

Signatures of functional responses to acute and chronic COVID-19

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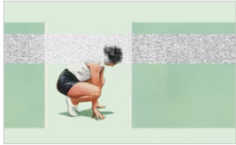
University of Washington 2023 Physics REU Program

Mentors: Zach Montague and Armita Nourmohammad

Introduction

PASC, the immune system, and our data

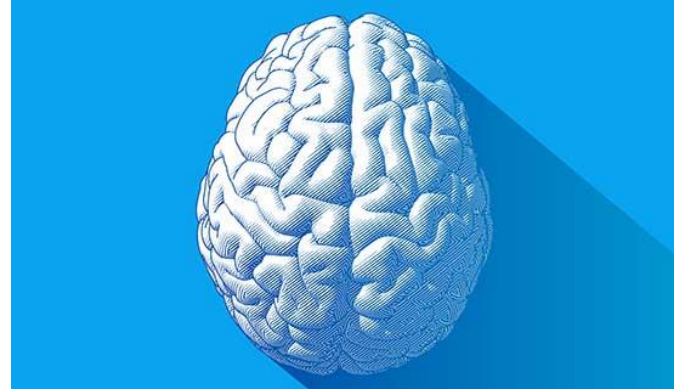
PASC (Long Covid)



Long COVID Has Forced a Reckoning for One of Medicine's Most Neglected Diseases

Only a couple dozen doctors specialize in chronic fatigue syndrome (ME/CFS). Now their knowledge could be crucial to treating millions more patients.

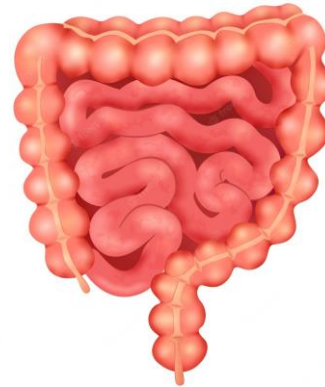
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The data is clear: long Covid is devastating people's lives and livelihoods

31-69%

The number of people who go on to develop PASC after having COVID-19



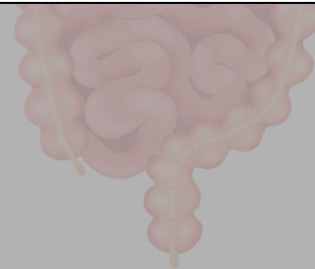
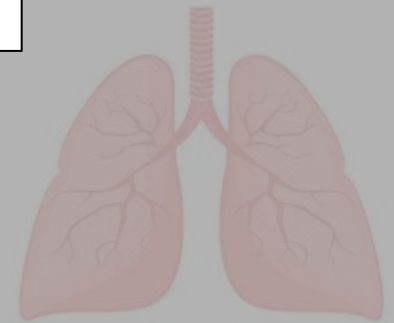
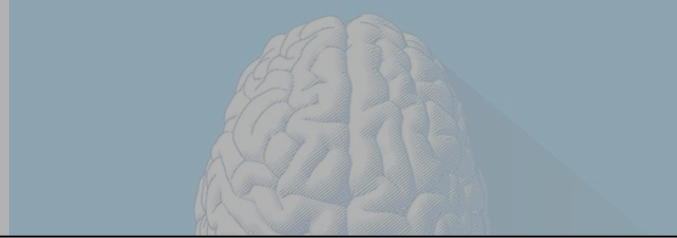
PASC (Long Covid)



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The data i
people's li

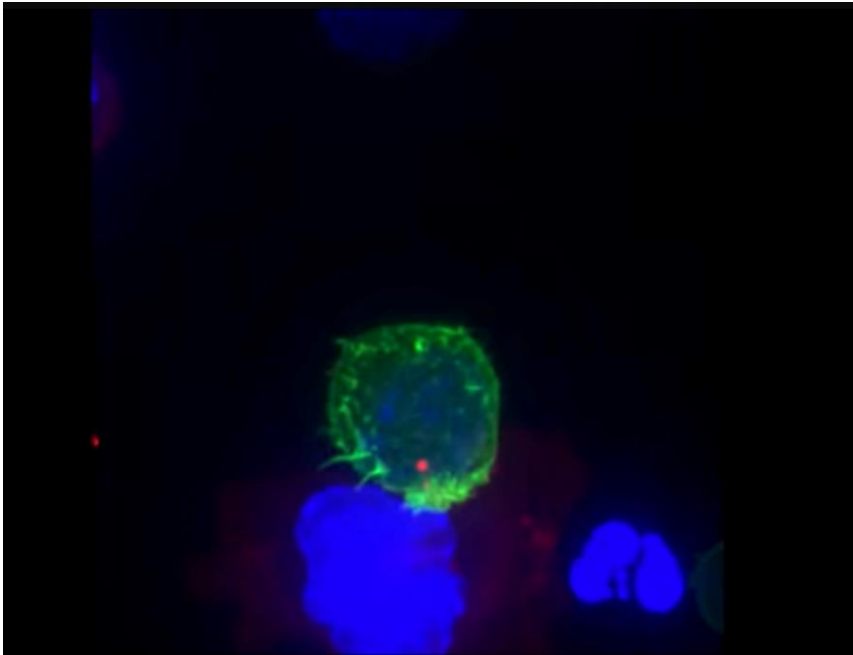


Can we identify predictors of PASC from the immune response to acute COVID-19?

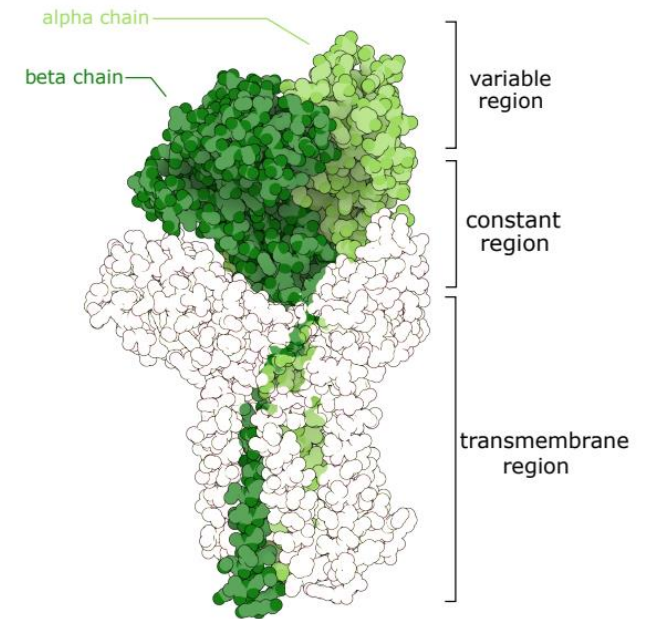
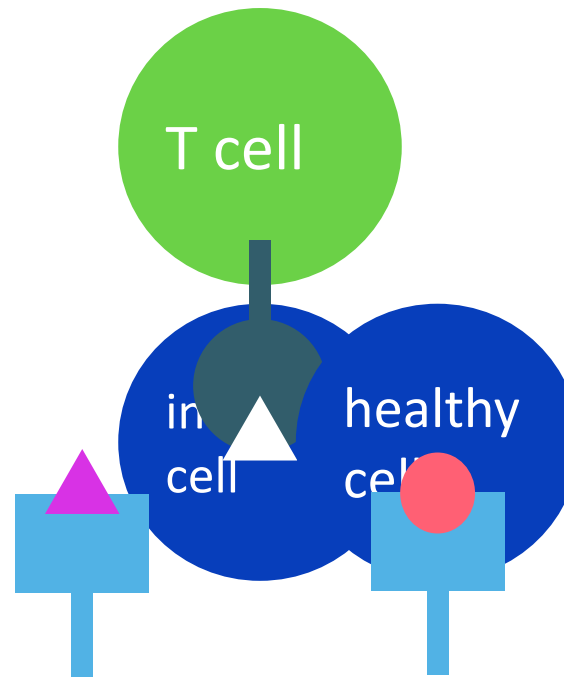
31-69%

