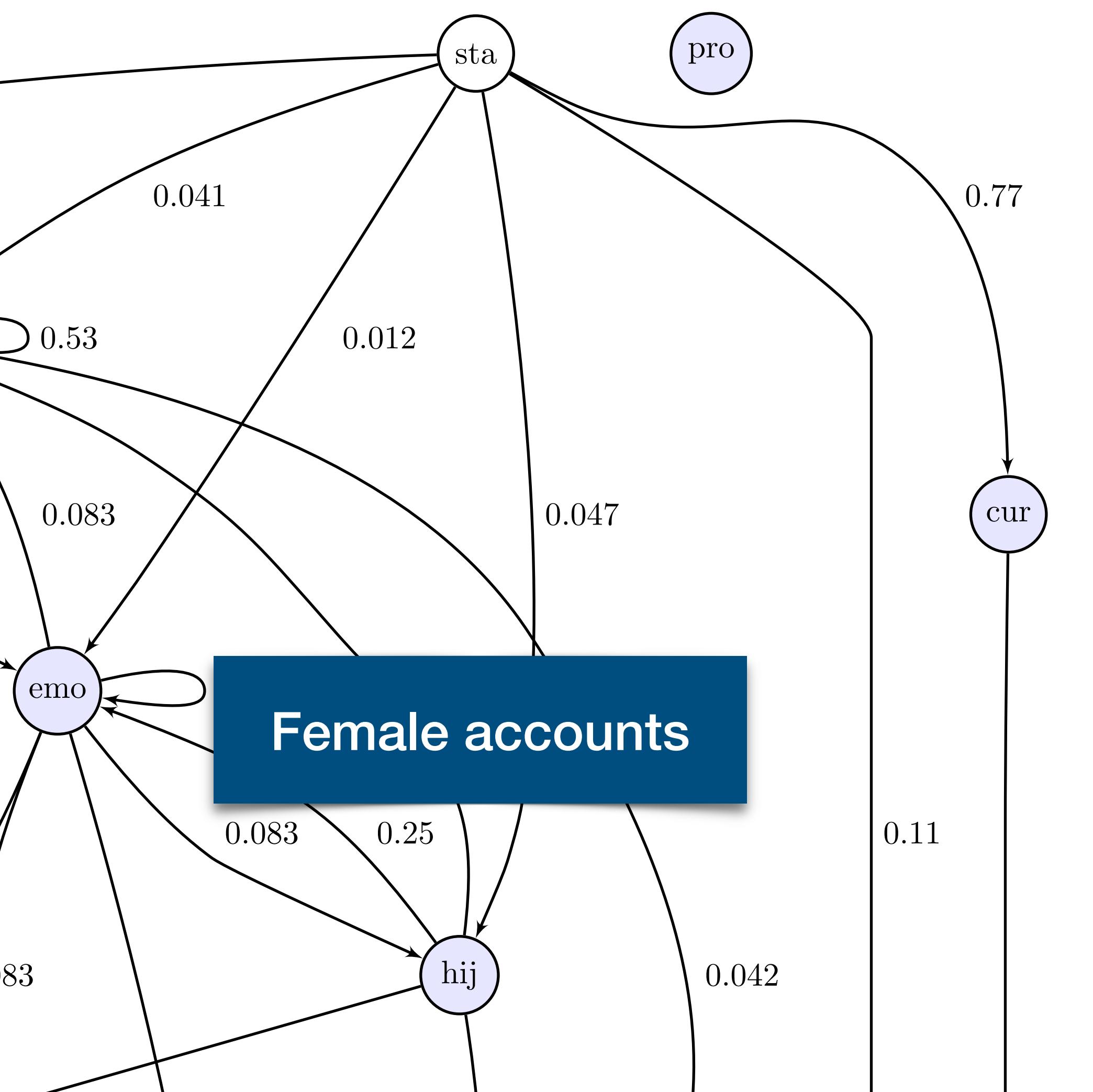
- Modeled action sequences as graphs; edge weights as probabilities of transitions
- Transitions from *action* to *other action* differed across the age and gender dimensions of victim accounts
- Illustrative example: *emo* → *cha* → *hij*









