

# Predicting breast cancer metastasis from blood samples

“On variance and other problems”

Einar Holsbø  
January, 2017

Q: can we predict metastasis from  
gene expression measurements in  
blood samples?

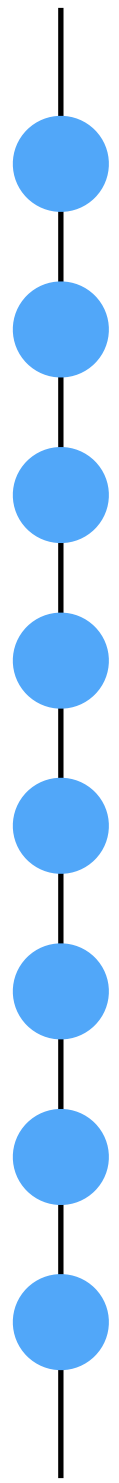
A: maybe

# Norwegian Women and Cancer (NOWAC)

- Prospective case–control study
- Blood samples + questionnaires

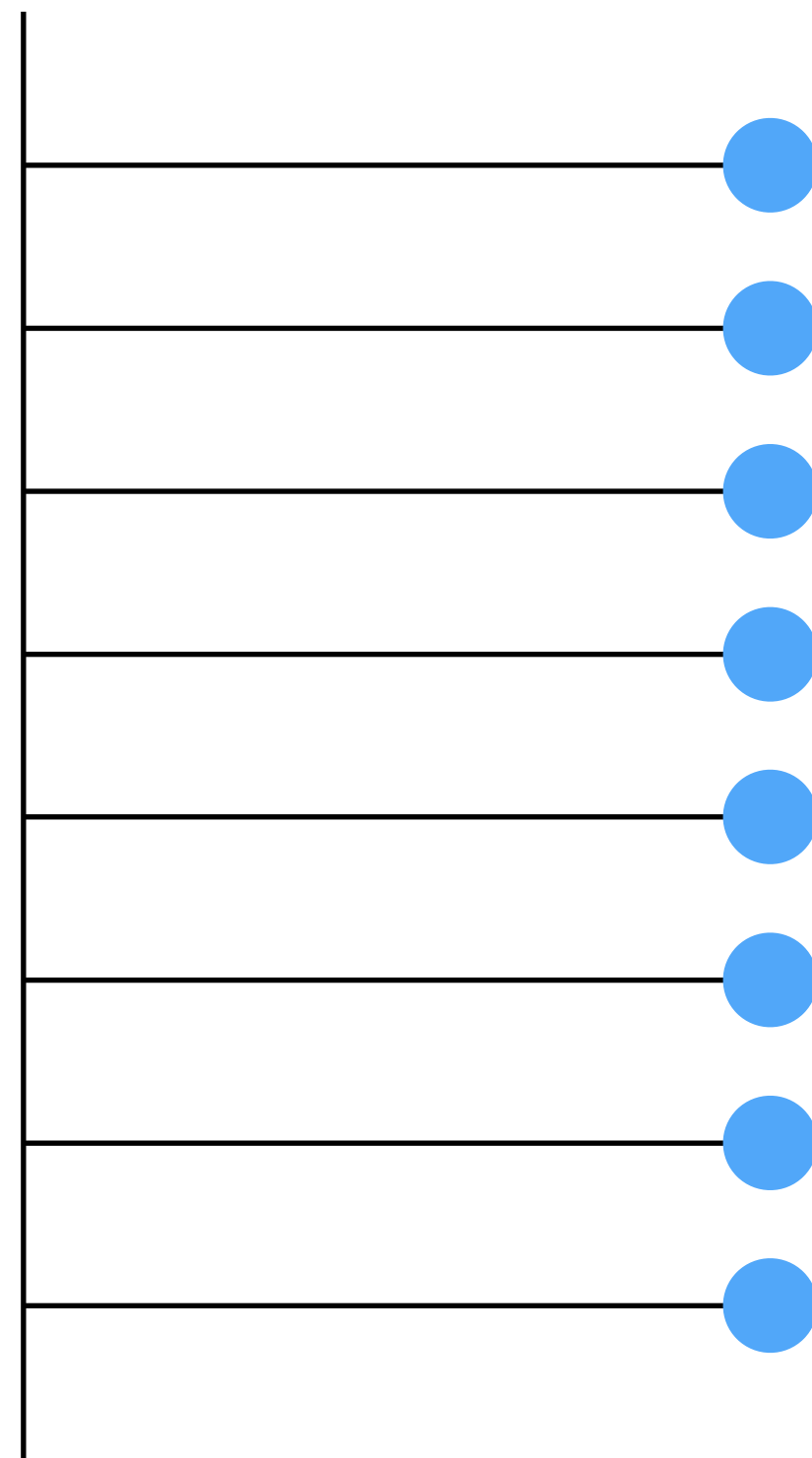
# Prospective

Enrollment



# Prospective

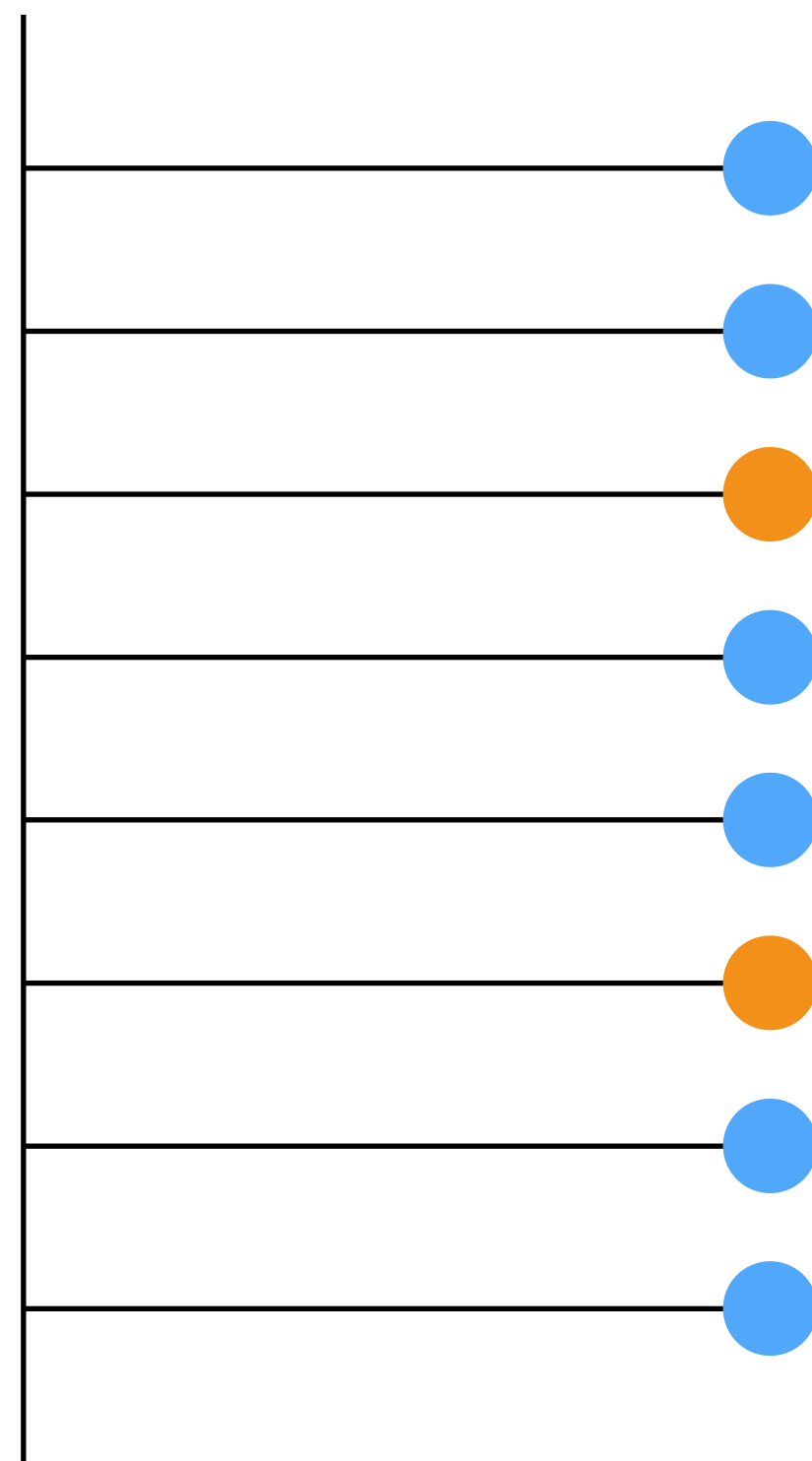
Enrollment