The number of people who go on to develop PASC after having COVID-19

A brief overview of T-cell function

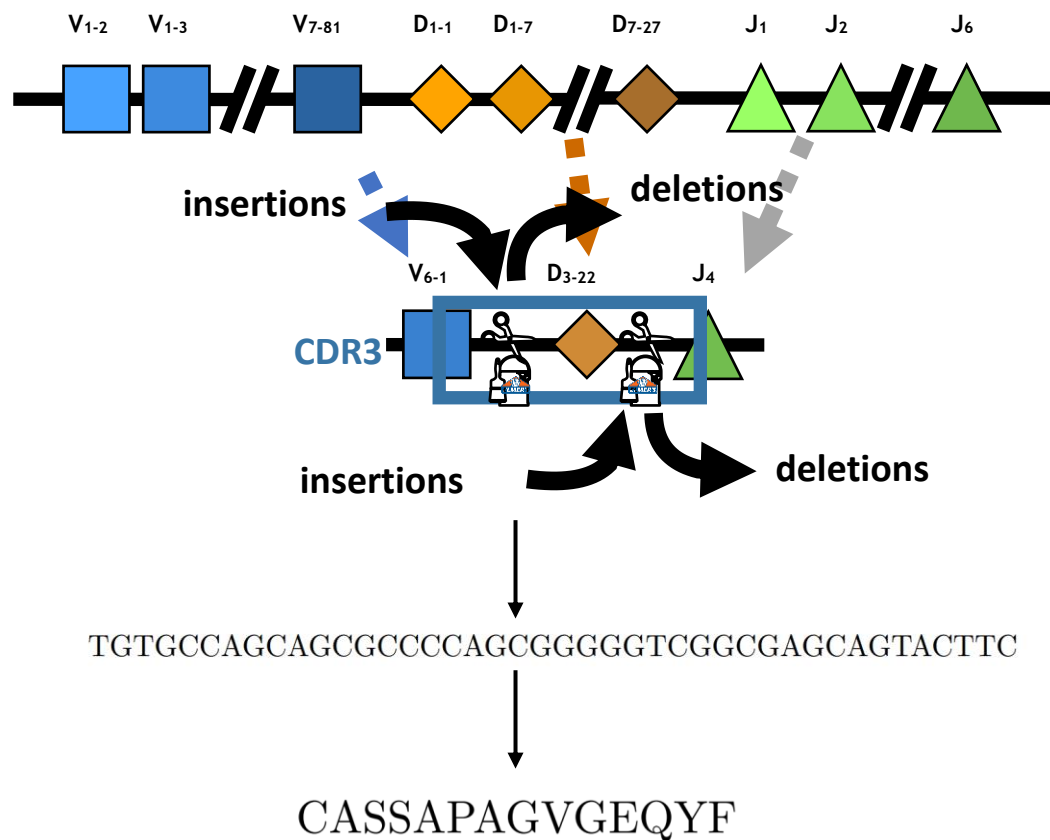


video: Griffiths lab, Cambridge Institute for Medical Research

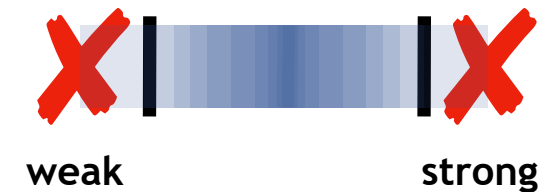


Generation of TCR diversity

VDJ recombination



testing for functionality



~ 10^{18} potential TCRs
 10^{12} in body
 10^6 in data – sampling variation!

Montague, 2023
 Janeway, 2001

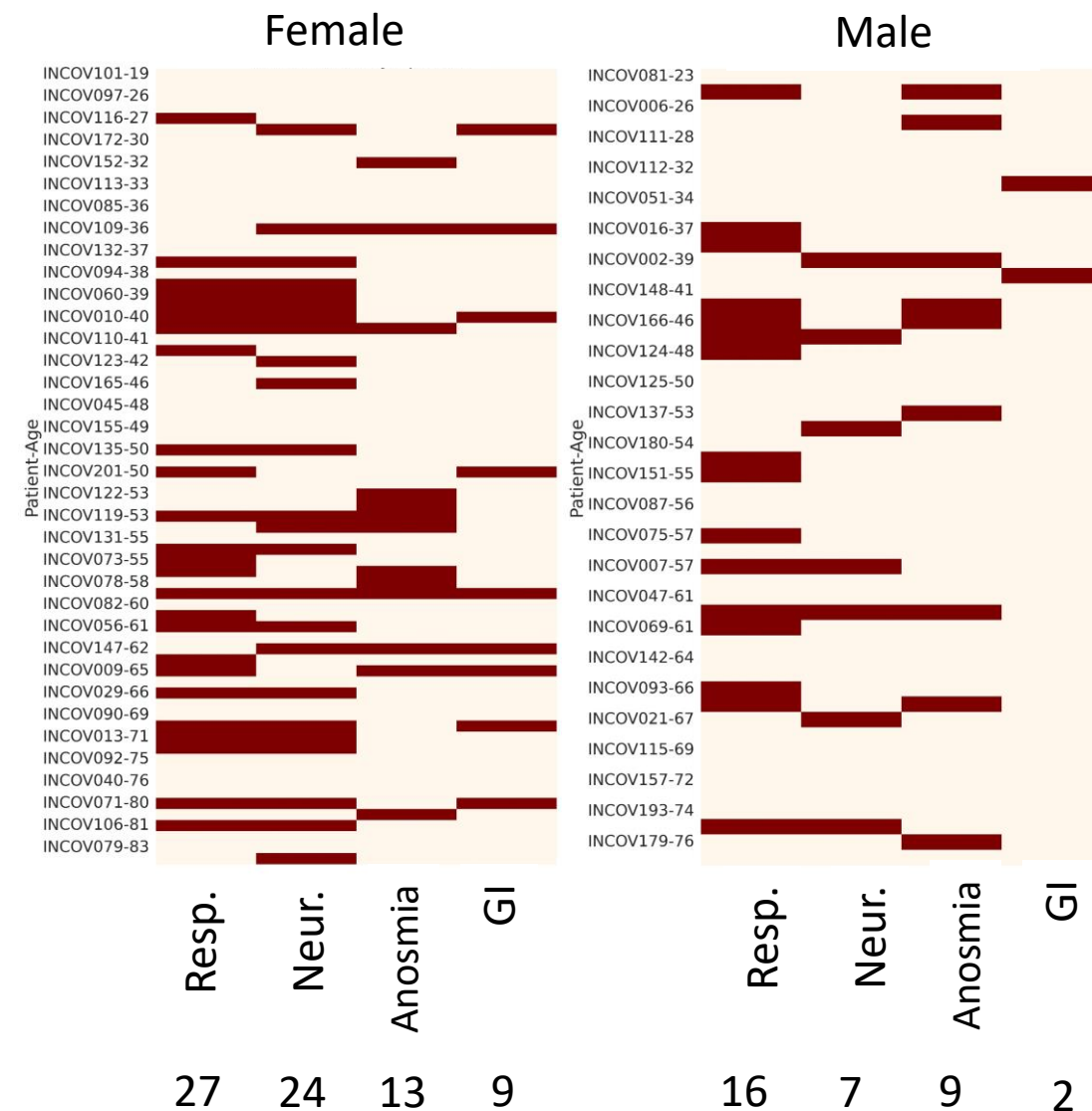
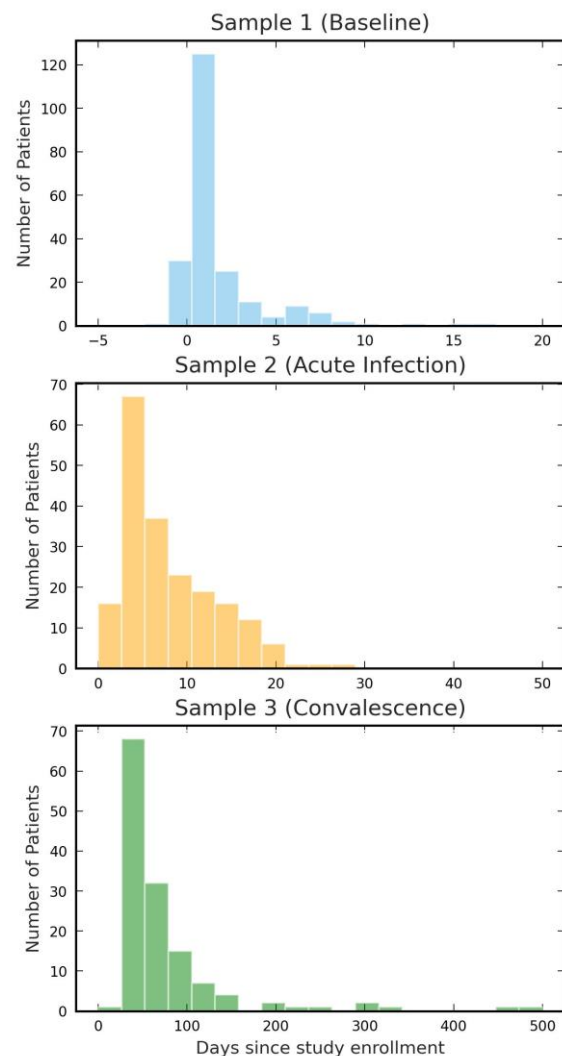
PASC immune repertoire data