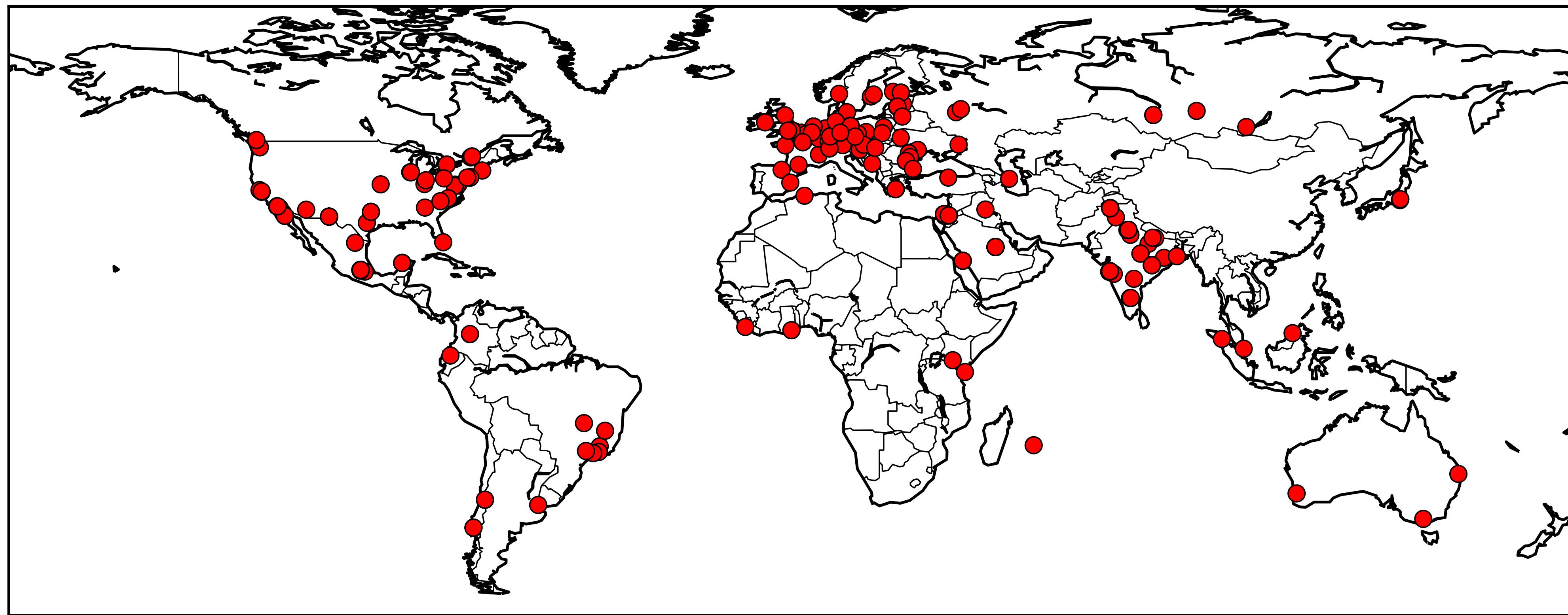


# Origins of Accesses

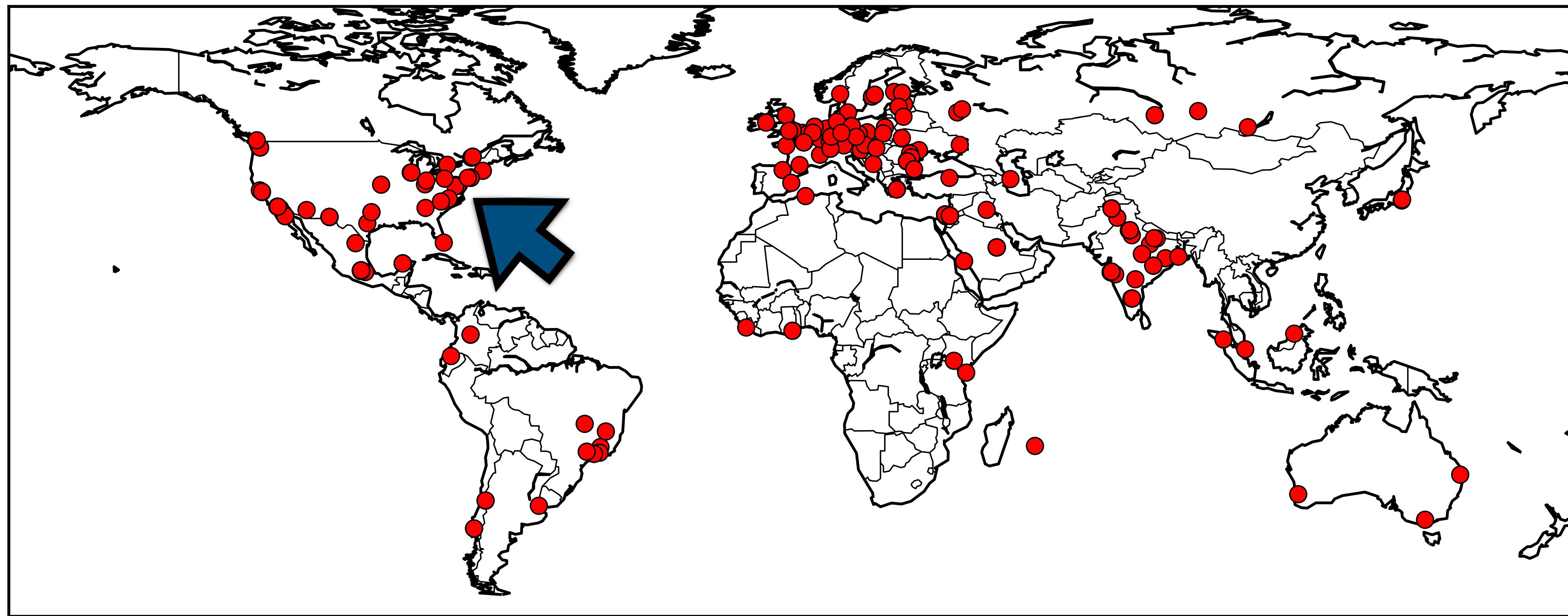
- 415 IP addresses (a mix of IPv4 and IPv6)
- 53 countries
- 39 TOR exit nodes