Time →

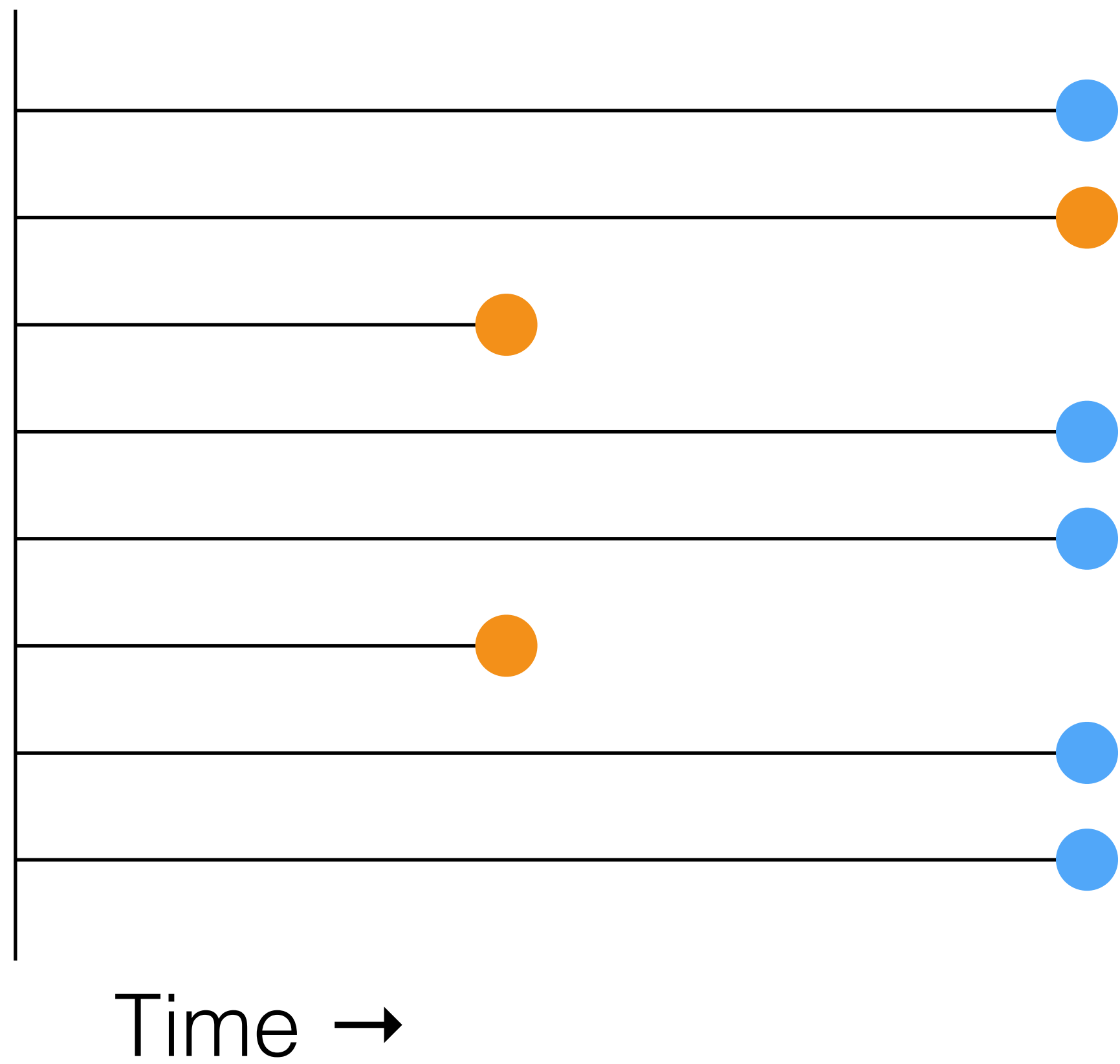
# Prospective

Enrollment



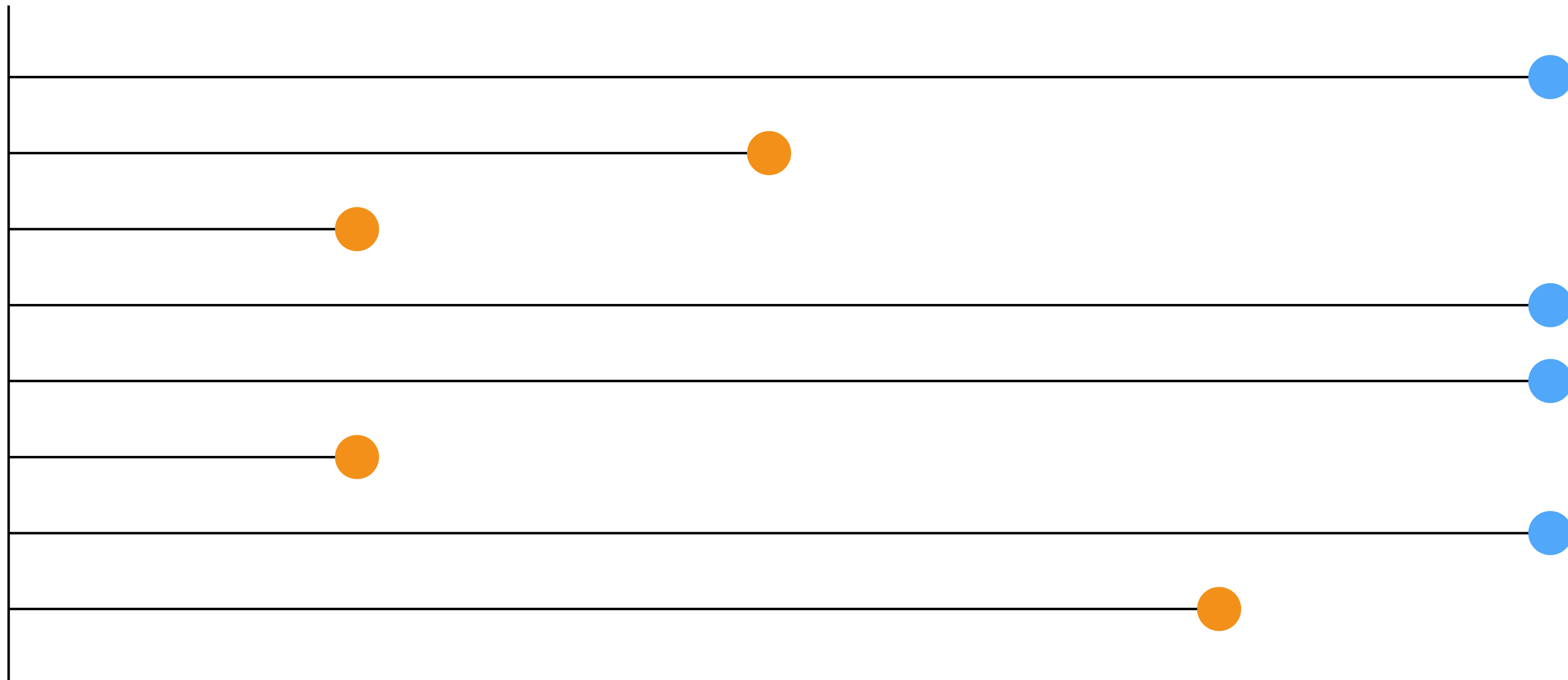
Time →

# Prospective

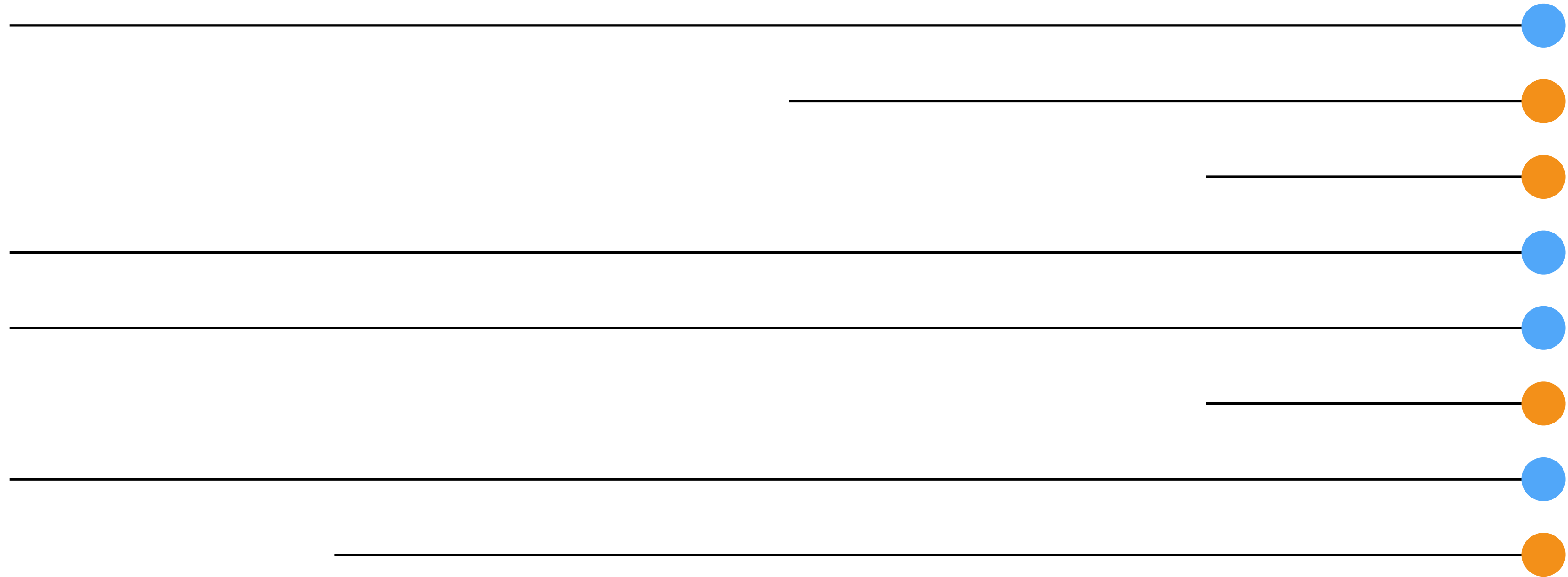




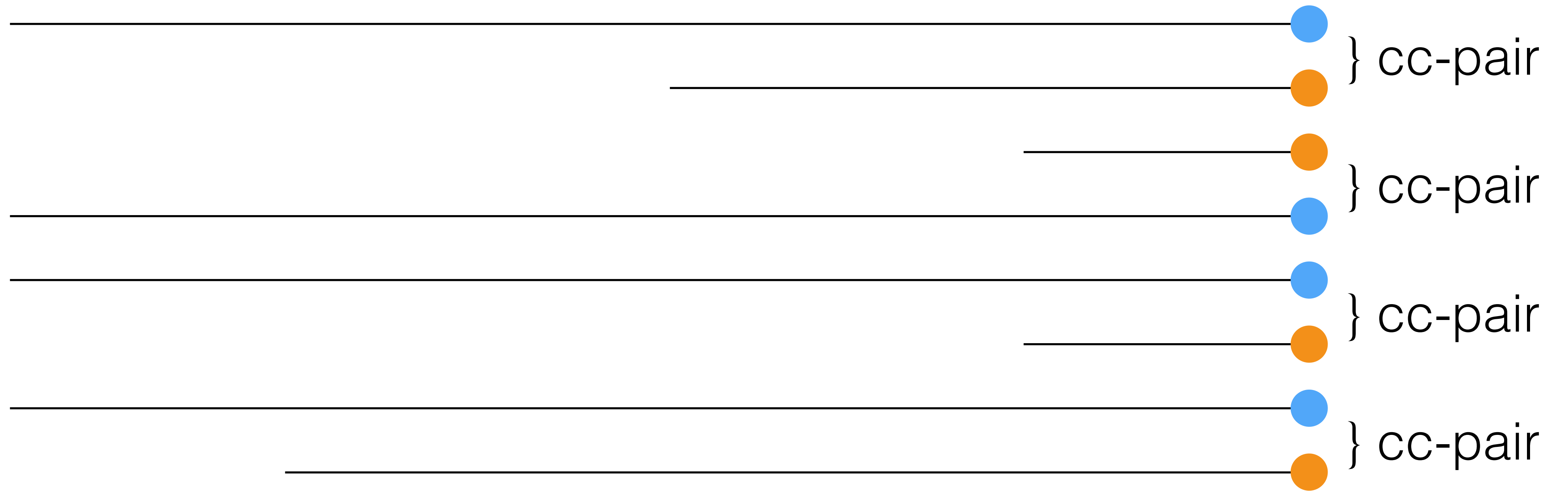
# Prospective



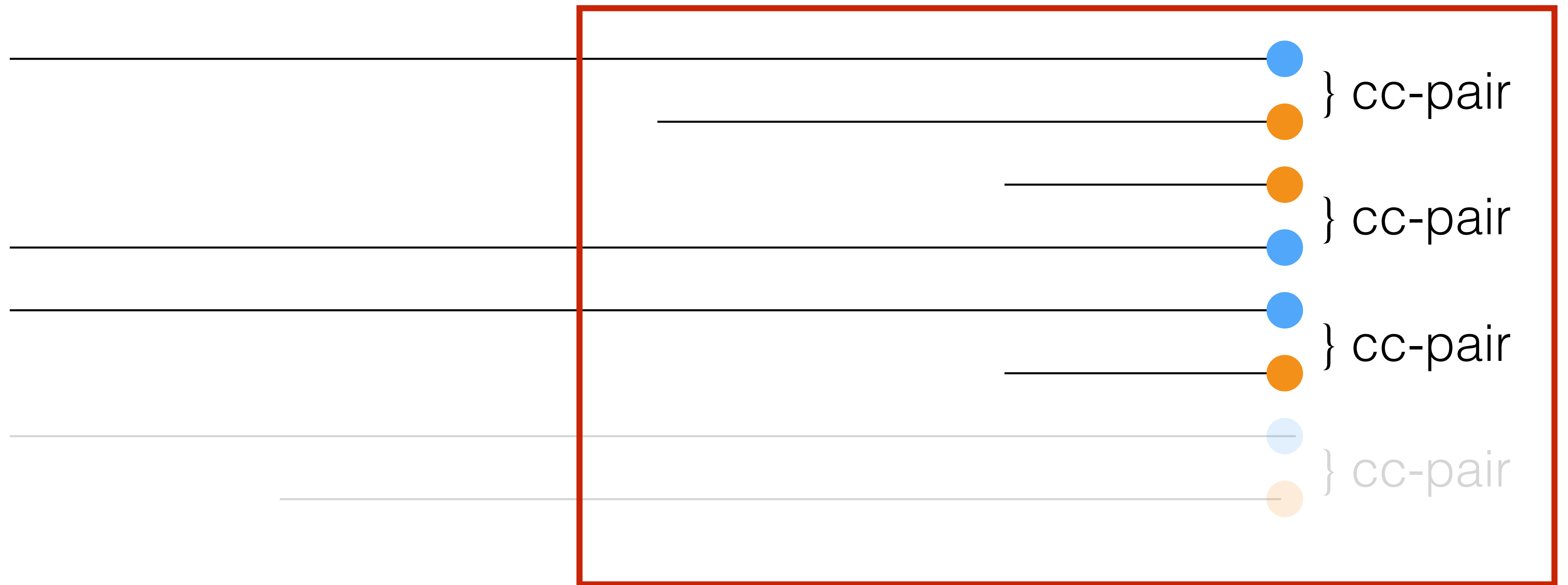
# Prospective



# Case-control



# Case-control



**1 year before diagnosis**

# Data at a glance

```
dim(gene_expression)
## [1] 88 12404
```

```
summary(days_to_diagnosis)
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      6.0   117.8   189.5   186.8   269.2   358.0
```

```
summary(metastasis)
## FALSE  TRUE
##    66    22
```

```
table(metastasis, stratum)
##           stratum
## metastasis screening interval clinical
##      FALSE           43           10           13
##      TRUE            6            6            10
```

# How to do predictive modelling

1. Pick some of your favorite models
2. Evaluate model performance by cross-validation
3. Fit tuning parameters by nested cross-validation

