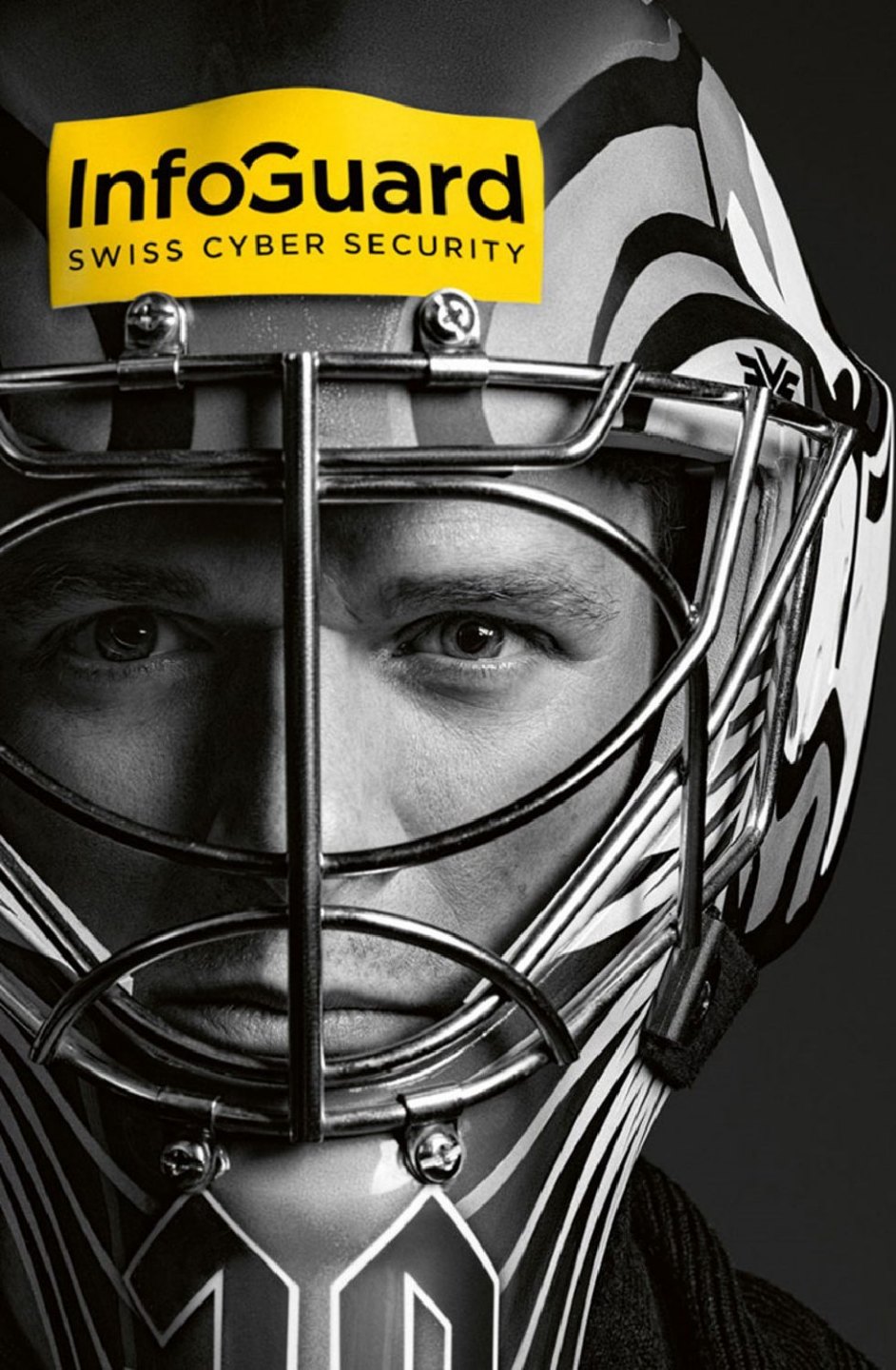




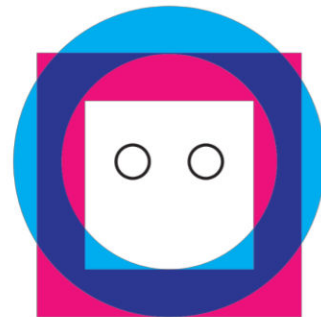
**„TELL ME WHAT YOU USE
AND I WILL TELL YOU WHO
YOU ARE“ (Ch.II)**