63
 patients
 without PASC
 ↓
28
 male
35
 female

61
 patients with
 PASC
 ↓
24
 male
37
 female

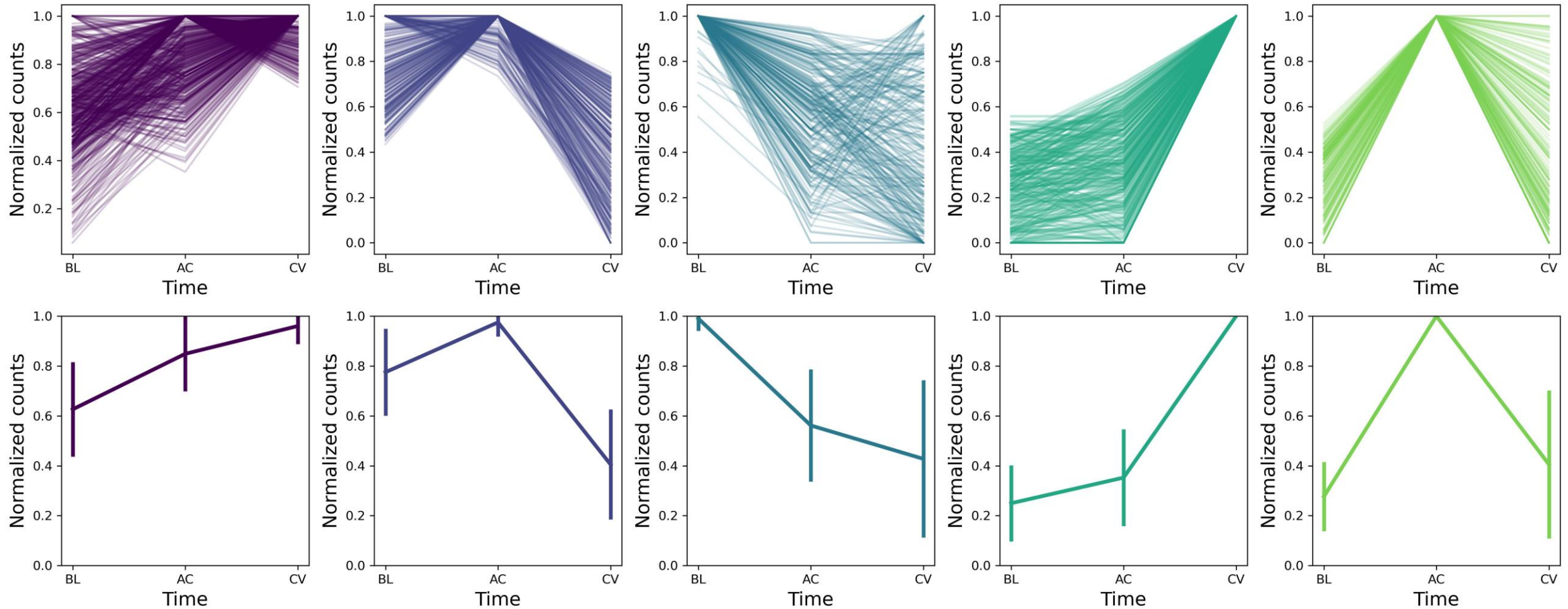
aged
19-86

274, 923
 avg # of unique TCRs per patient