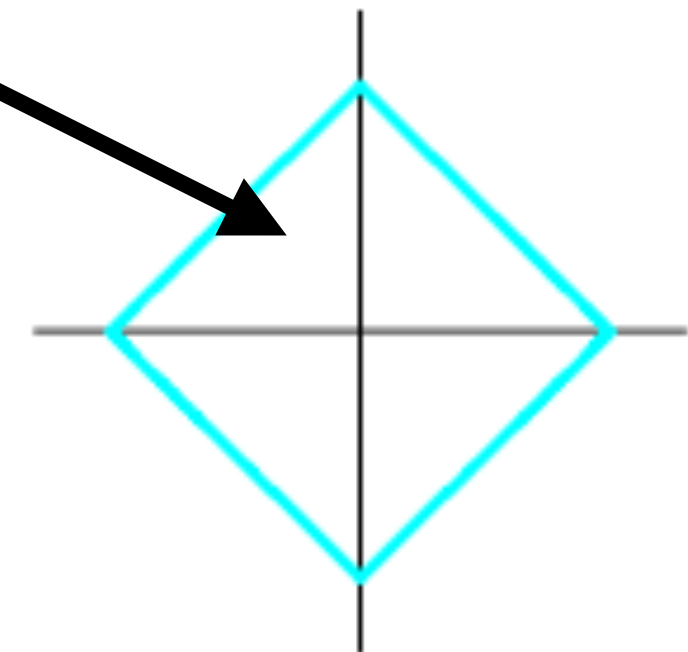
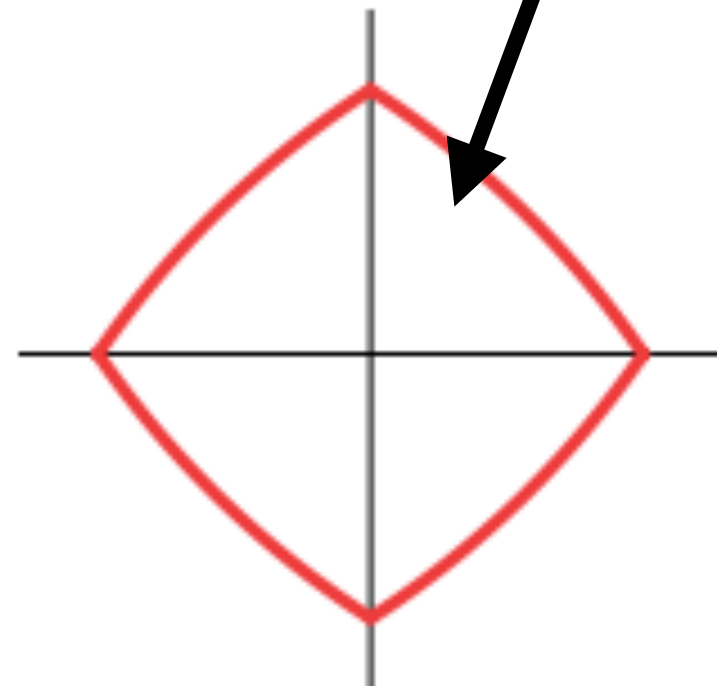
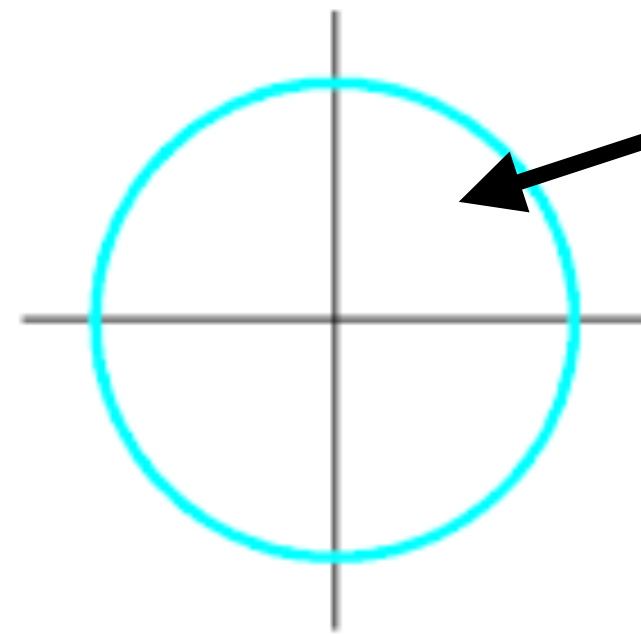
# Some models

Penalized logistic regression

$$\text{find } \hat{\beta} \text{ s.t. } \log \frac{p(Y|x)}{1 - p(Y|x)} = \hat{\beta}_0 + \hat{\beta}_1 x_1 + \dots$$