Swiss Cyber Storm 2023



INTRODUCTION

PROFILE

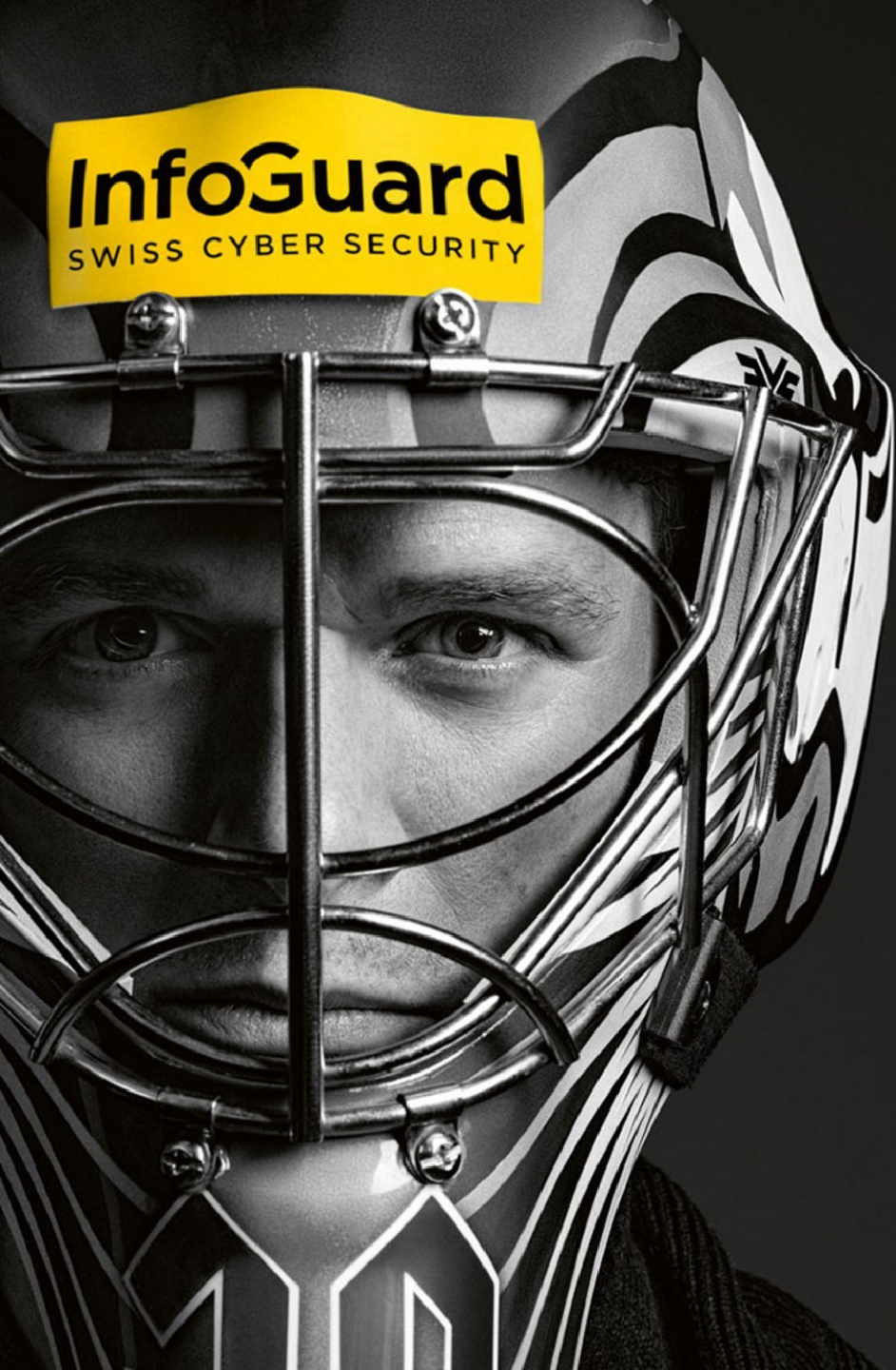


HUMAN-IST
HUMAN CENTERED
INTERACTION
SCIENCE
& TECHNOLOGY

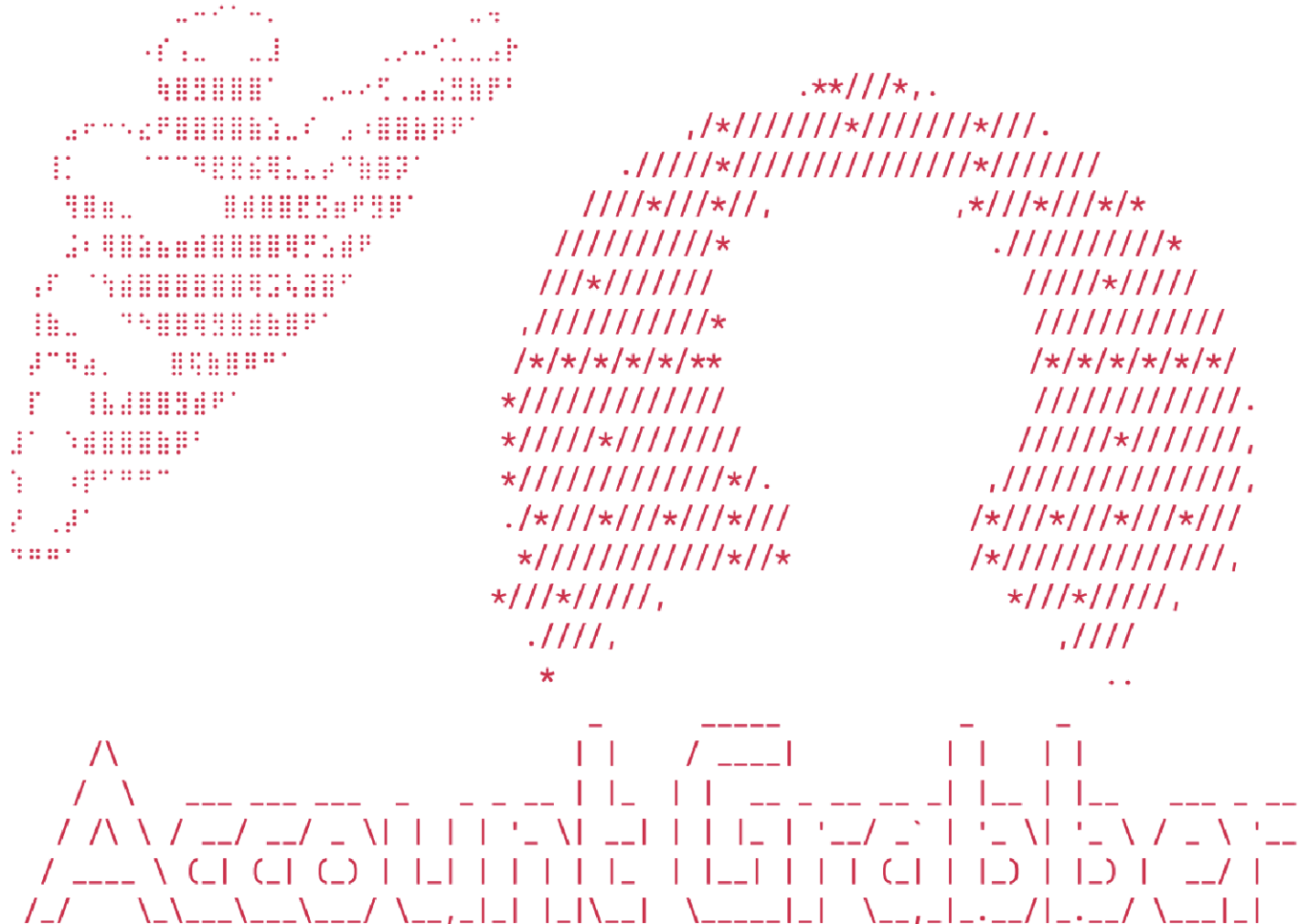
- Penetration Tester@Infoguard AG
- External PhD Student@Human-IST Institute, University of Fribourg
- Applied science research collaboration

WHAT IS THIS TALK ABOUT?

- interdisciplinary security research
- work in progress
- no final conclusions (yet)
- interesting observations, hypotheses
- outlook



TOOLING



WHAT IS THE TOOL?

- Open source intelligence (OSINT) project
- Does user enumeration on webapps
- Analyzed 300-400 webapps
- Tool currently supports >110
- Enhancements w/ IPA

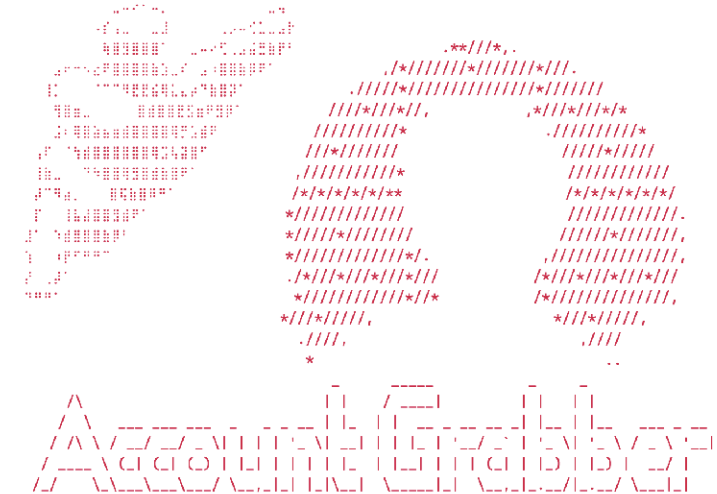


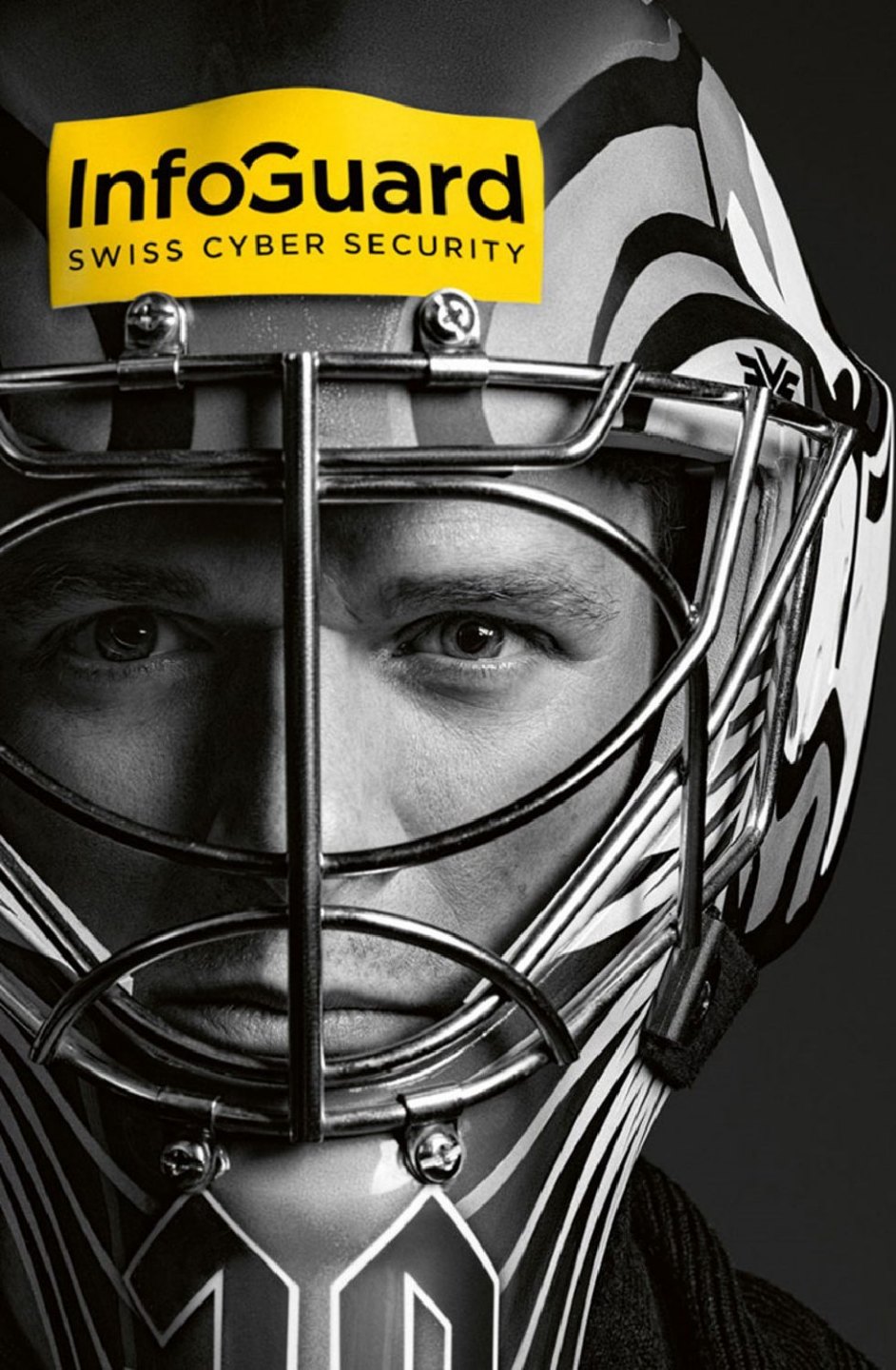
WHAT DOES THE OUTPUT LOOK LIKE?

e-mail	first name	last name	gender	byear									
			male	1973	1	0	0	0	0	0	0	0	0
			female	1962	1	1	1	1	1	0	0	0	0
			male	1975	1	0	0	0	0	0	0	0	0
			female	1971	1	0	1	1	0	0	0	0	0
			male	1961	0	0	0	0	0	0	0	0	0
			female	1985	1	0	0	0	0	0	0	0	0
			female	1972	0	0	0	0	0	0	0	0	0
			female	1979	1	0	0	0	0	0	0	0	0
			male	1955	0	0	0	0	0	0	0	0	0
			female	1972	0	0	0	0	0	0	0	0	0
			male	1962	1	0	0	0	0	0	0	0	0
			female	1984	1	0	1	0	0	0	0	0	0
			female	1991	0	0	0	0	0	0	0	0	0
			female	1976	1	0	0	0	0	0	0	0	0
			female	1987	1	0	1	0	0	0	0	0	0
			male	1983	1	1	0	0	0	0	0	0	0
			female	1991	0	0	0	0	0	0	0	0	0
			female	1987	1	0	0	0	0	0	0	0	0
			male	1965	0	0	0	0	0	0	0	0	0
			male	1960	0	1	0	0	0	0	0	0	0
			female	1970	1	0	0	0	0	0	0	0	0

WHAT IS THE STATUS?