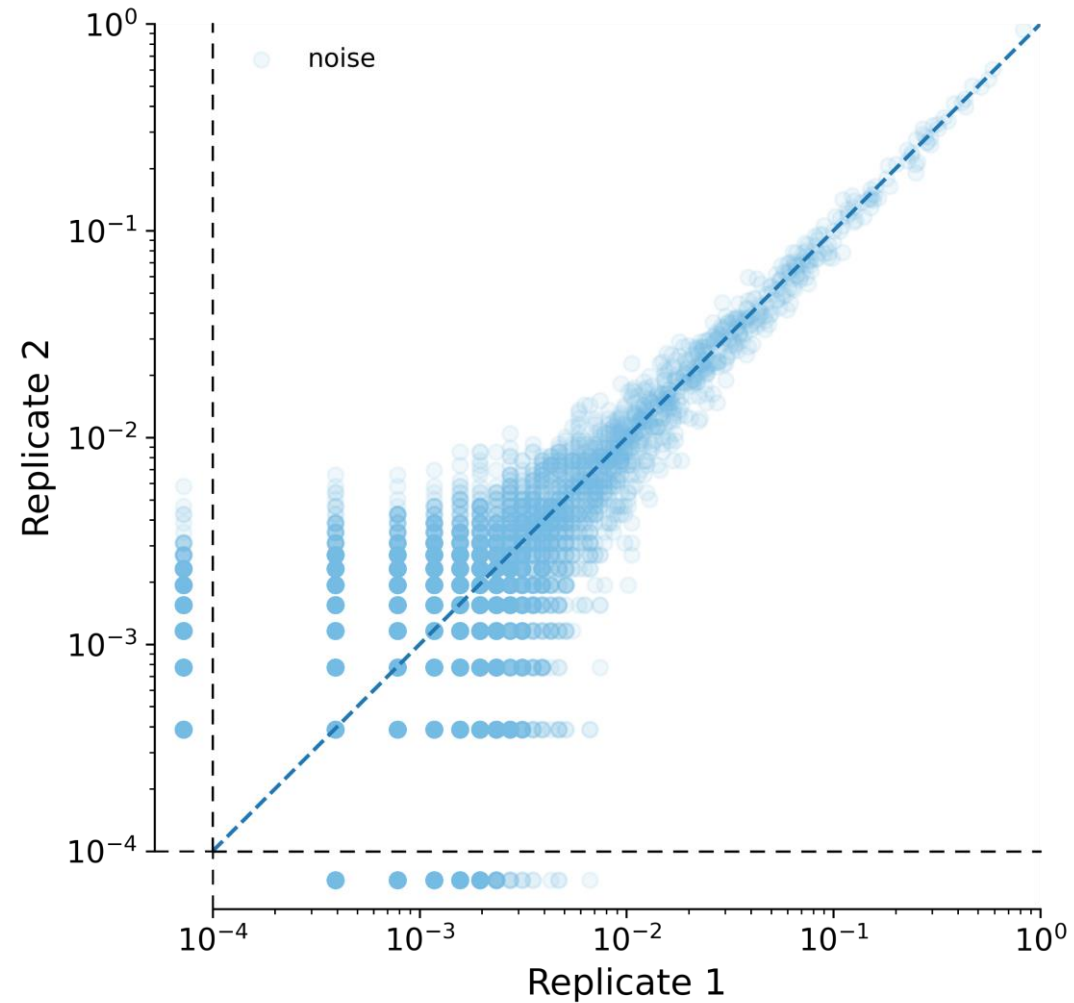


T Cell Population Dynamics Analysis

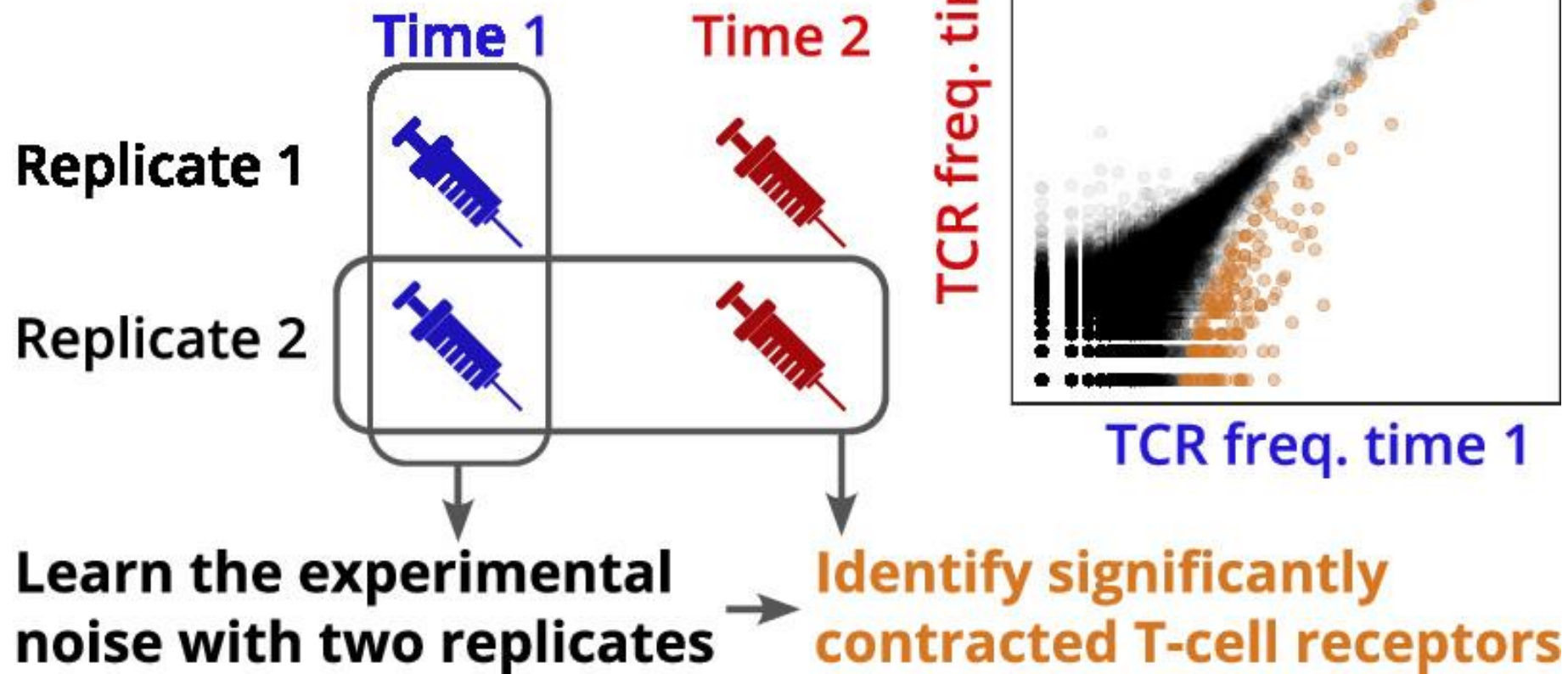
Clone Read Trajectories



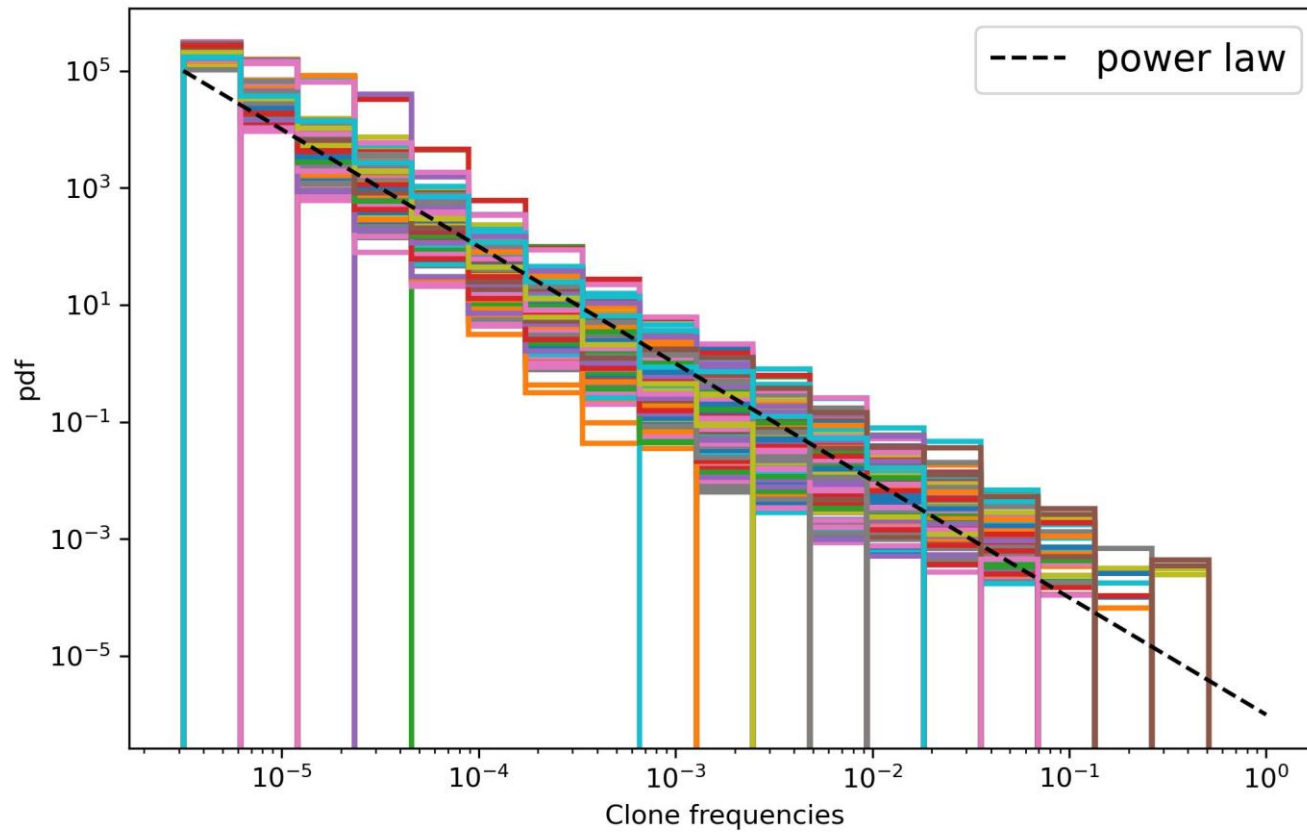
Undersampling leads to variation in TCR reads