# Some models

$\hat{\beta}$  can only be inside these shapes

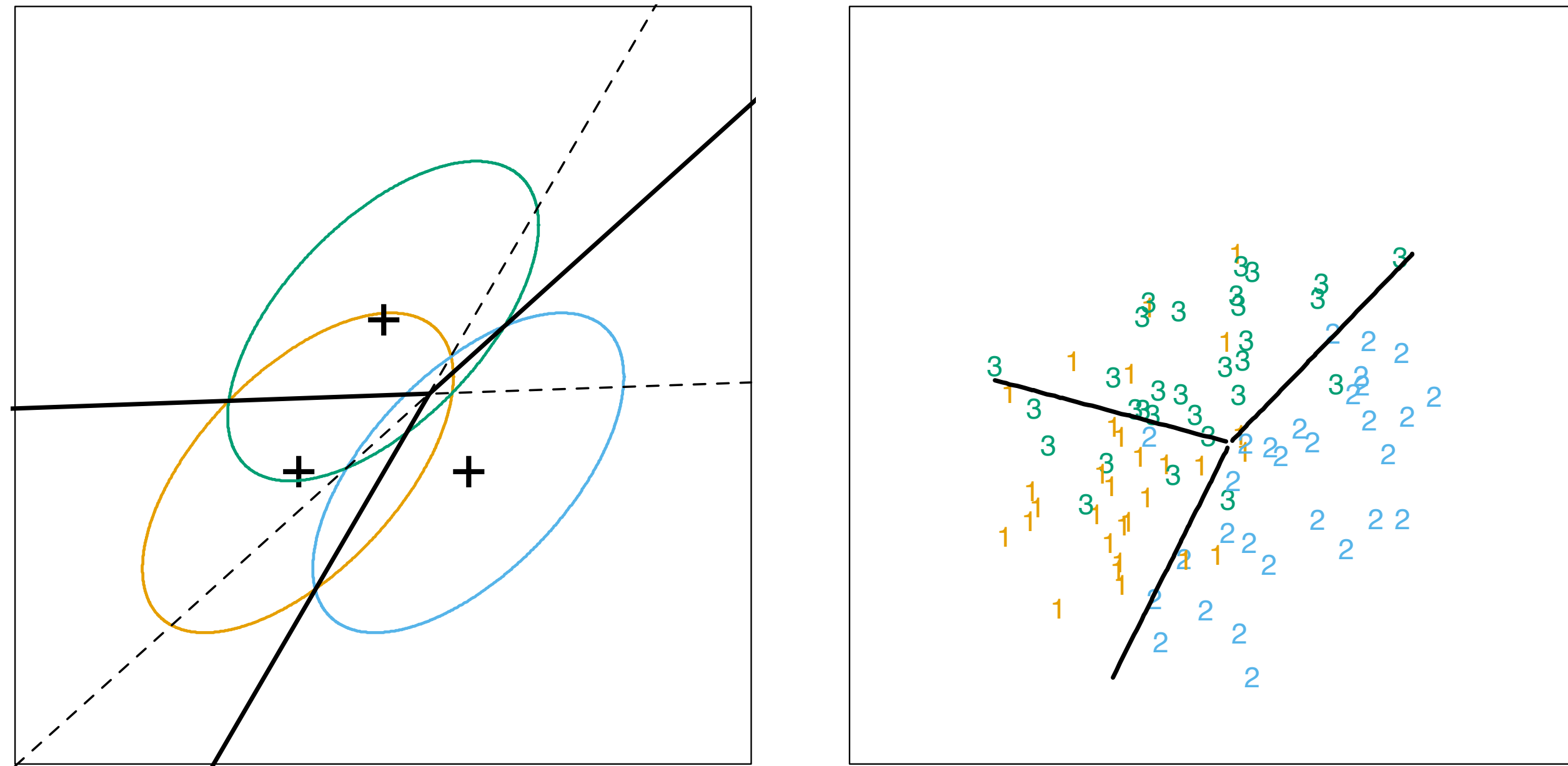


From Hastie, Tibshirani, and Friedman: The Elements of Statistical Learning



# Some models

## Nearest centroids



From