*Caveat:* Some may be VPNs and proxies

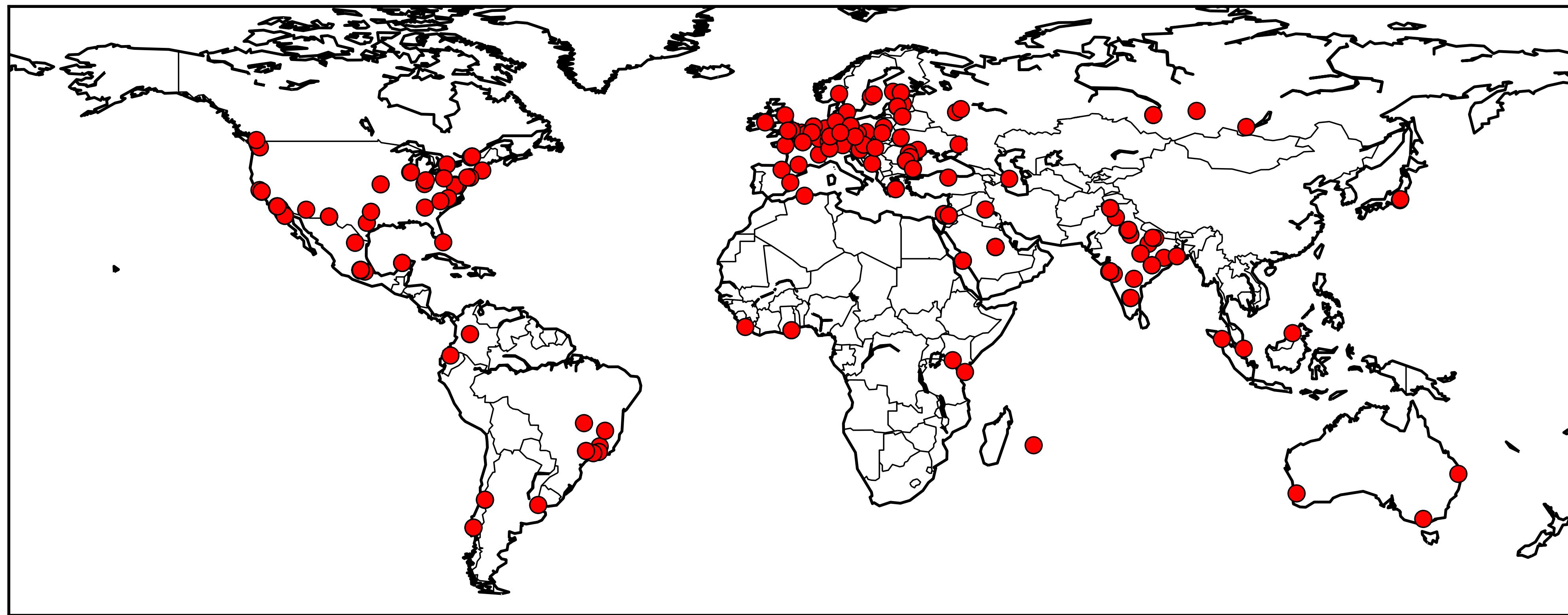
# Origins of Accesses



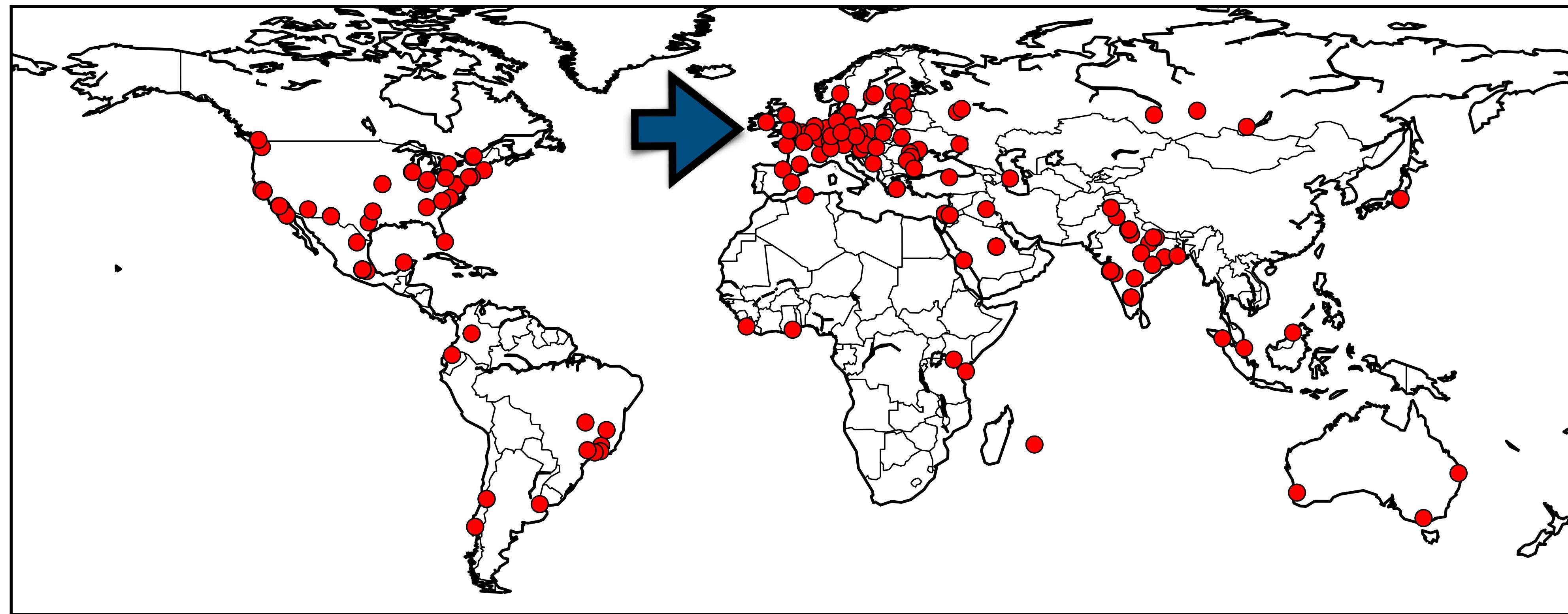
# Origins of Accesses



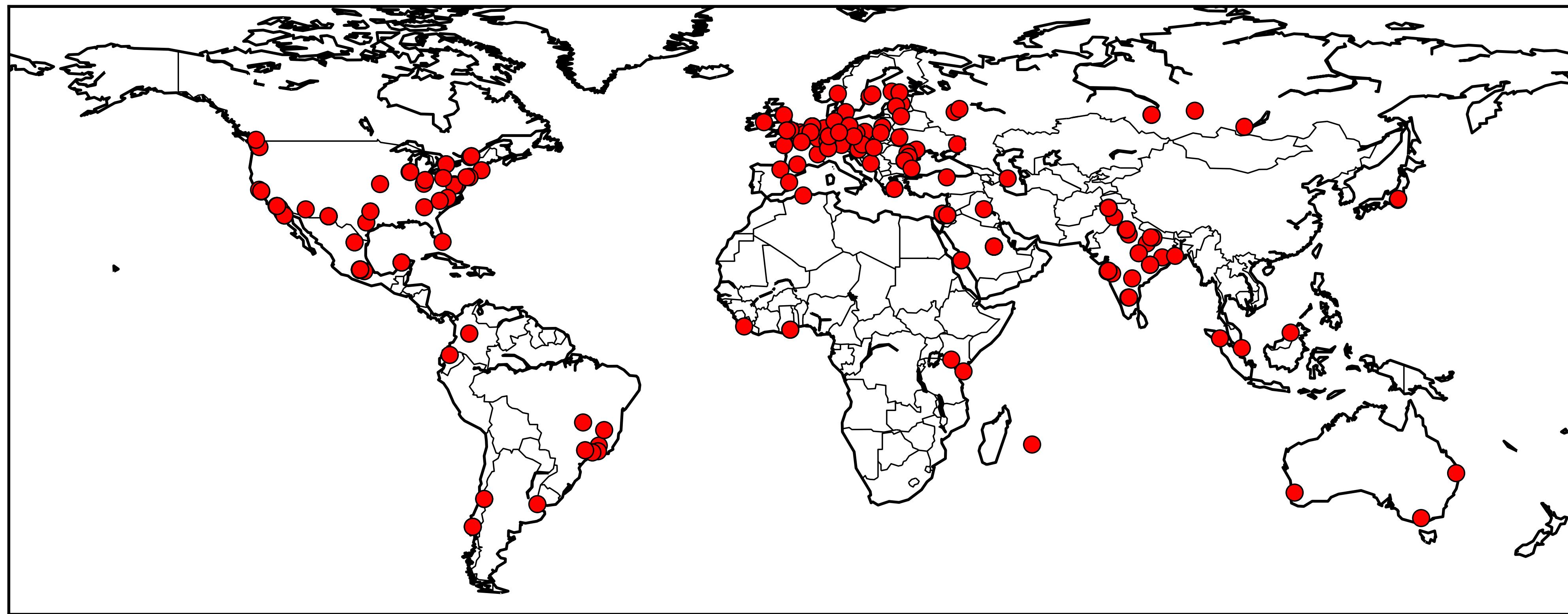
# Origins of Accesses



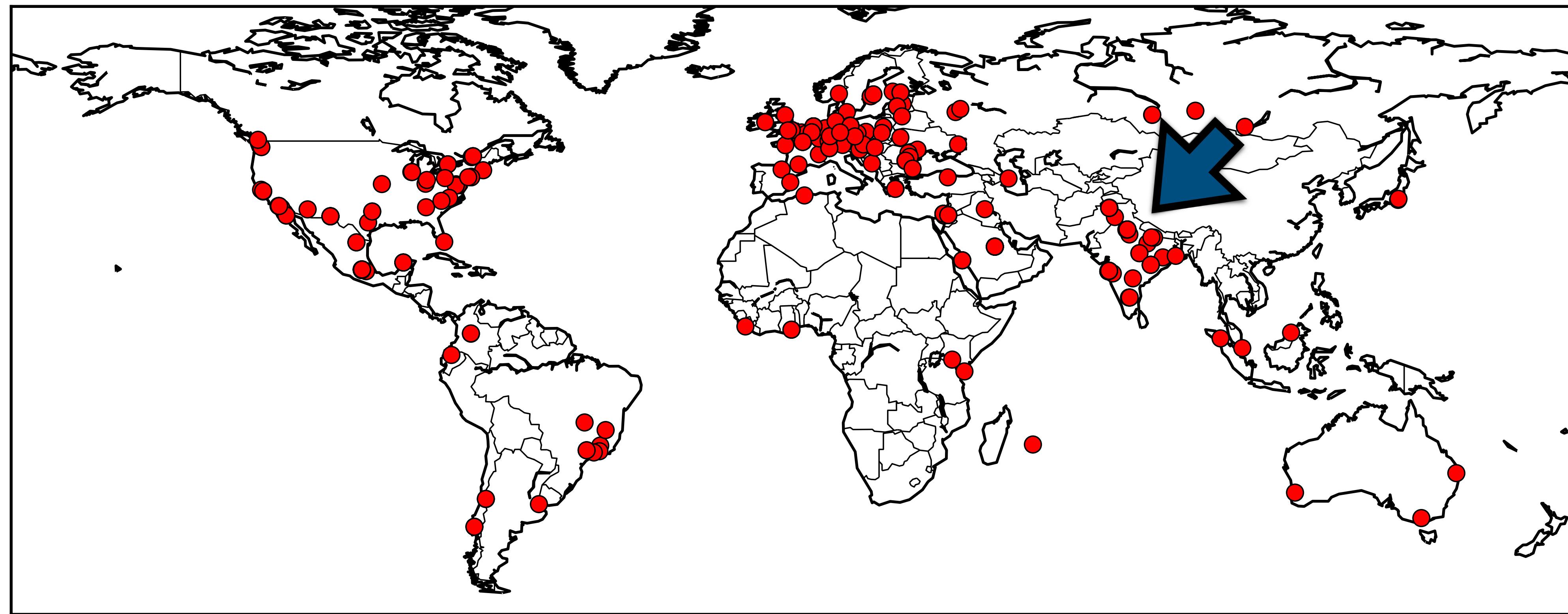