NoisET

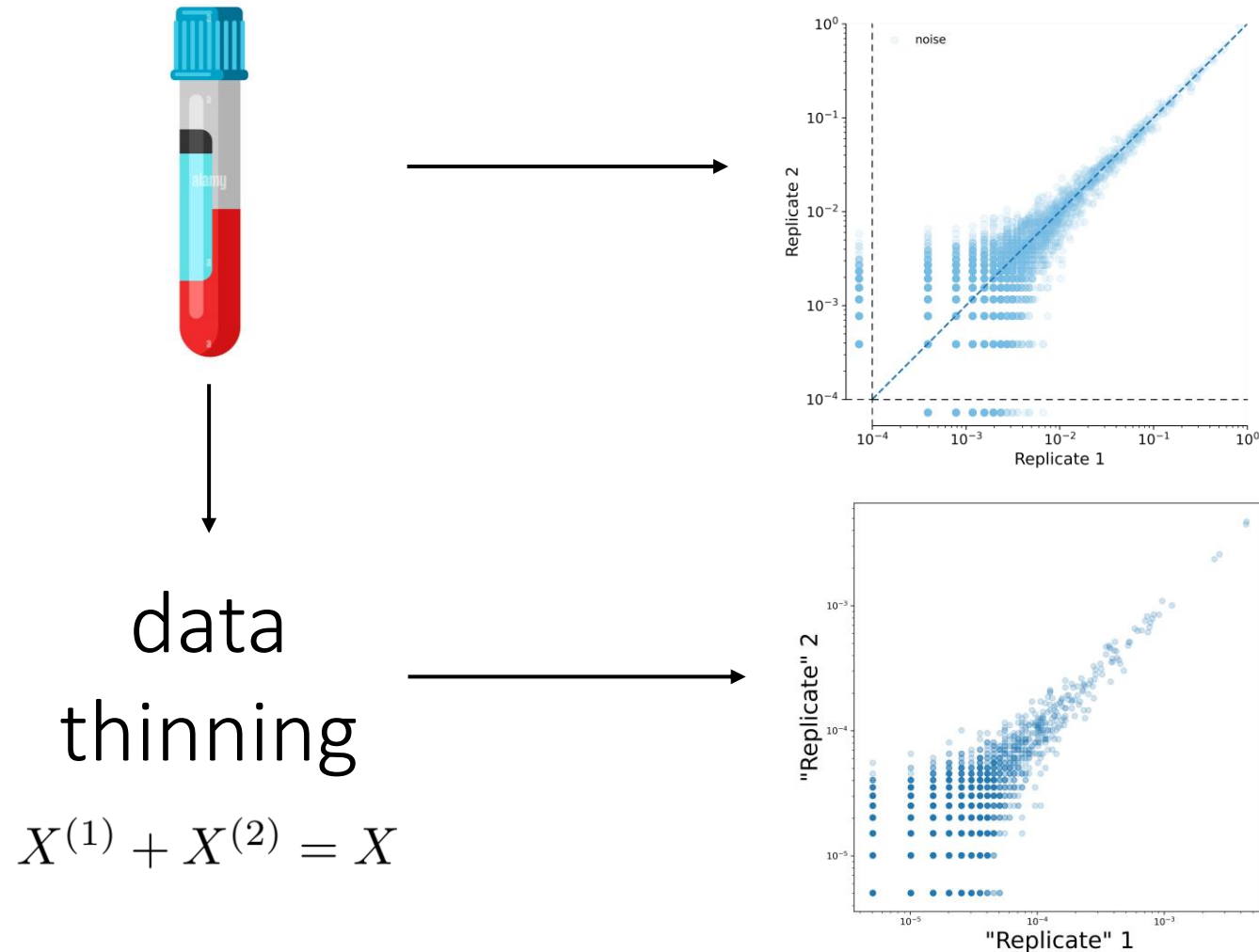


Clone size distributions follow a power law



$$\rho(f) = C f^{-2}$$

Data thinning in lieu of biological replicates



Noise Model Choices

$$P(\hat{n}|f) = \text{Pois}(f N_r)$$

$$P(\hat{n}|f) = \text{NegBin}(\hat{n}; N_r f, N_r f + a(N_r f)^b)$$

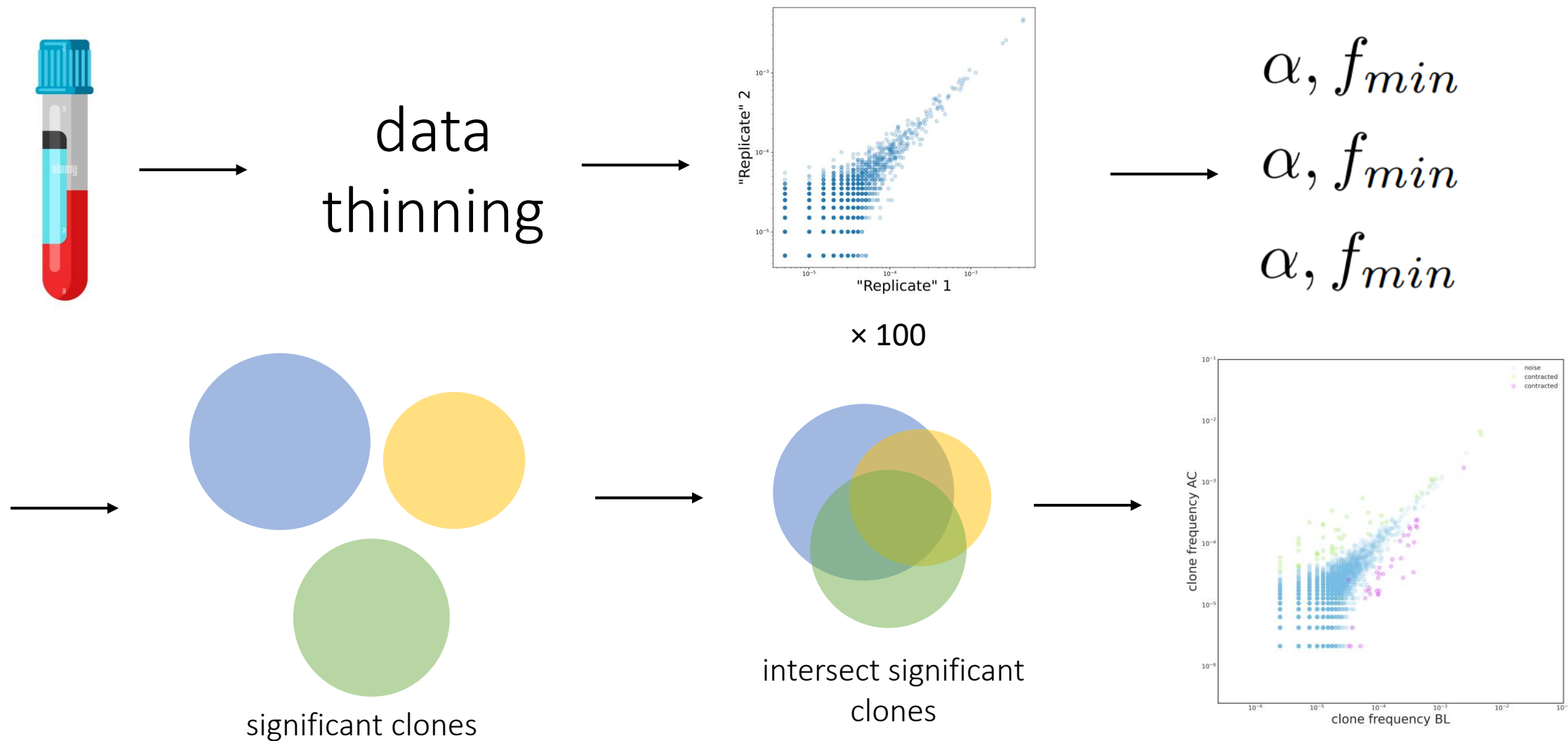
Expansion Model

$$f(t_1) = f \longrightarrow f(t_2) = fe^s$$

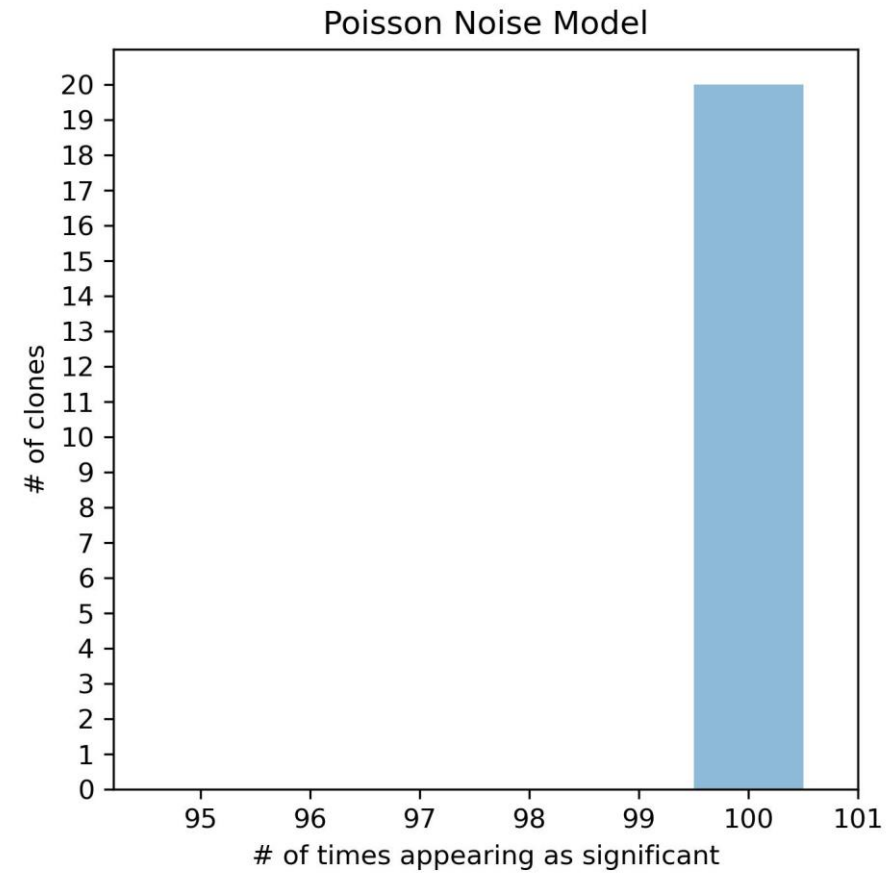
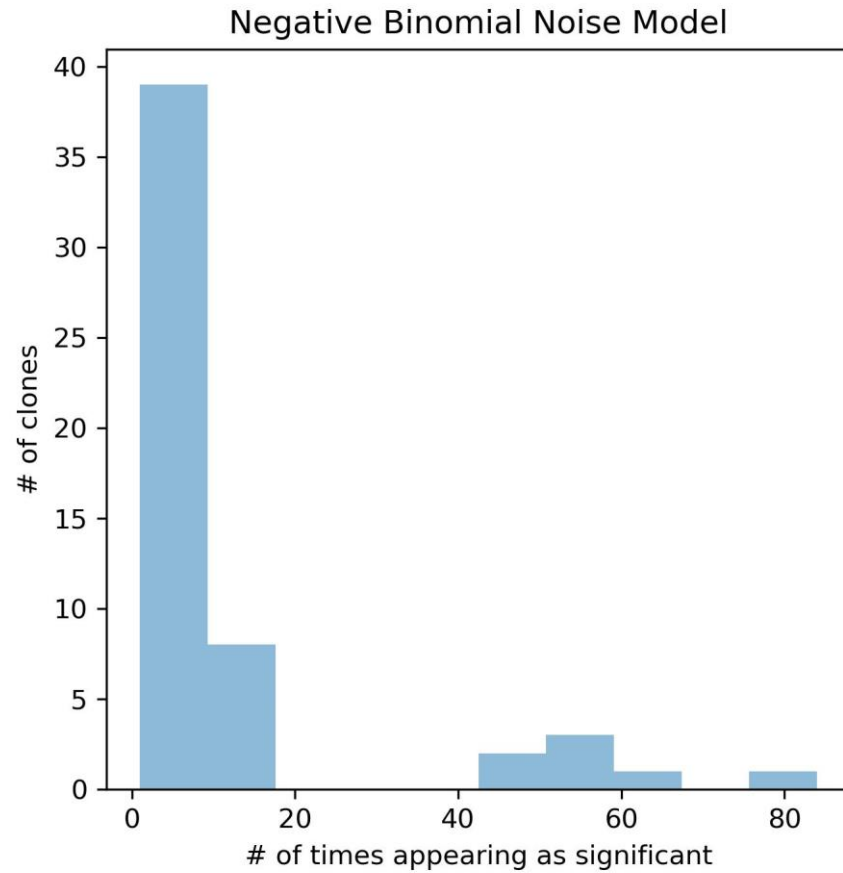
$$P((\hat{n}_i(t_1) = \hat{n}_1, \hat{n}_i(t_2) = n_2) | \gamma, \bar{s}) = \iint df_1 \rho(f_1) ds P(s | \gamma, \bar{s}) P(n_1 | f_1) P(n_2 | fe^s)$$

$$P(s | \hat{n}_1, \hat{n}_2) = \frac{P(\hat{n}_1, \hat{n}_2 | s, \gamma, \bar{s}) P(s | \gamma, \bar{s})}{P(\hat{n}_1, \hat{n}_2)}$$