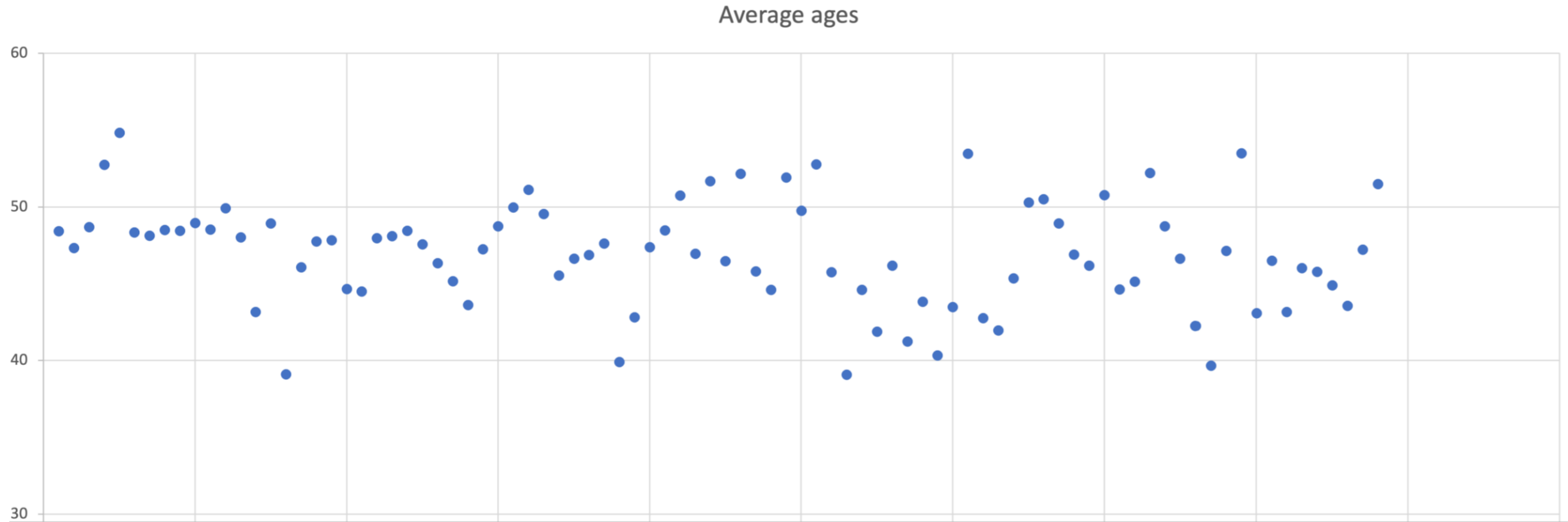
- Still in data acquisition phase
- Current dataset w/ 400 random emails, «training data»
- gender/age are known (self-declared), signup often only 2 options, whole spectrum uncovered, not an ideal repr. of community