Hastie, Tibshirani, and Friedman: The Elements of Statistical Learning

# Cross validation



# Cross validation



# Cross validation

Fit model ->

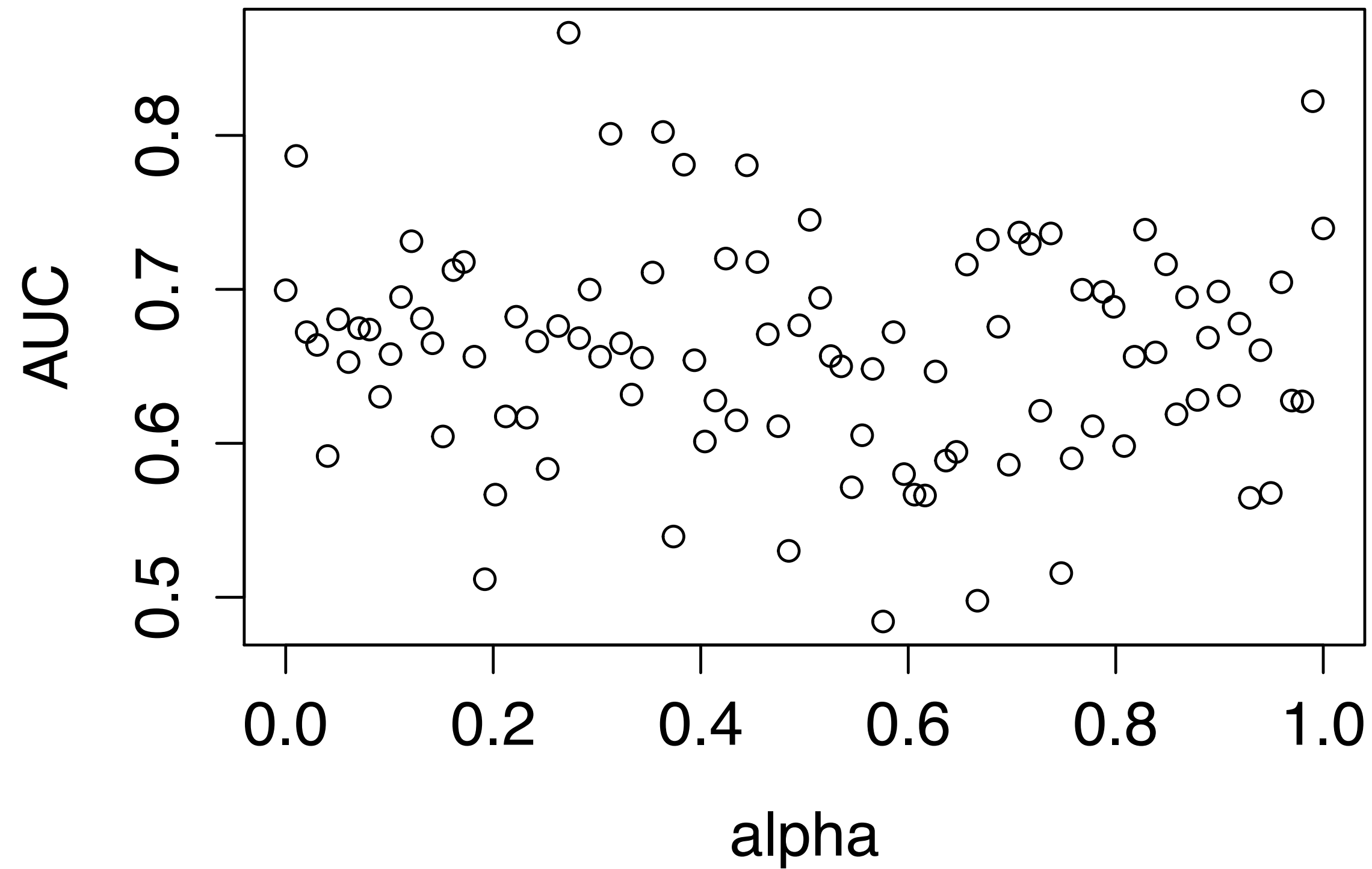


Evaluate ->