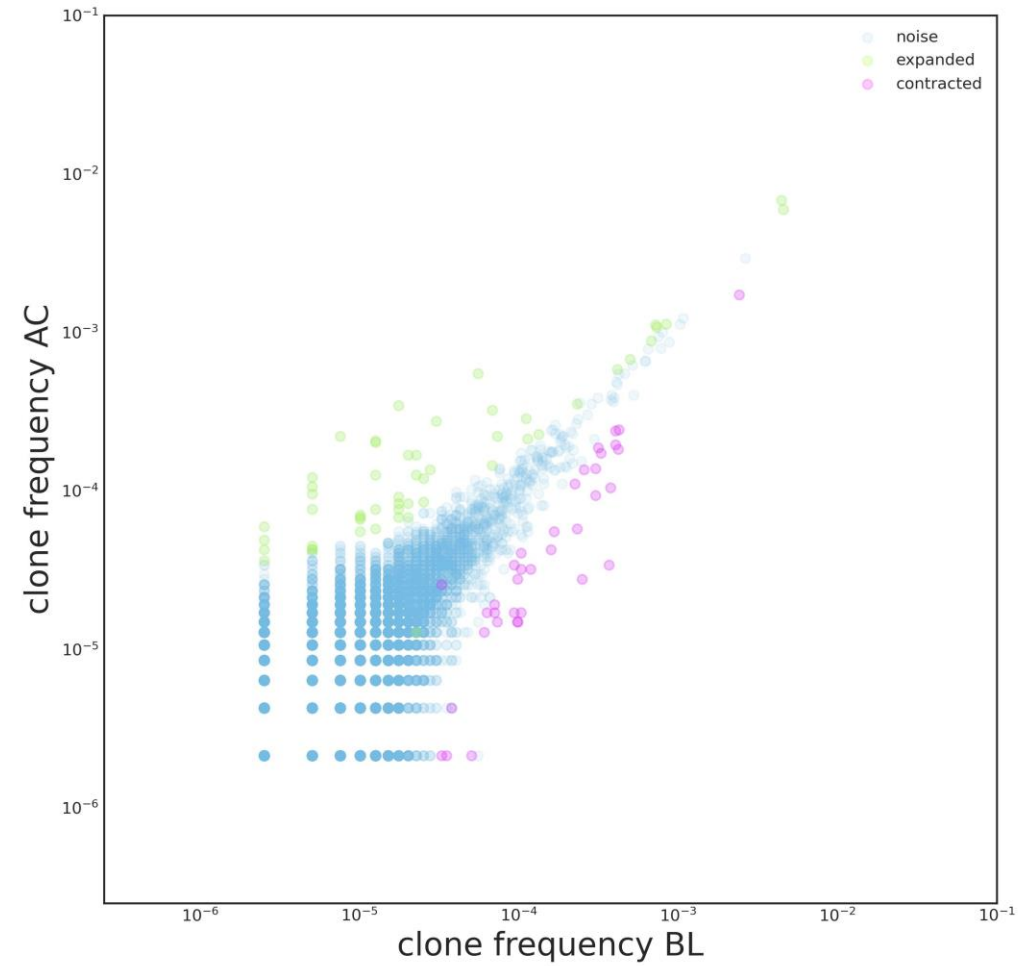
Algorithm



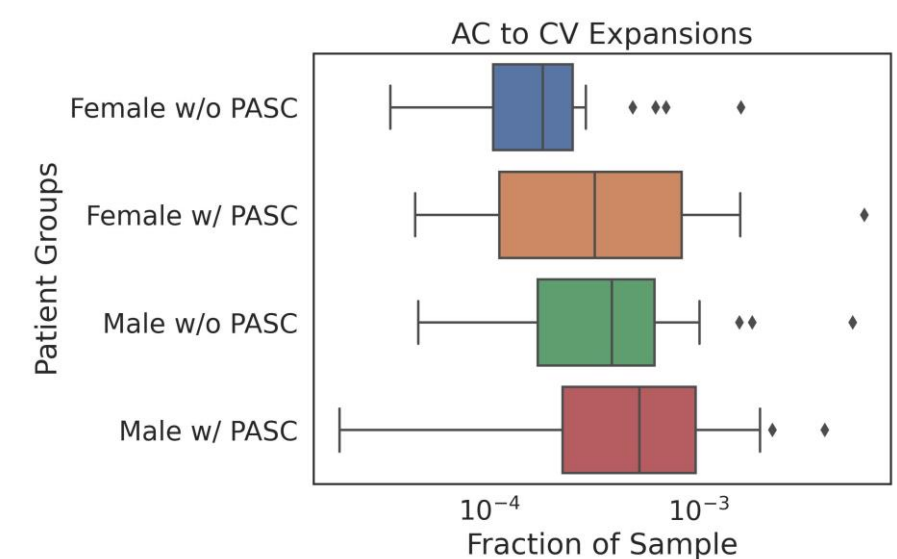
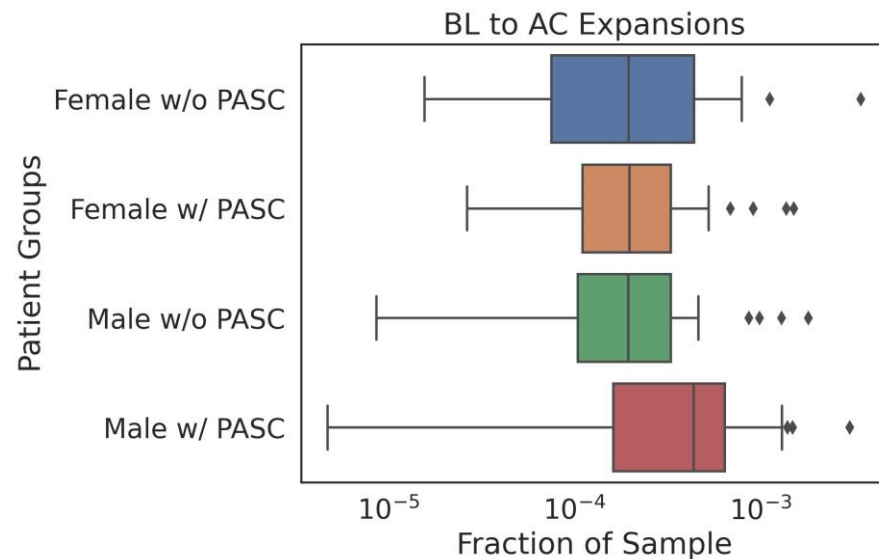
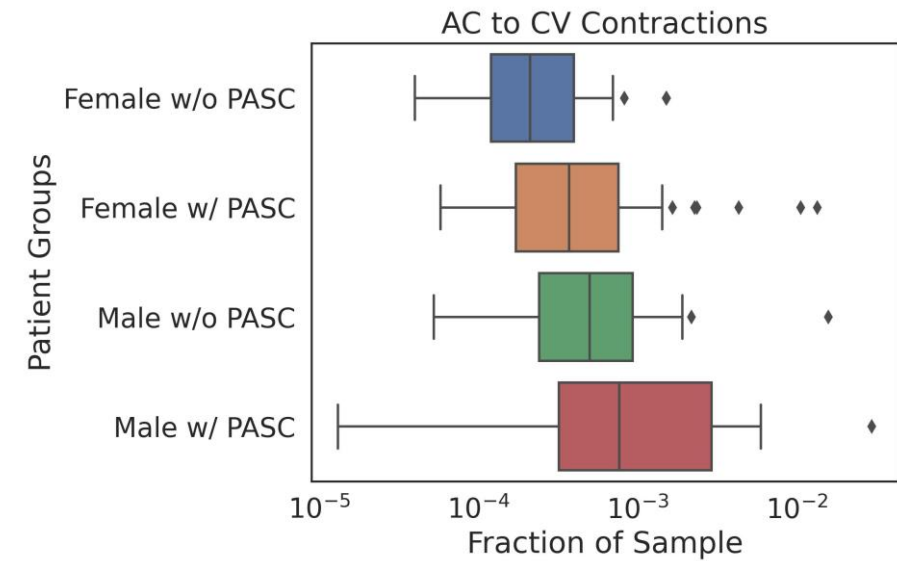
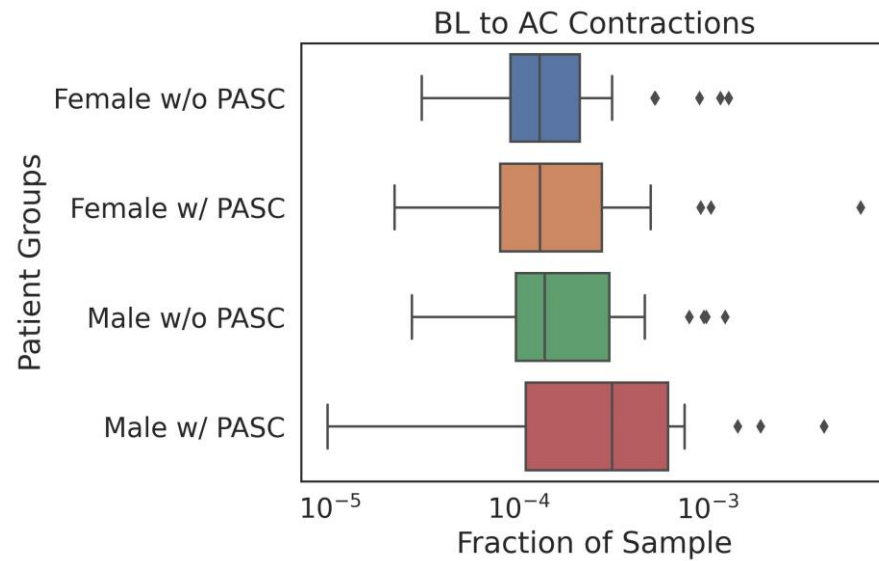
Choosing between noise models