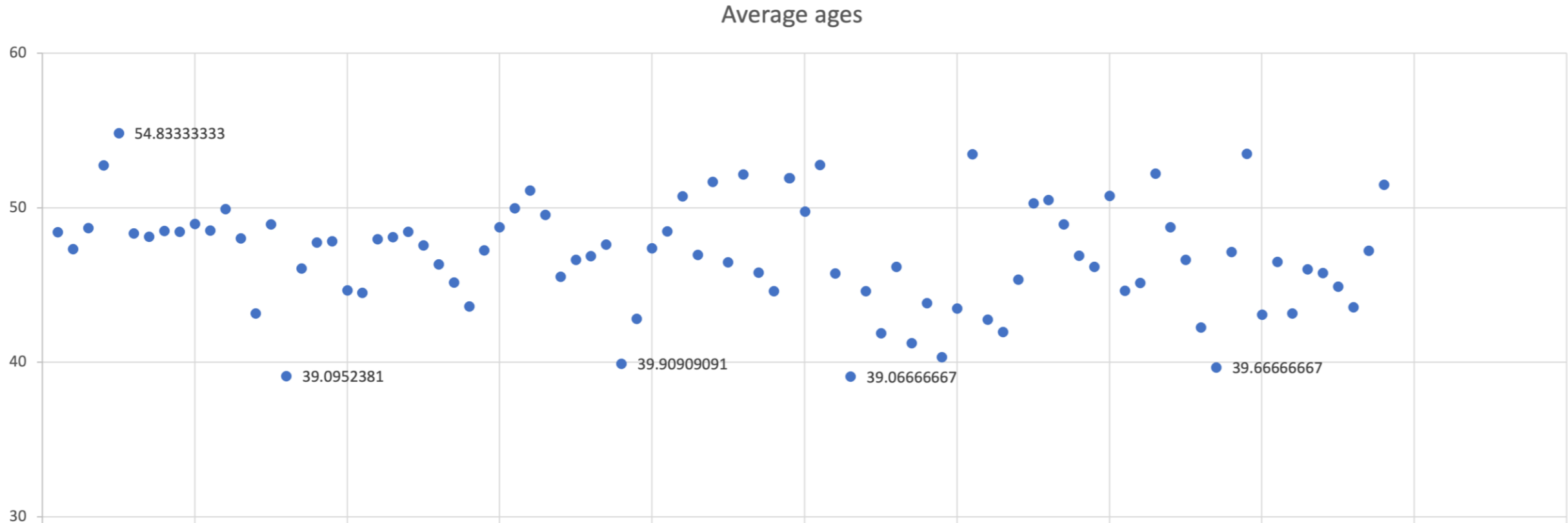


OBSERVATIONS

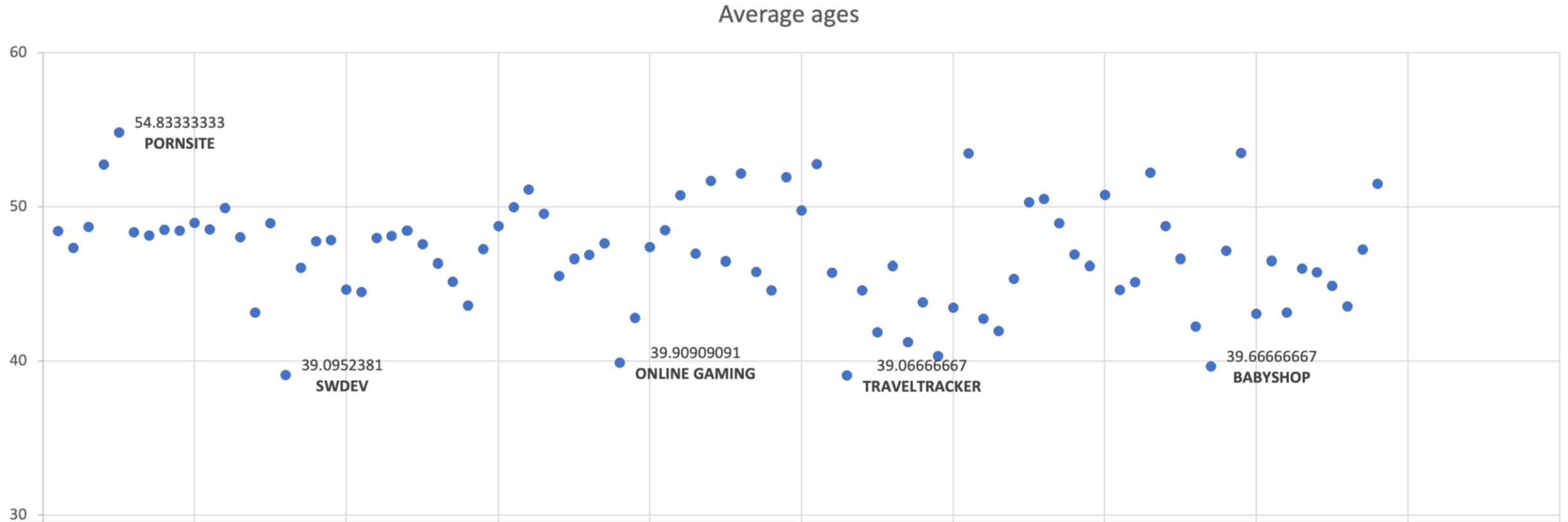
OBSERVATIONS I – AVERAGE AGES



OBSERVATIONS I – AVERAGE AGES MINIMA/MAXIMA

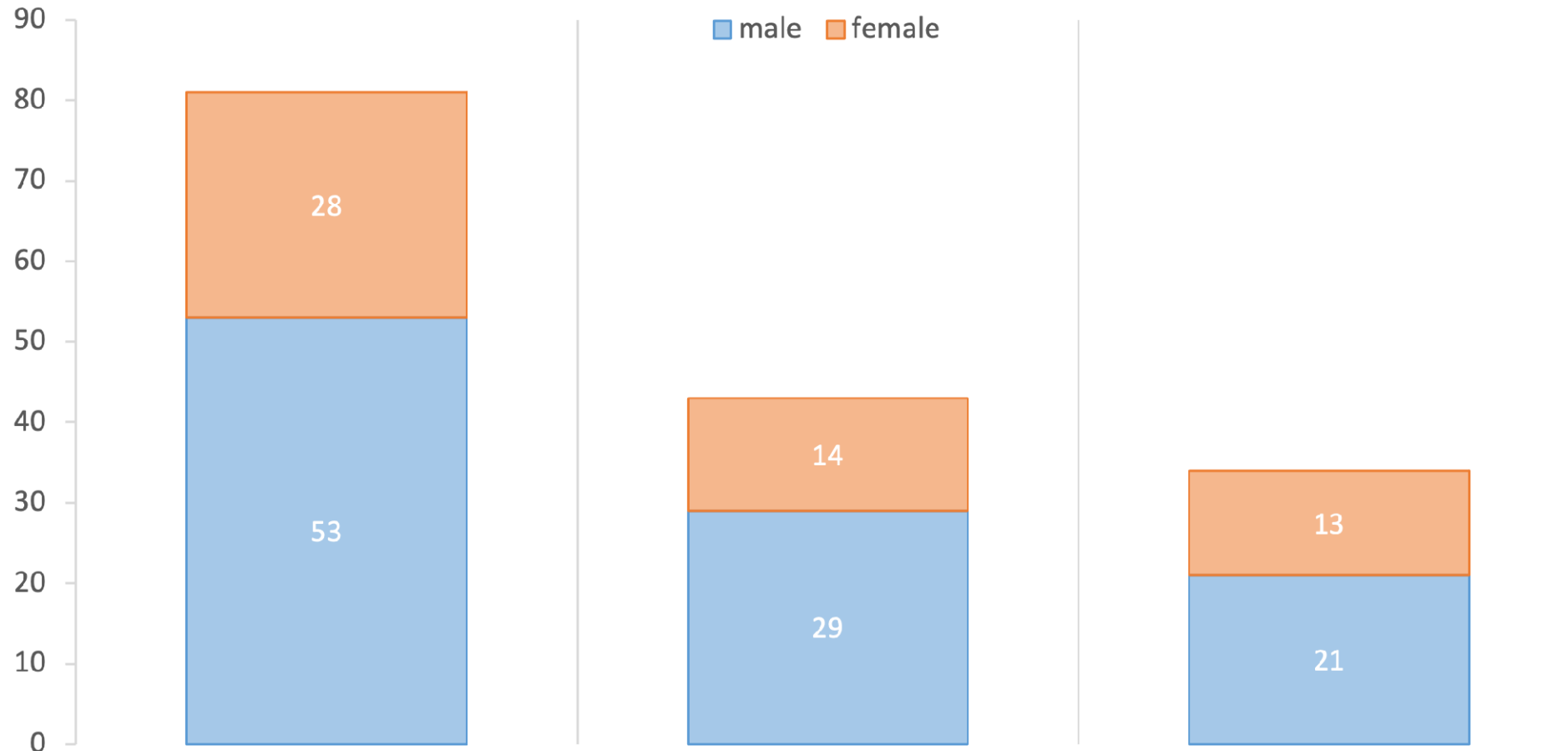


OBSERVATIONS I – AVERAGE AGES MINIMA/MAXIMA



OBSERVATIONS II – GENDER ID DISTRIBUTION TECHSHOPS

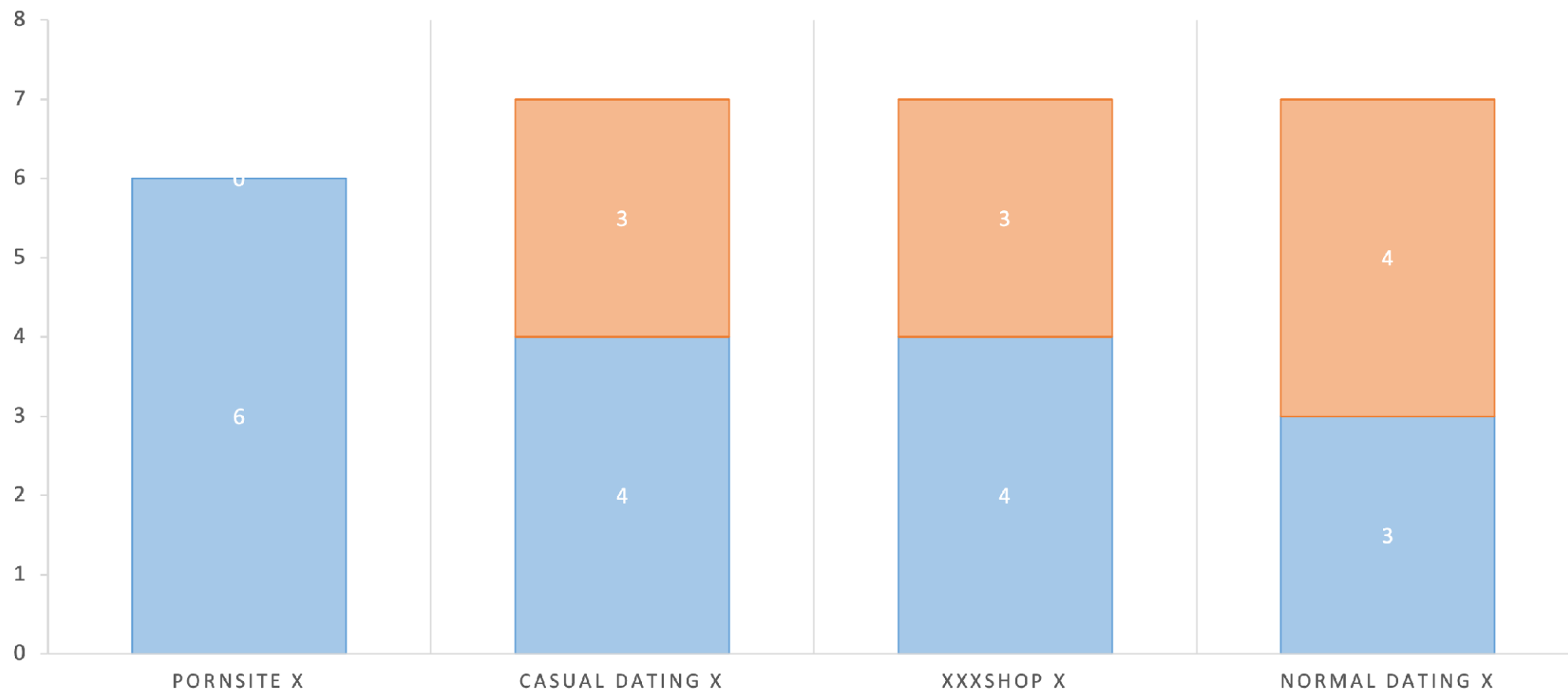
ONLINE SHOPS TECHNOLOGY



OBSERVATIONS II – GENDER ID DISTRIBUTION ADULT