# Origins of Accesses



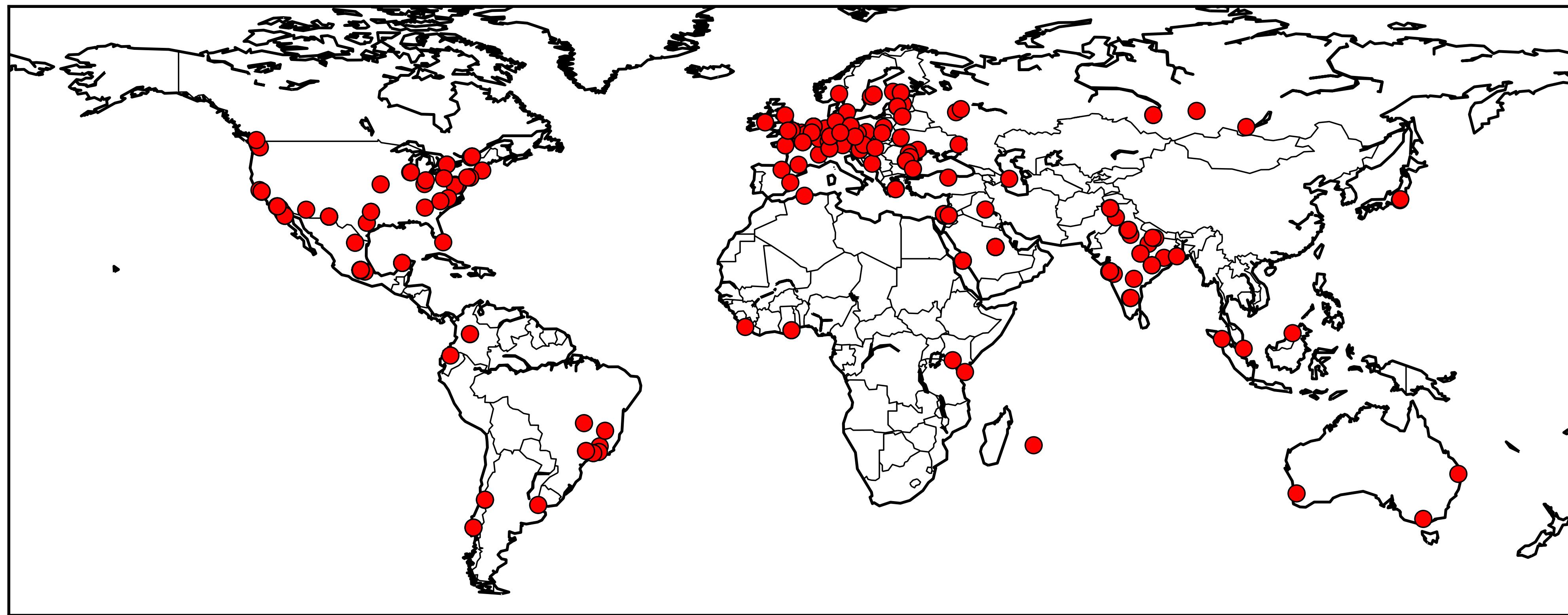
# Origins of Accesses



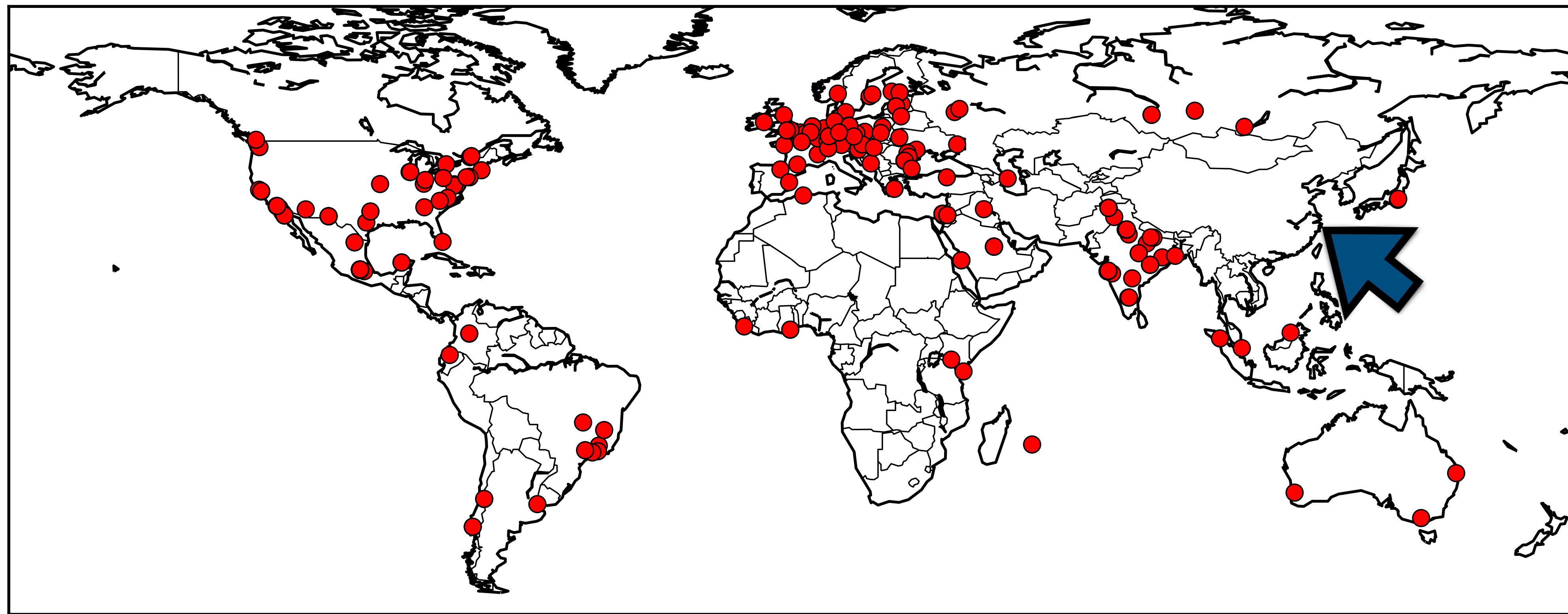
# Origins of Accesses



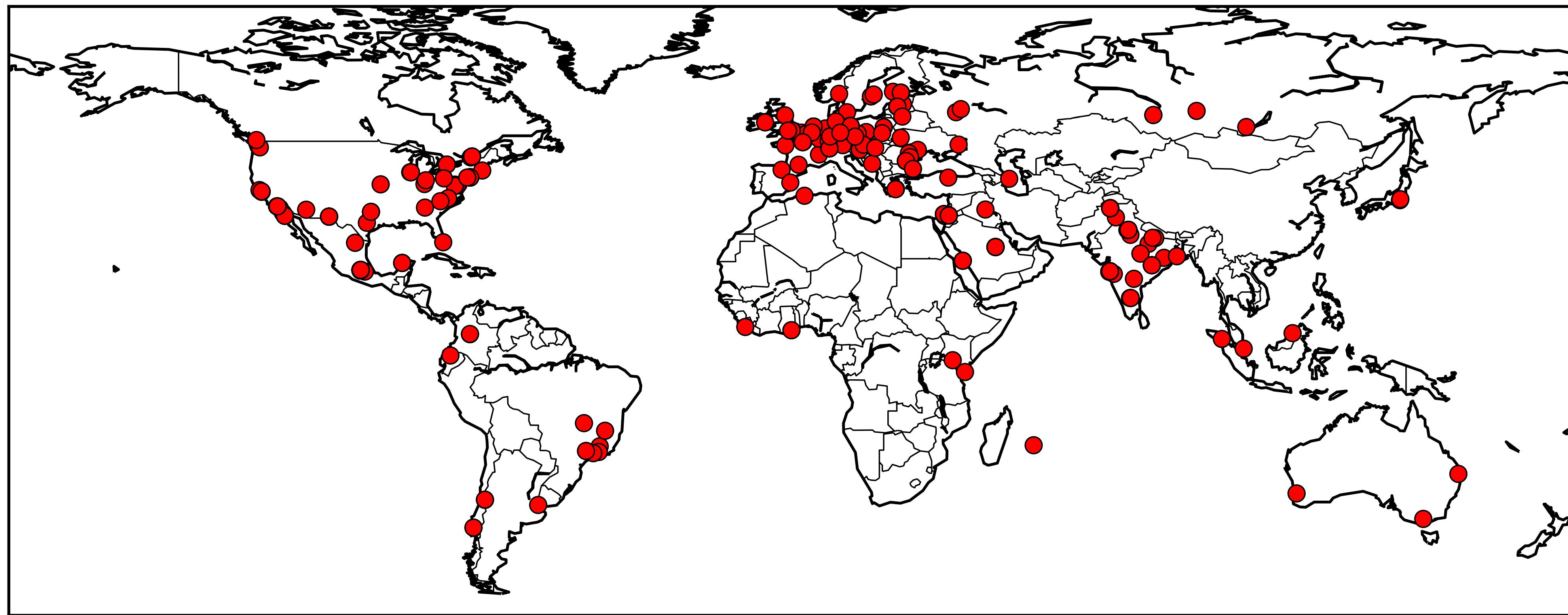
# Origins of Accesses



# Origins of Accesses



# Origins of Accesses



# Summing Up

---

# Implications

- Need to rethink detection and mitigation systems
- Along demographic (and other?) attributes
- More work needs to be done in this area

# Ethics

- Used test accounts; isolated from regular Facebook social graph
- Used publicly available stock photos and social posts
- Facebook contacts kept an eye on the accounts
- Obtained IRB ethics approval

# Thanks!

- Jeremiah Onaolapo, *University of Vermont*
  - ✉ [jeremiah.onaolapo { at } uvm.edu](mailto:jeremiah.onaolapo@uvm.edu)
  - 🏠 [www.uvm.edu/~jonaolap](http://www.uvm.edu/~jonaolap)
- Nektarios Leontiadis, *Facebook*
- Despoina Magka, *Facebook*
- Gianluca Stringhini, *Boston University*