Detecting Expansions and Contractions

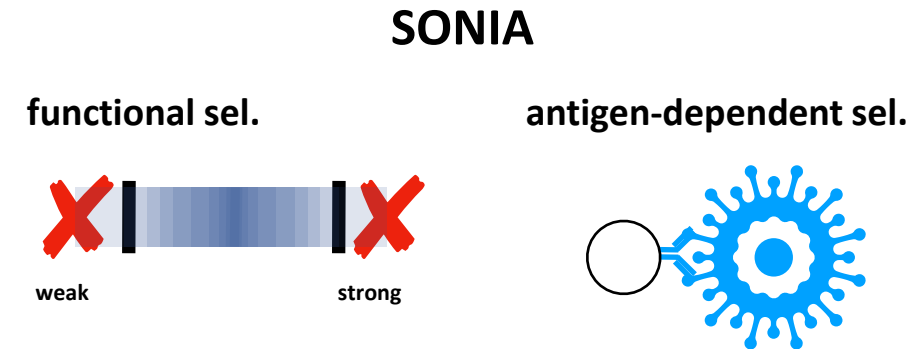
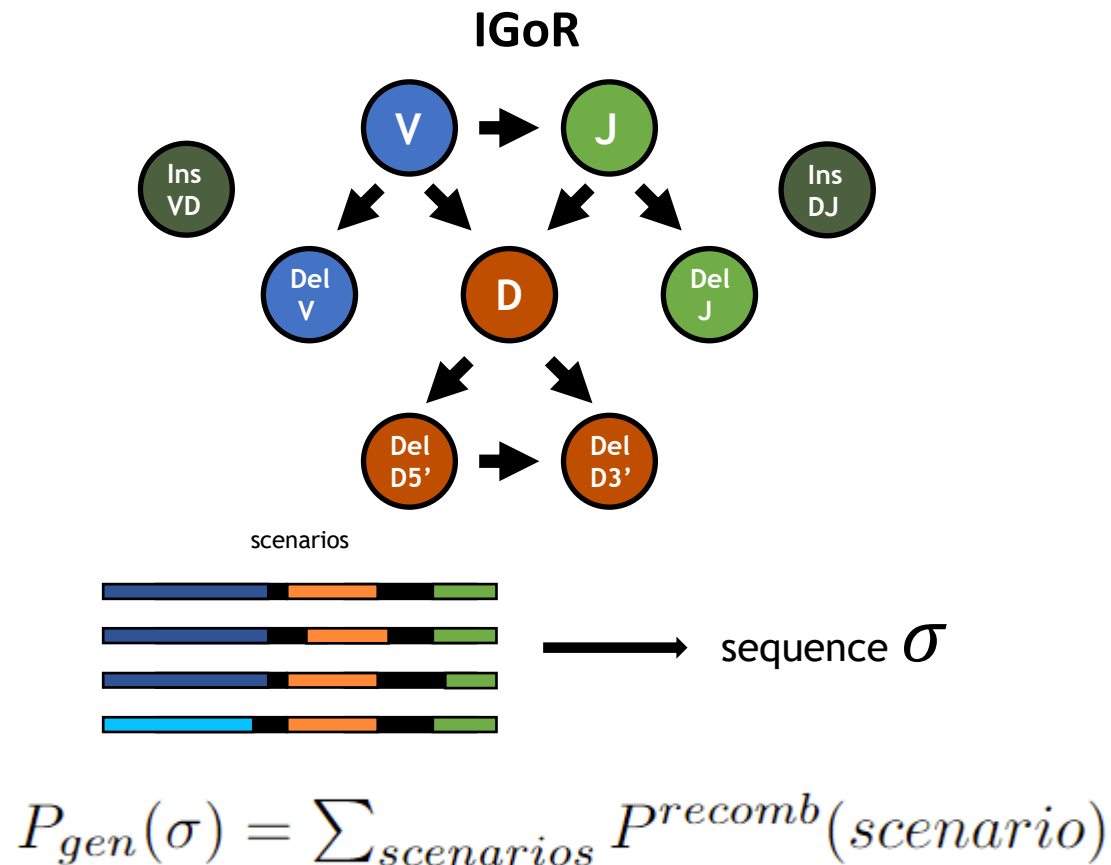