EROTICA / ADULT / DATING

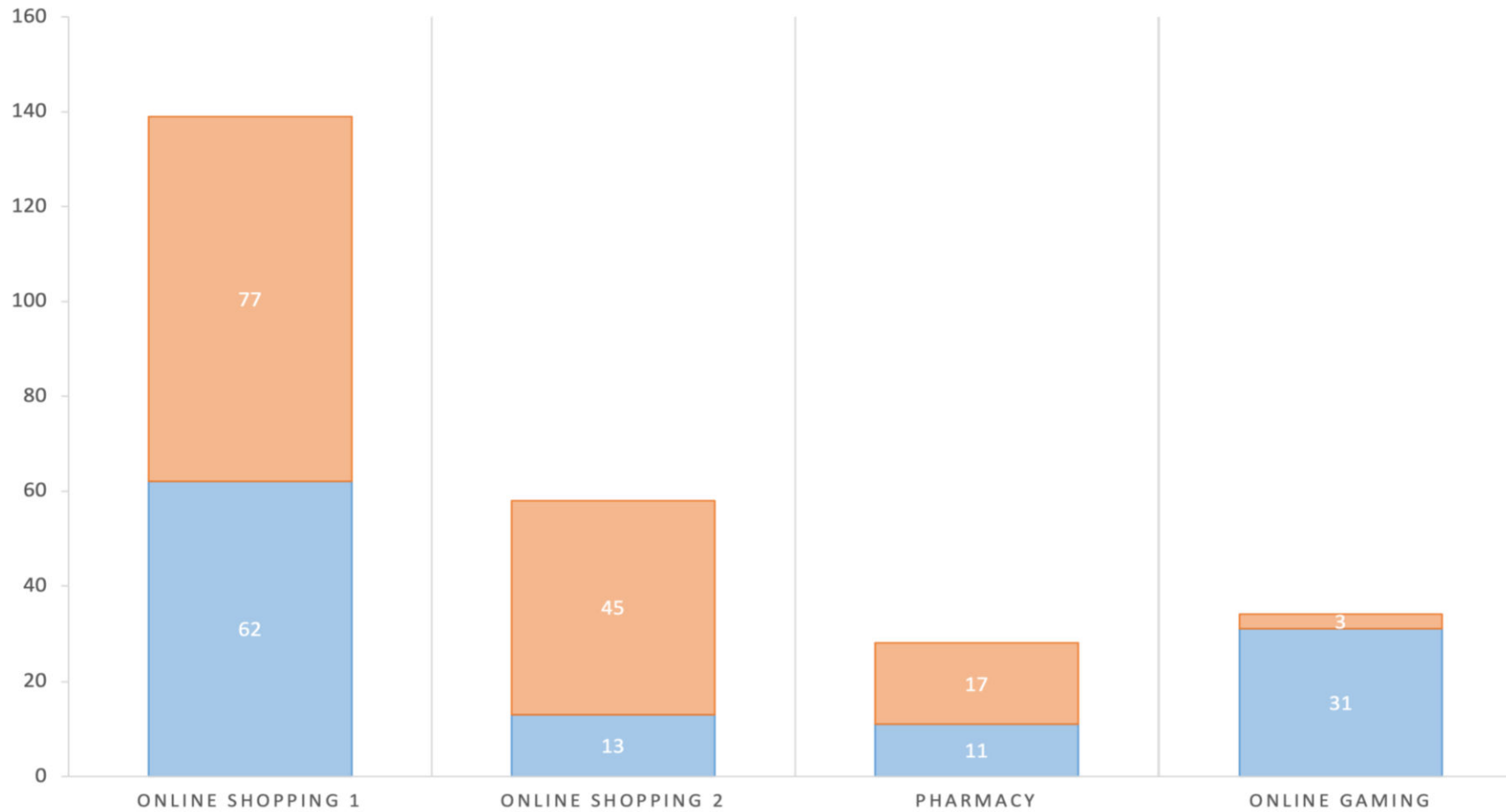
■ male ■ female



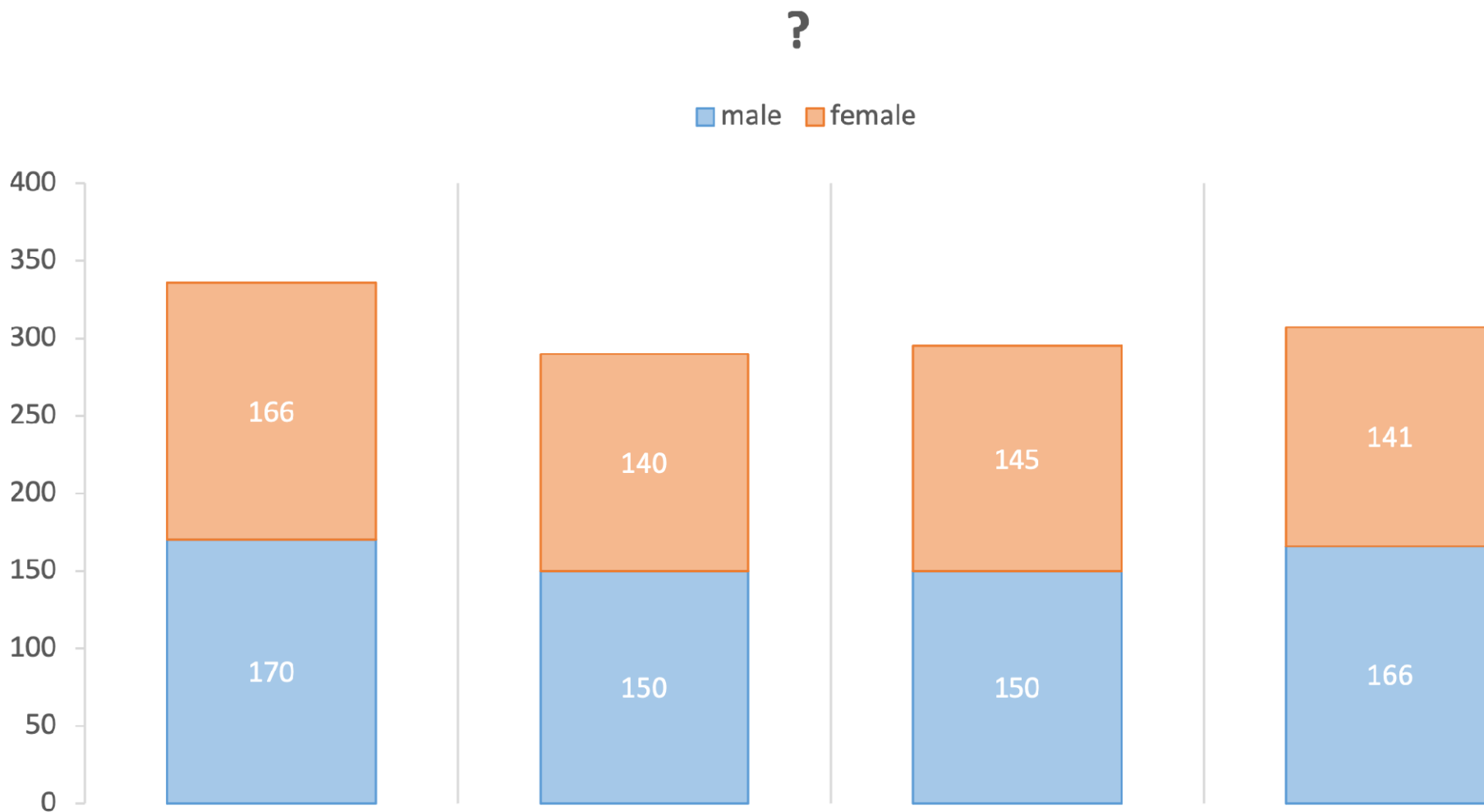
OBSERVATIONS II – GENDER ID DISTRIBUTION MISC

VARIOUS

■ male ■ female



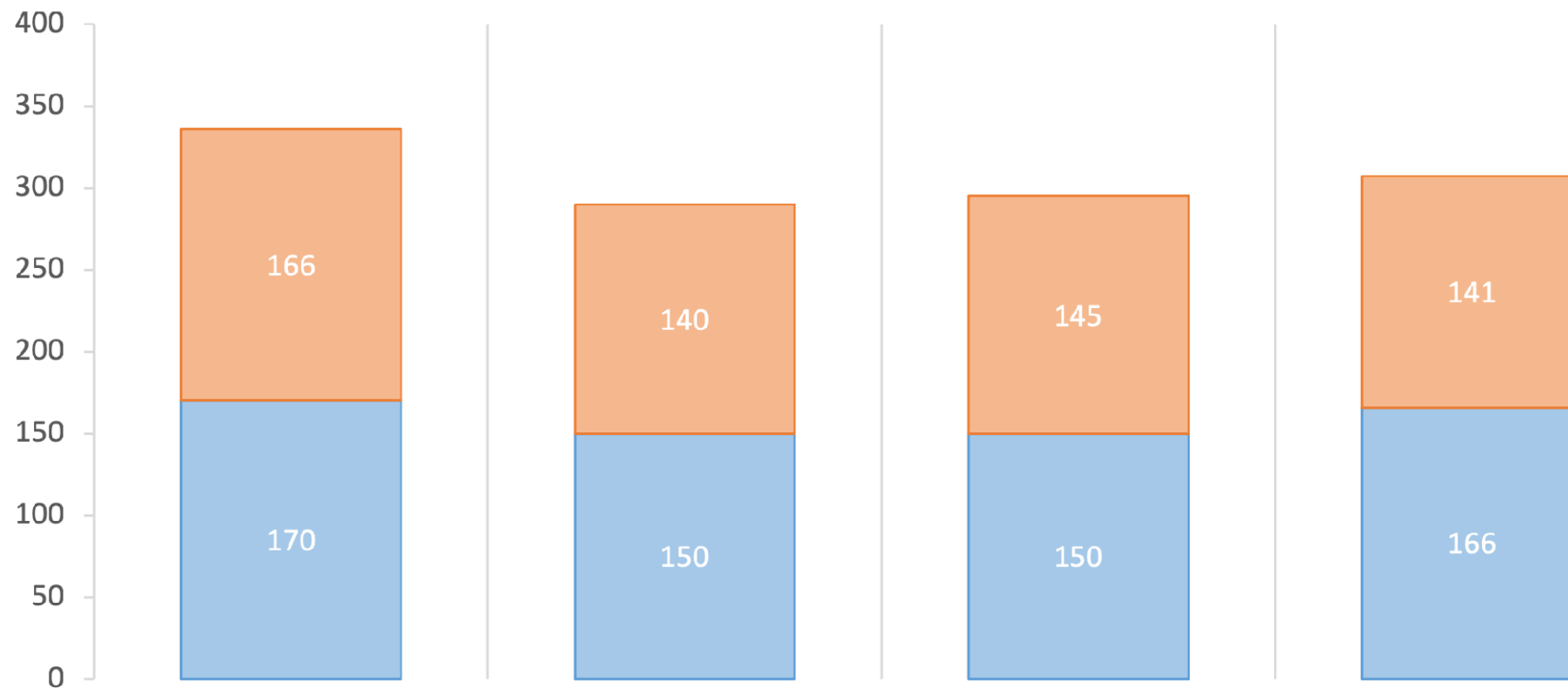
OBSERVATIONS II – ?



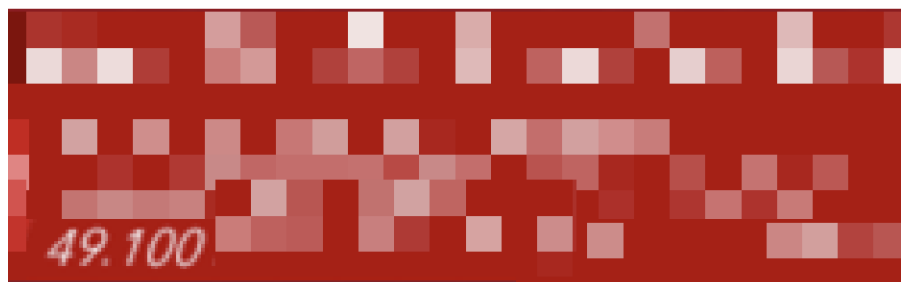
OBSERVATIONS II

BIG TECH

■ male ■ female



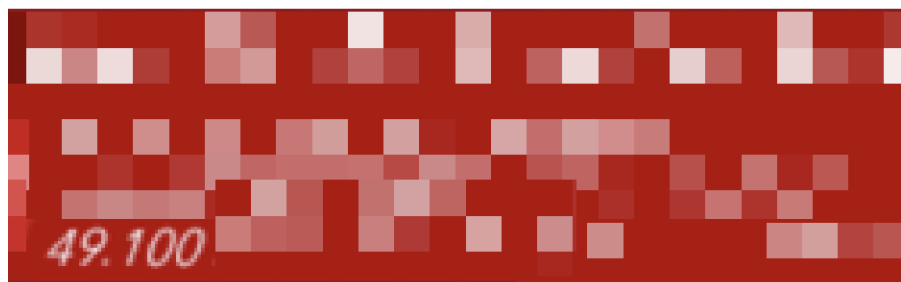
OBSERVATIONS III – XXX FORUM (STATISTICS)



Assumption:

- $100/8700000 * 49100 \sim 0.5\%$ of population registered
- $8700000/49100 \sim 177$, every 177th citizen registered

OBSERVATIONS III – XXX FORUM (STATISTICS)



Assumption:

- $100/8700000 * 49100 \sim 0.5\%$ of population registered
- $8700000/49100 \sim 177$, every 177th citizen registered

400 dataset:

- #144 and #353 registered
- blackmailable? → use throwaway e-mail address



PROFILING

PROFILING I - FINDING RELATIONSHIPS: PROTOTYPES

1	0	0	0	0	0	0	0	0	0	1	0	0	0
1	1	1	0	0	0	0	0	0	0	1	0	0	1
1	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	1	0	0	0	0	0	0	1	0	0	0	0
0	0	1	1	0	0	0	0	0	1	0	0	1	0
1	0	1	0	0	0	0	0	0	1	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	1	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	1	0	0	0	0
1	0	0	1	1	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0	1	0	0	0	0
0	0	1	0	0	0	0	0	0	1	0	0	0	0
1	0	0	1	0	0	0	0	0	0	0	0	0	0
0	0	1	0	0	0	0	0	0	1	0	0	0	0
1	0	1	1	0	0	0	0	0	0	0	0	0	0
1	1	1	1	0	0	0	0	0	1	1	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	1	0	0	0	0	0	0	0	0	1	0	0
1	0	0	0	0	0	0	0	0	1	0	0	0	1
1	1	0	0	0	0	0	0	0	1	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	1
1	0	1	0	0	0	0	0	0	1	0	0	0	1

Prototype 1
„Outdoor enthusiast“

Prototype 2
„Hipster Coder“

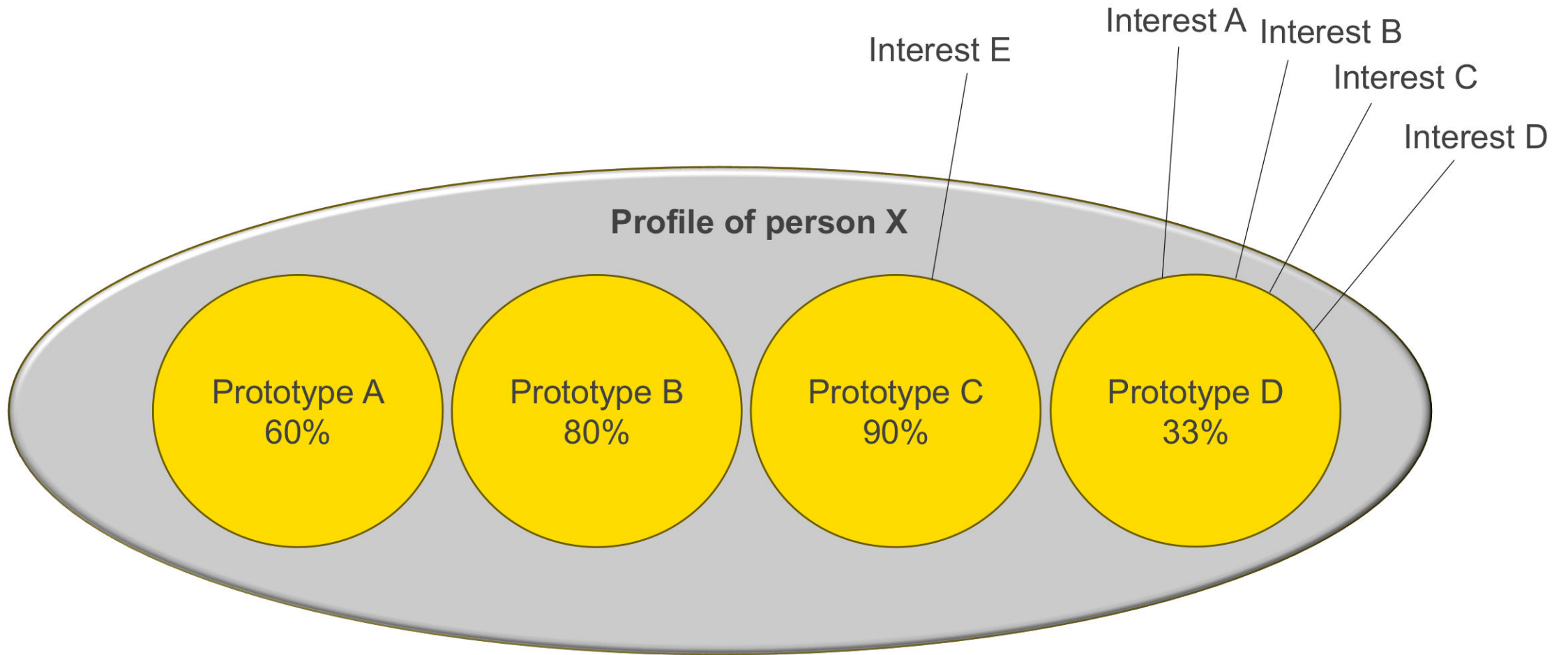
PROFILING I - FINDING RELATIONSHIPS: PROTOTYPES

1	0	0	0	0	0	0	0	0	0	1	0	0	0
1	1	1	0	0	0	0	0	0	1	0	0	1	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	1	0	0	0	0	0	0	1	0	0	0	0
0	0	1	1	0	0	0	0	0	1	0	0	1	0
1	0	1	0	0	0	0	0	0	1	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	1	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	1	0	0	0	0
1	0	0	1	1	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0	1	0	0	0	0
0	0	1	0	0	0	0	0	0	1	0	0	0	0
1	0	0	1	0	0	0	0	0	0	0	0	0	0
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1	0	1	1	0	0	0	0	0	0	0	0	0	0
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1	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	0
1	0	1	0	0	0	0	0	0	0	0	1	0	0
1	0	0	0	0	0	0	0	0	1	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	0	0	1
1	0	1	0	0	0	0	0	0	1	0	0	0	1

Prototype 1
„Outdoor enthusiast“
50% match

Prototype 2
„Hipster programmer“
71.4% match

PROFILING I - FINDING RELATIONSHIPS: PROTOTYPES

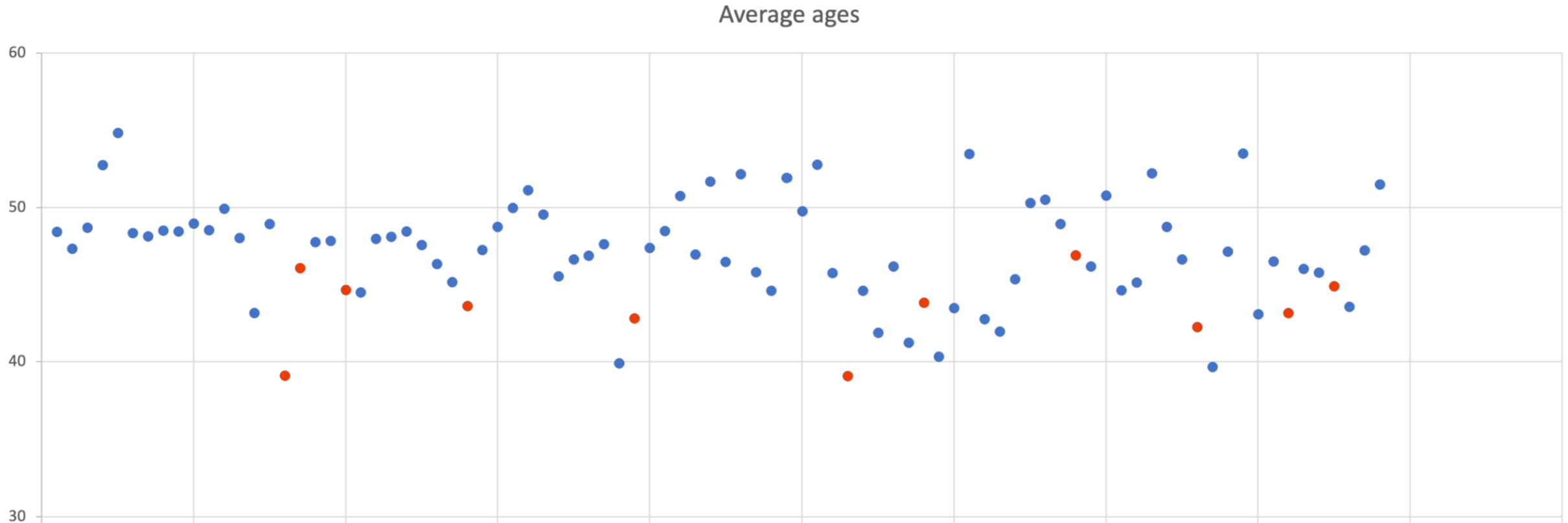


Fact that s.o. is NOT registered at platform X may also be valuable (meta-)information (or # of regs)