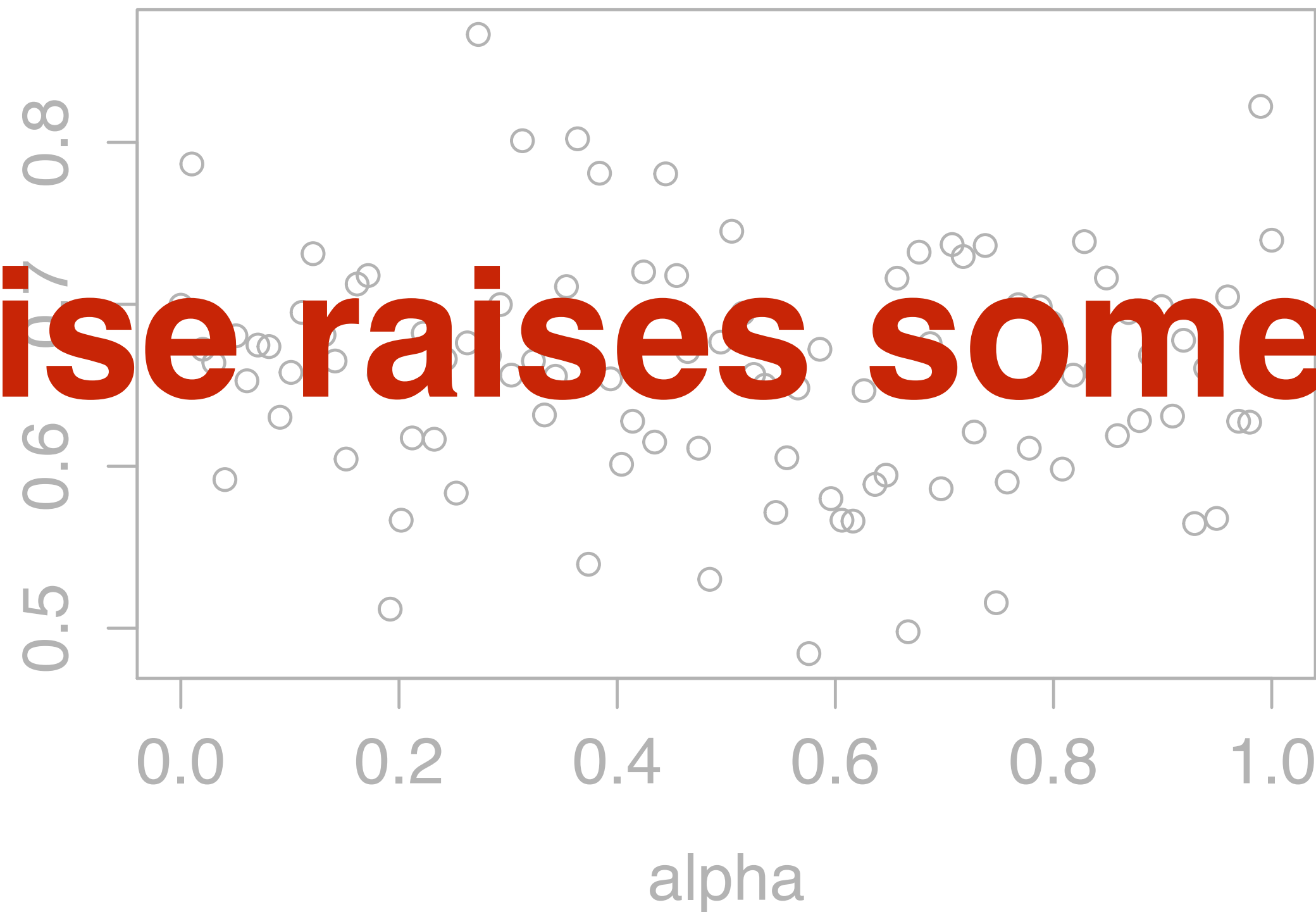
# Fit tuning parameters????????

Finding the “best” parameter alpha by cross-validation



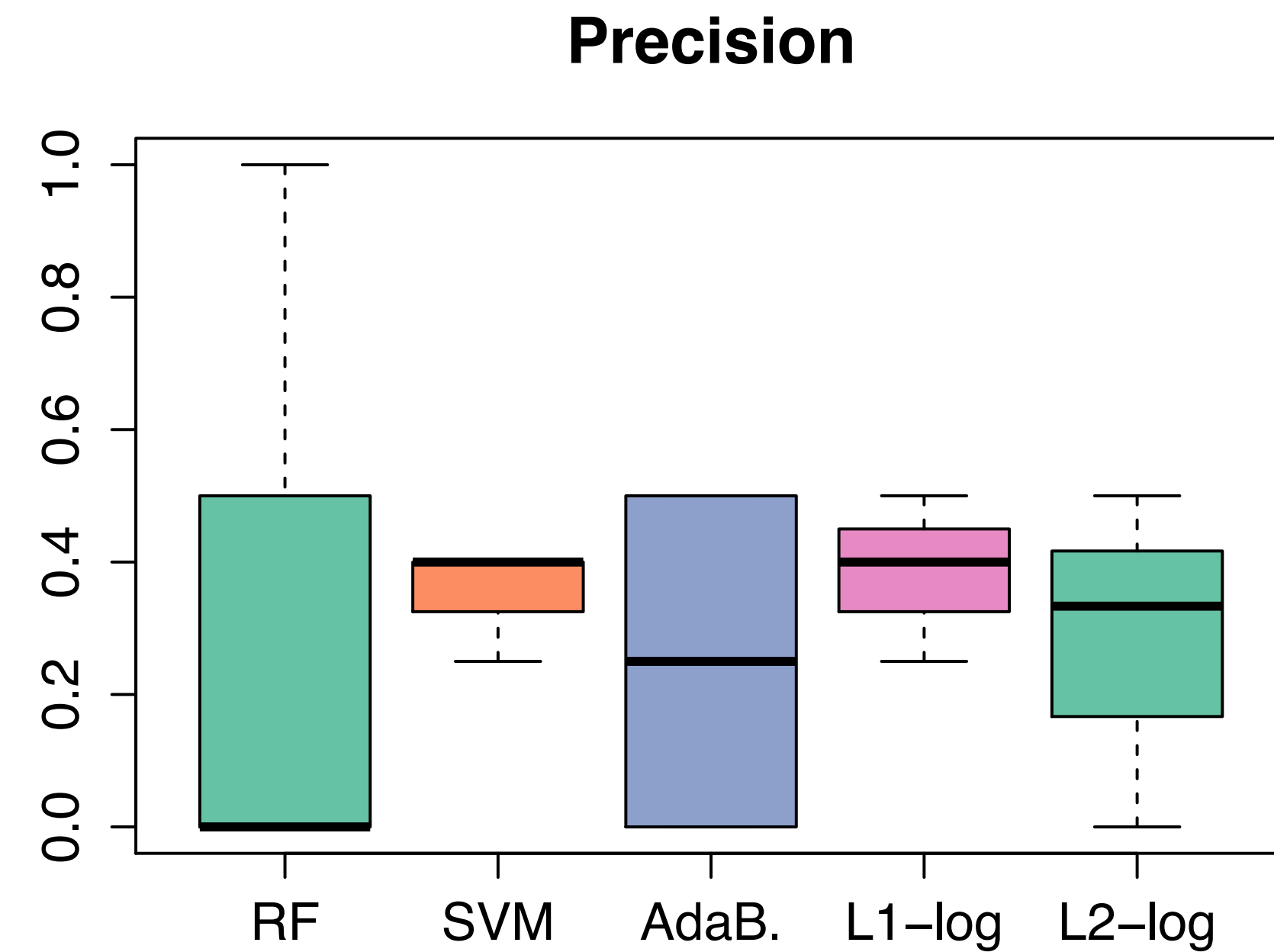
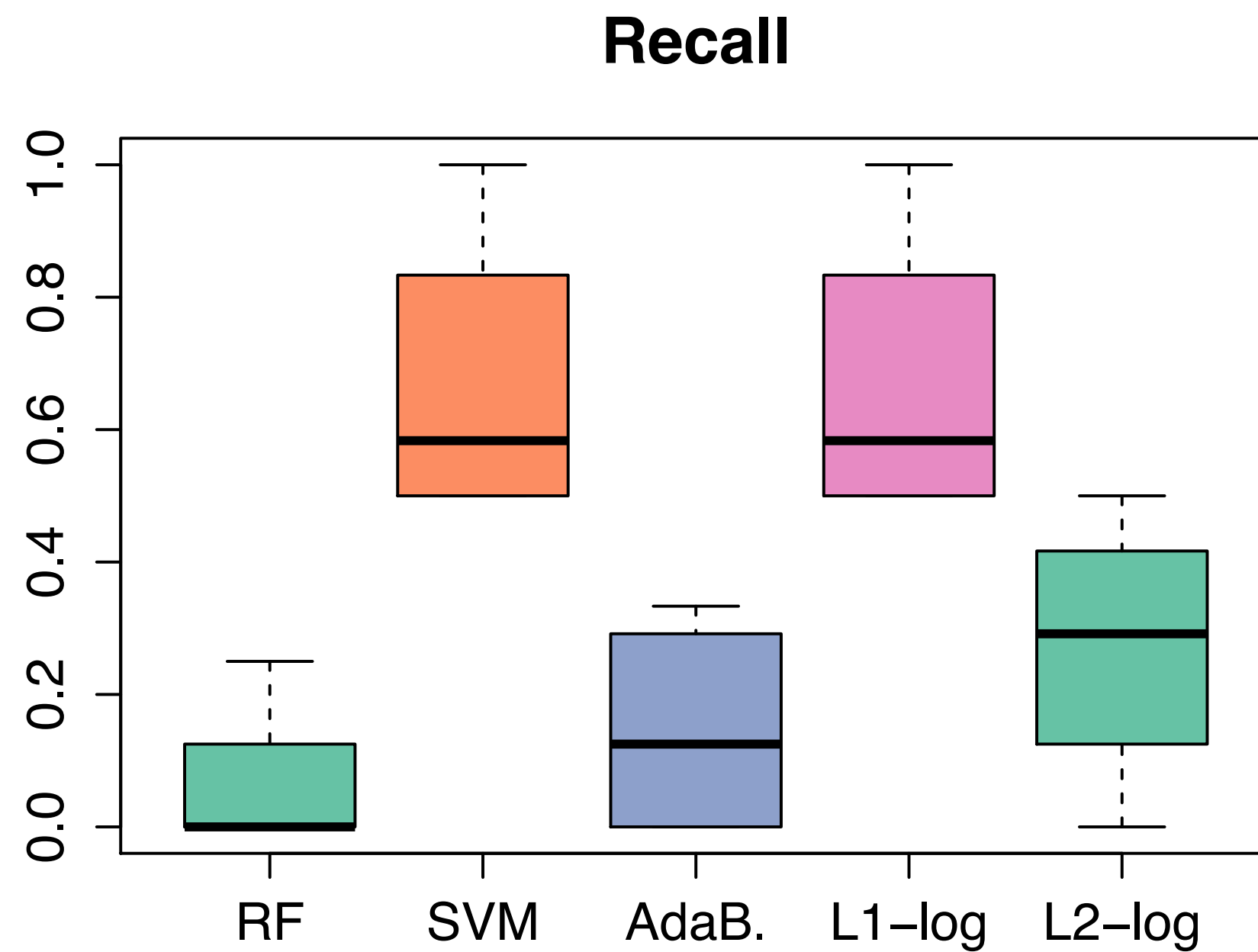
# Fit tuning parameters????????