Fraction responding by demographic group

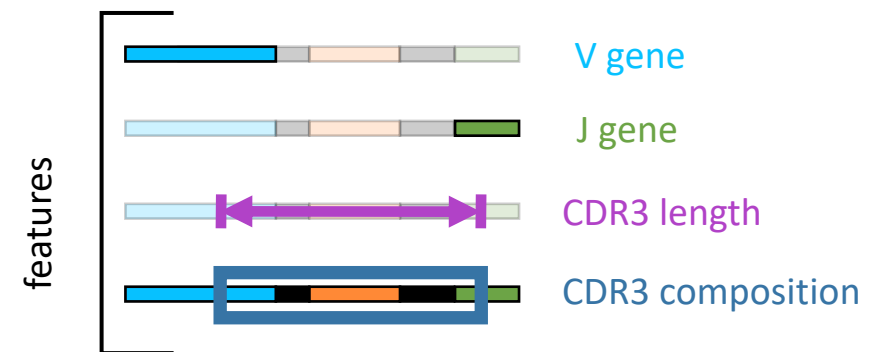


Generation and Selection Probability Analysis

Modeling generation and selection



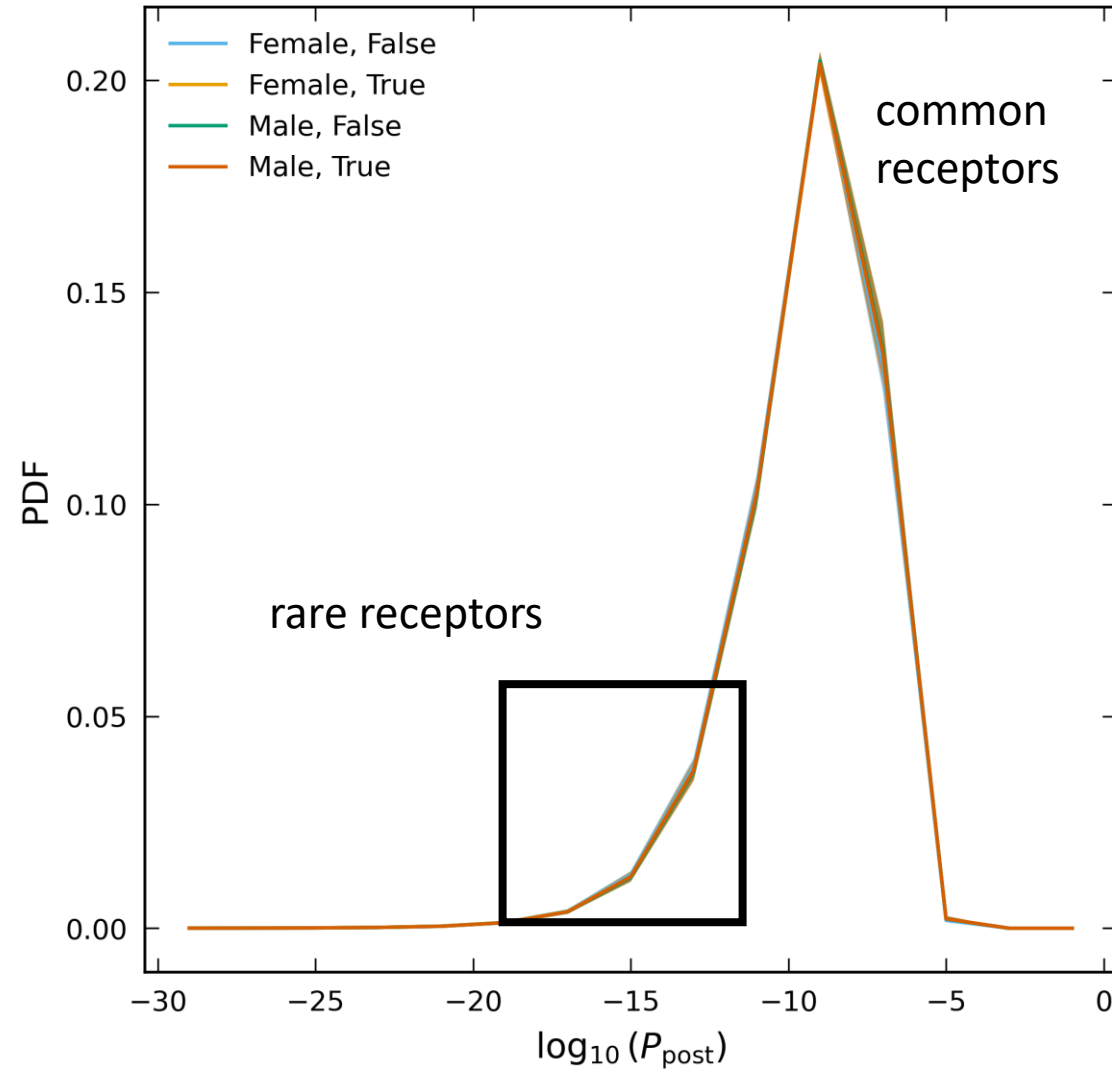
$$P_{post}(\sigma) = P_{gen}(\sigma) e^{\sum_f \lambda_f}$$



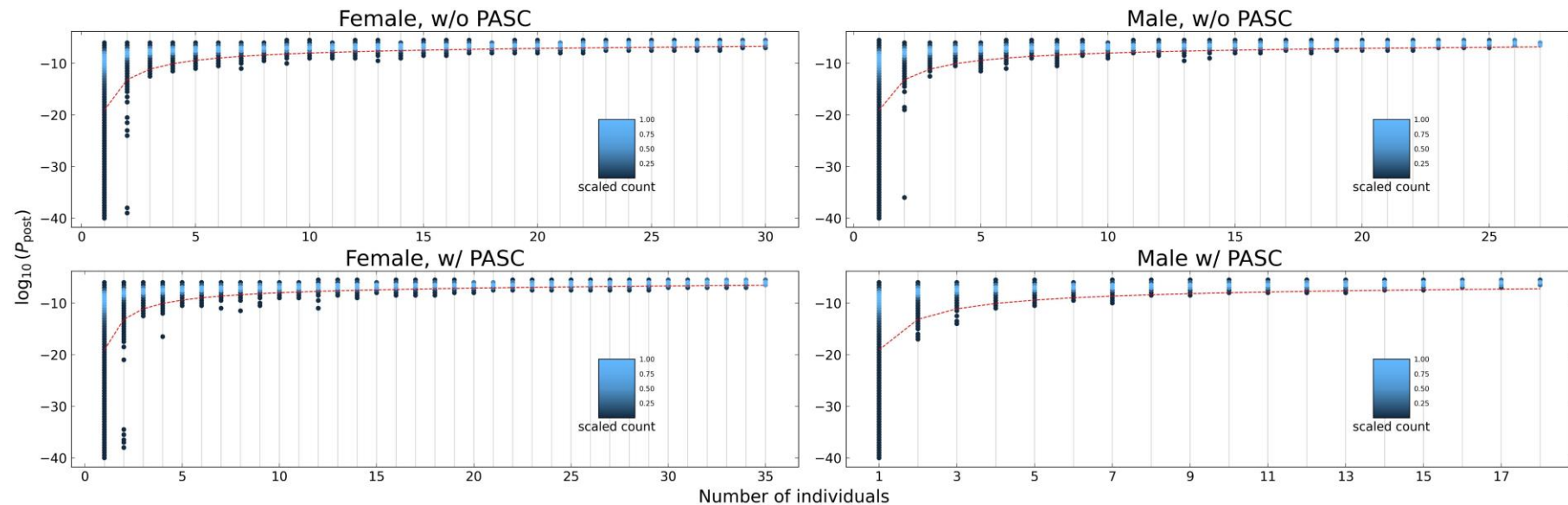
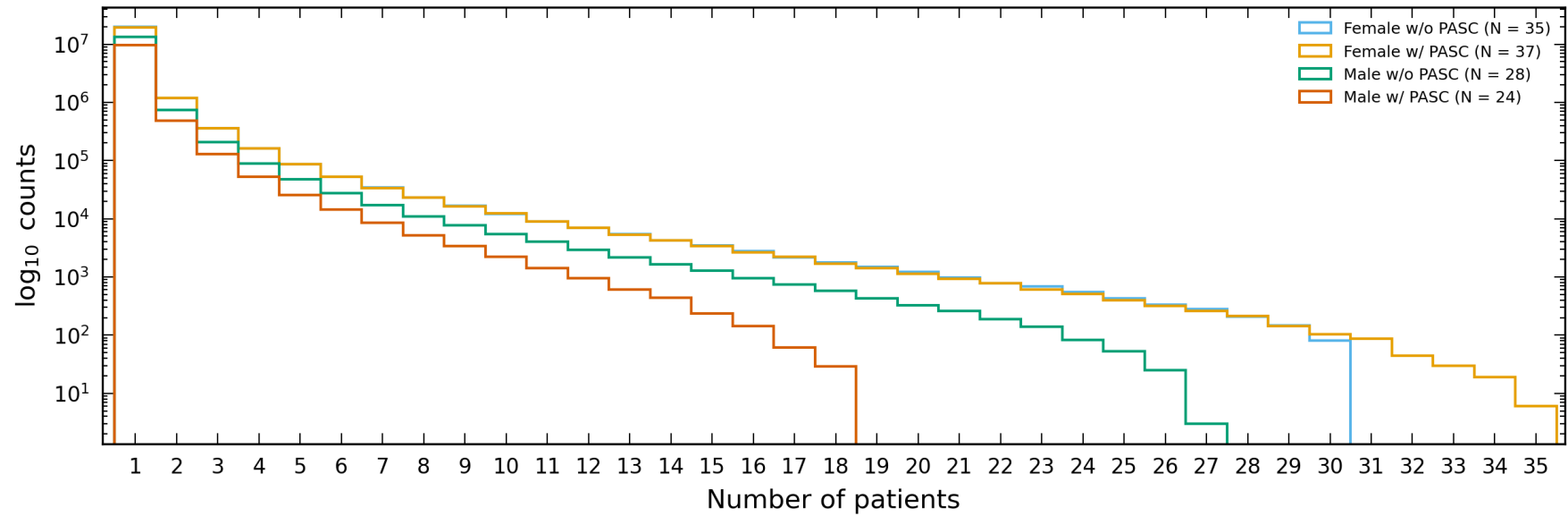
P_{post} distributions

Why might individuals share clones?

- Convergent recombination
- Experimental biases
- Exposure history



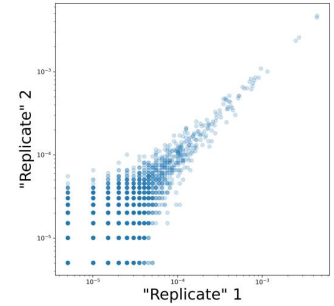
Clone sharing and P_{post} in shared clones



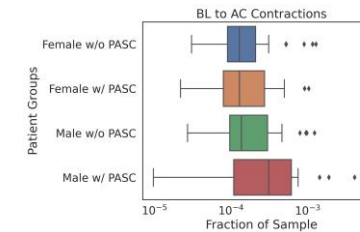
Discussion

Project Summary

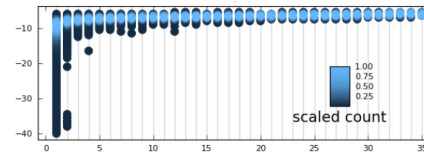
- Developed data thinning method for expansion analysis



- Detected expanding and contracting clones



- Identified rare, shared clones



Future Outlook

- Compare receptor distributions using information theoretic measures
- Refine data thinning technique
- Network analysis of responding clone sequences
- Integrate analysis with electronic health record data

Acknowledgements

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