PROFILING II – PREDICTING PROPERTIES

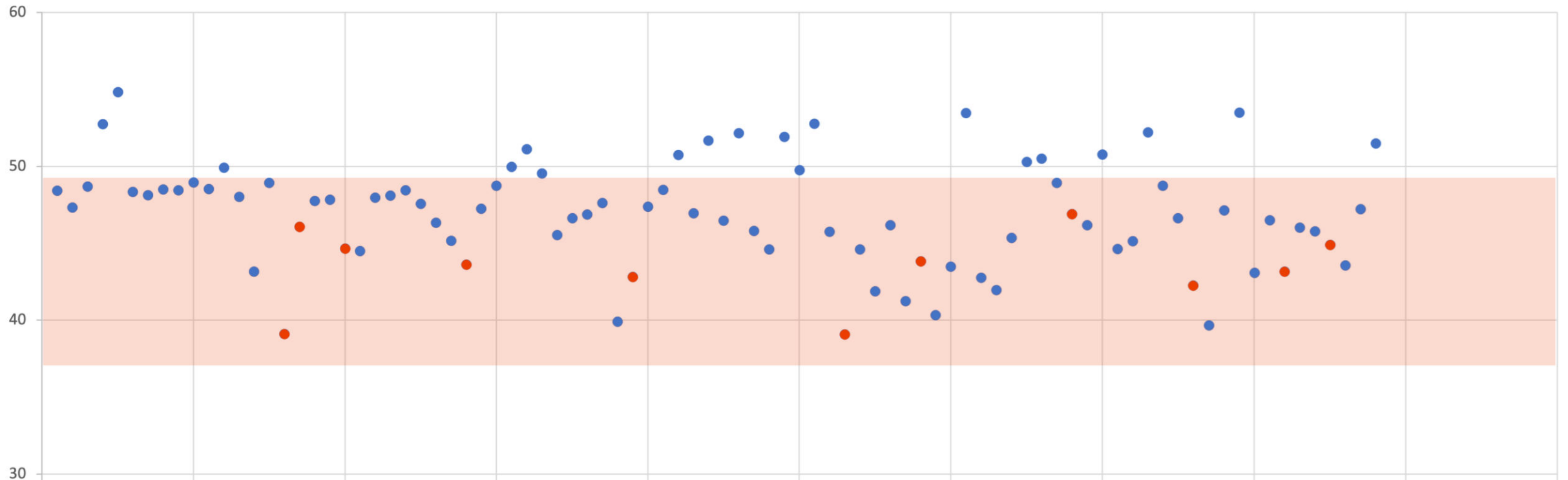
- What if we do not know age, gender id, (profession)?
- Using the data and tool, can we predict these for an unknown person behind an arbitrary email?
- (estimate userbase?)

PROFILING II – PREDICTING AGE

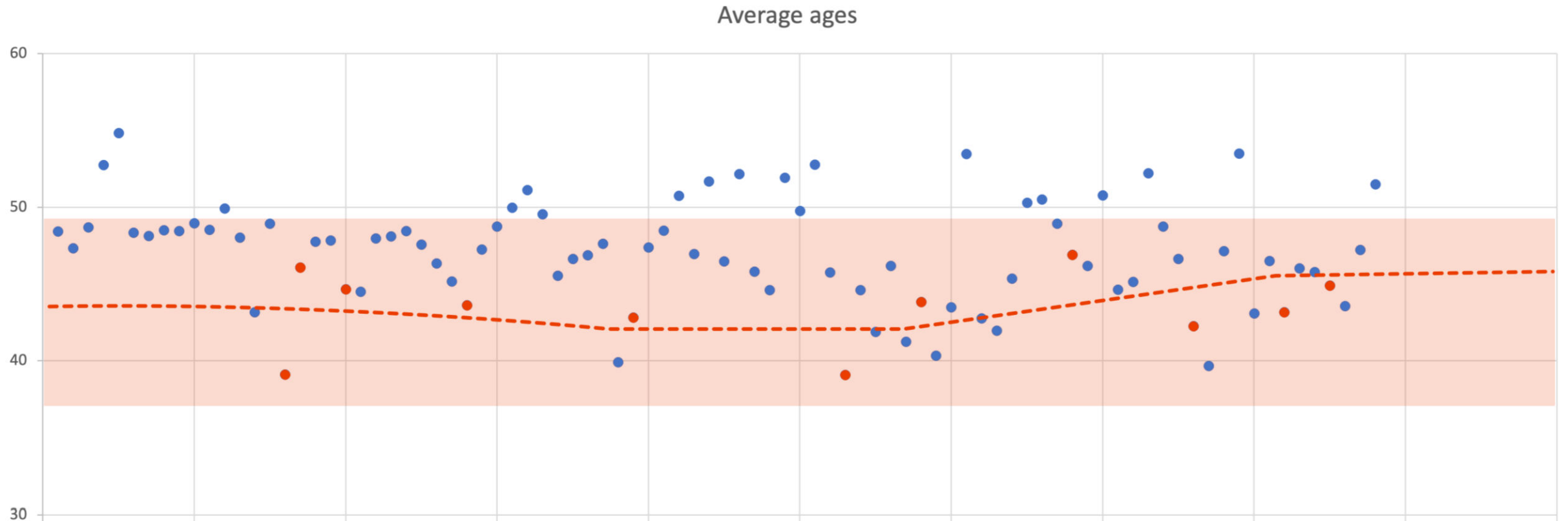


PROFILING II – PREDICTING AGE

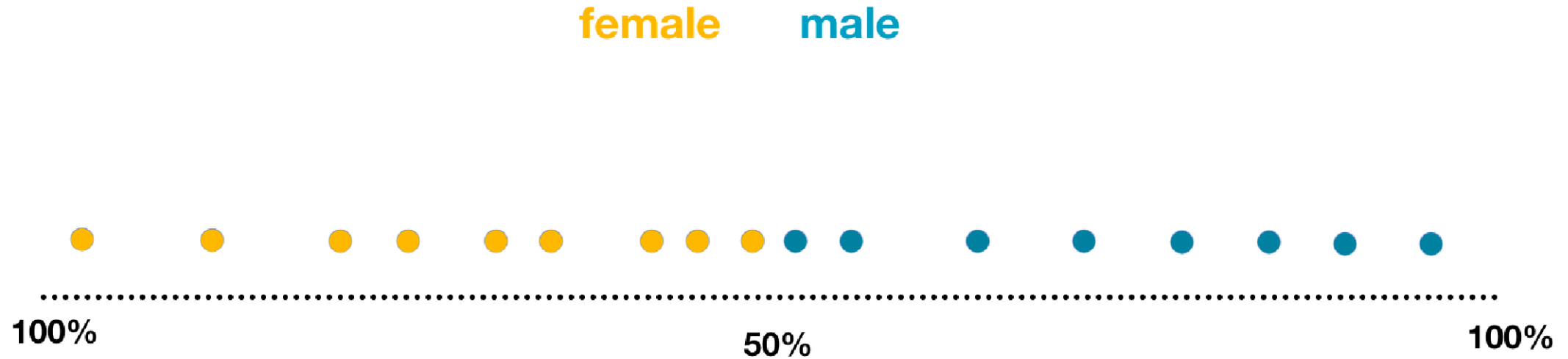
Average ages



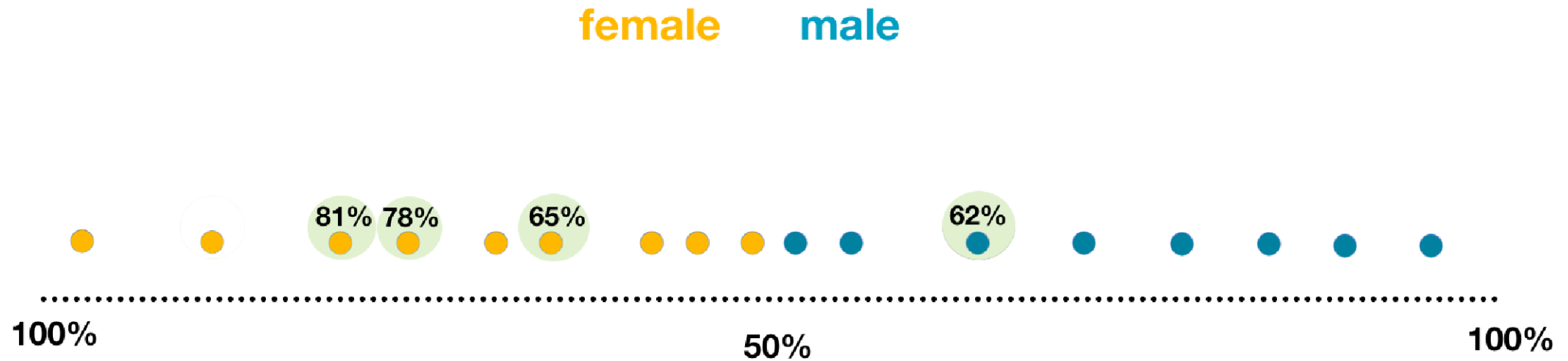
PROFILING II – PREDICTING AGE



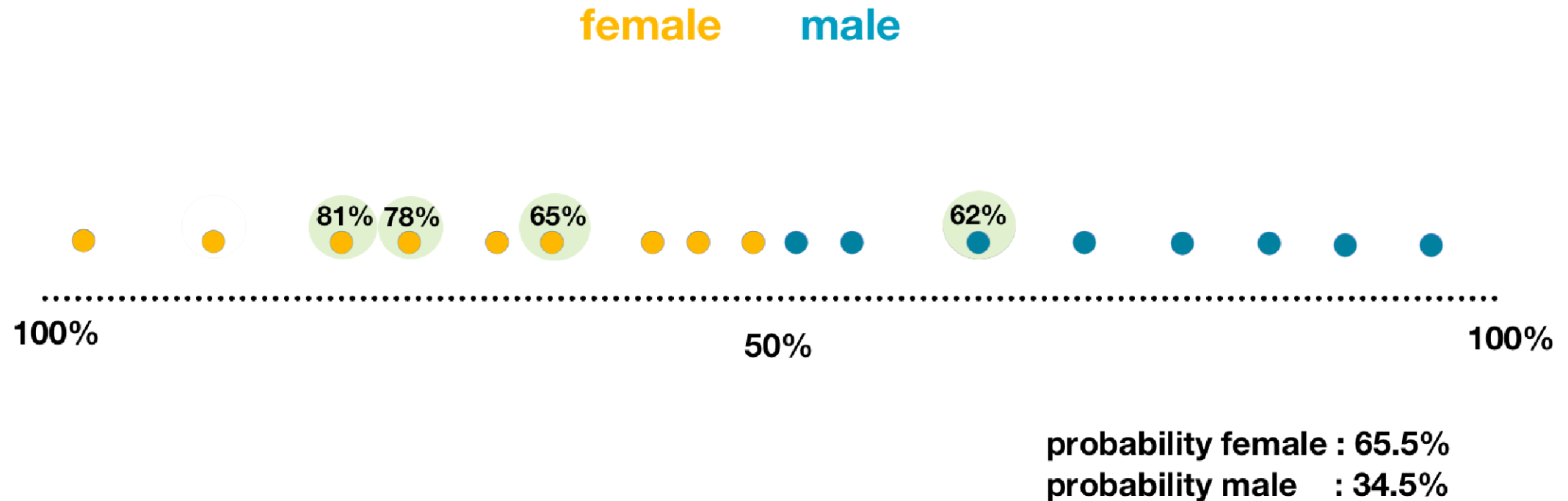
PROFILING II – PREDICTING GENDER ID

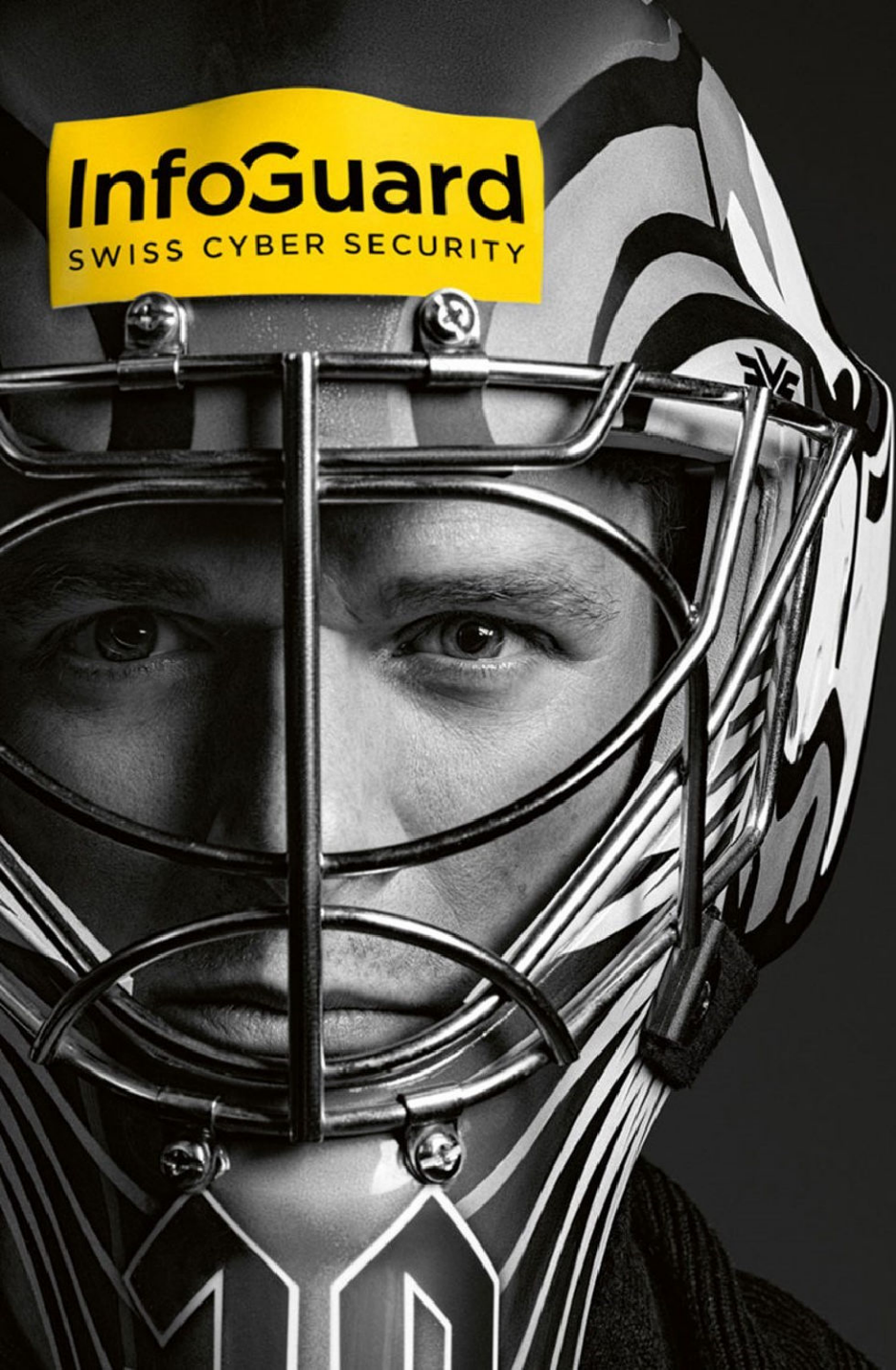


PROFILING II – PREDICTING GENDER ID



PROFILING II – PREDICTING GENDER ID

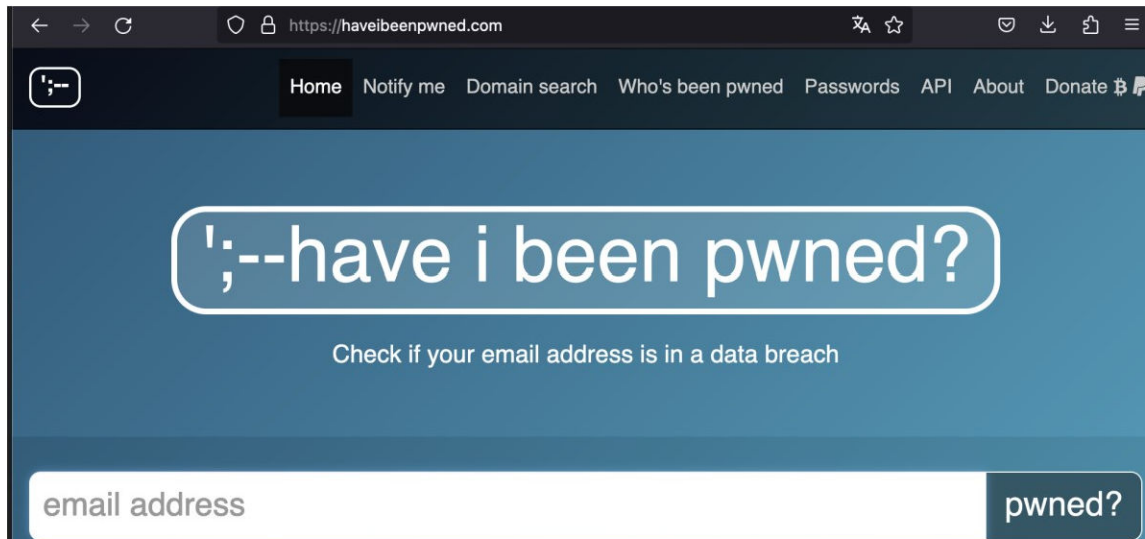




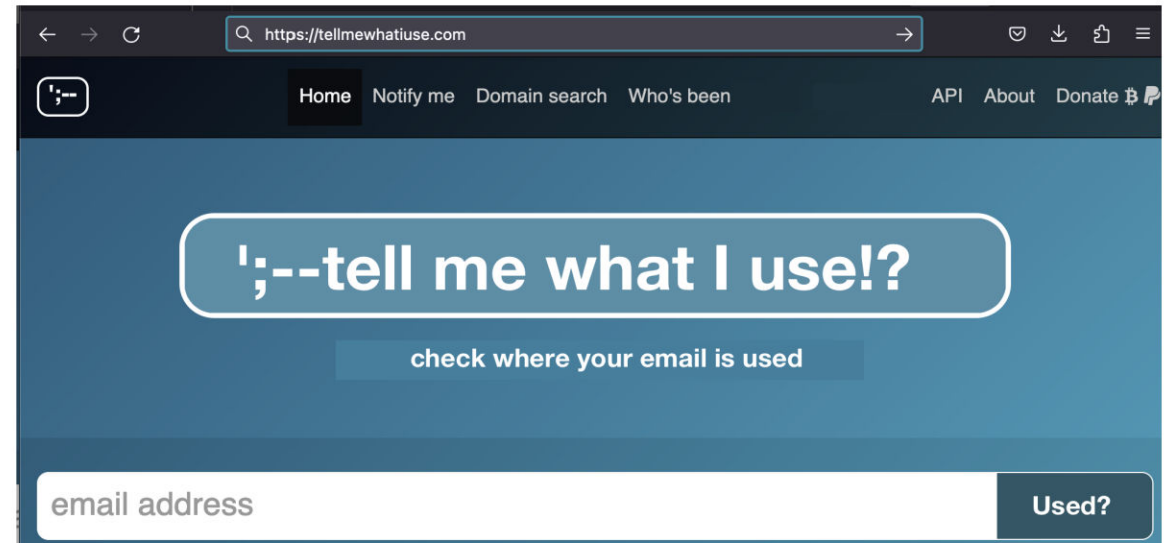
OUTLOOK

SERVICE?

haveibeenpwned.com



tellmewhatius.com



FINAL REMARKS

- Still in data acquisition phase
- Analysis of results incomplete
- Trends already visible with few data (400)
- Does not scale well, a study would need more data to be meaningful
- Evaluating ML methods, model training, tbd
- More advanced predictions: political orientation? religion?
- If its possible, it will (or is) be done

QUESTIONS/DISCUSSION





THANK YOUR FOR YOUR ATTENTION!

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