Finding the “best” parameter alpha by cross-validation

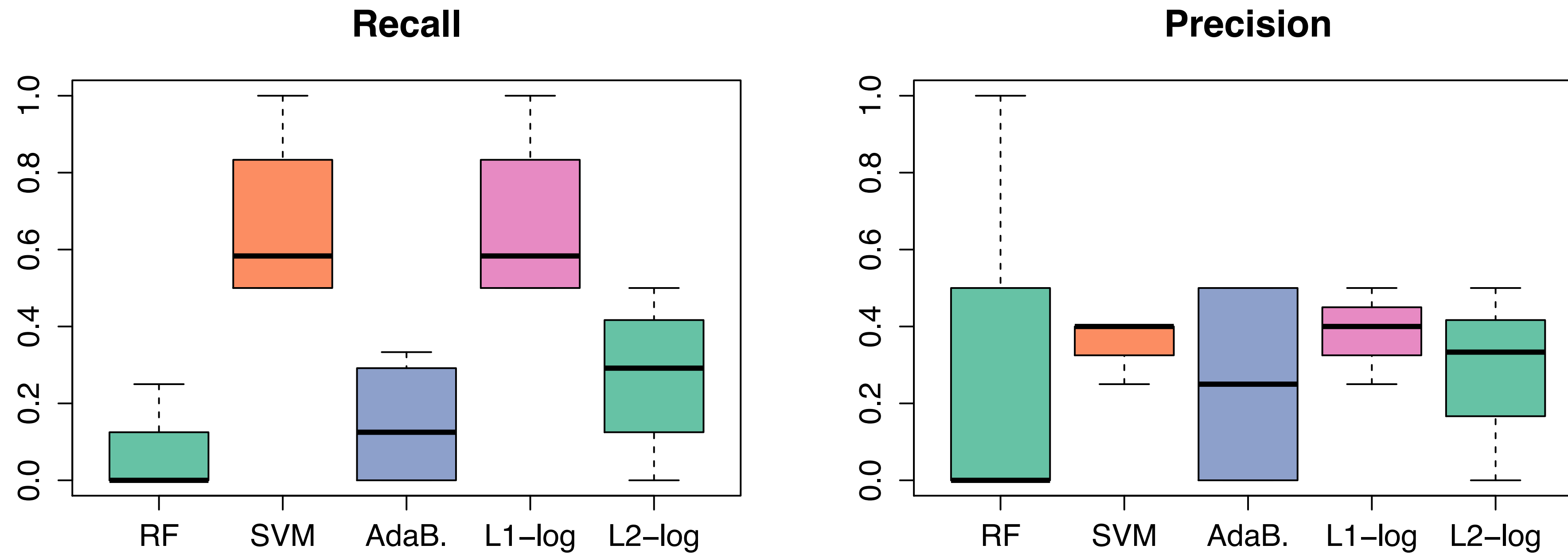


**This exercise raises some questions**

# Cross validation is almost useless to me



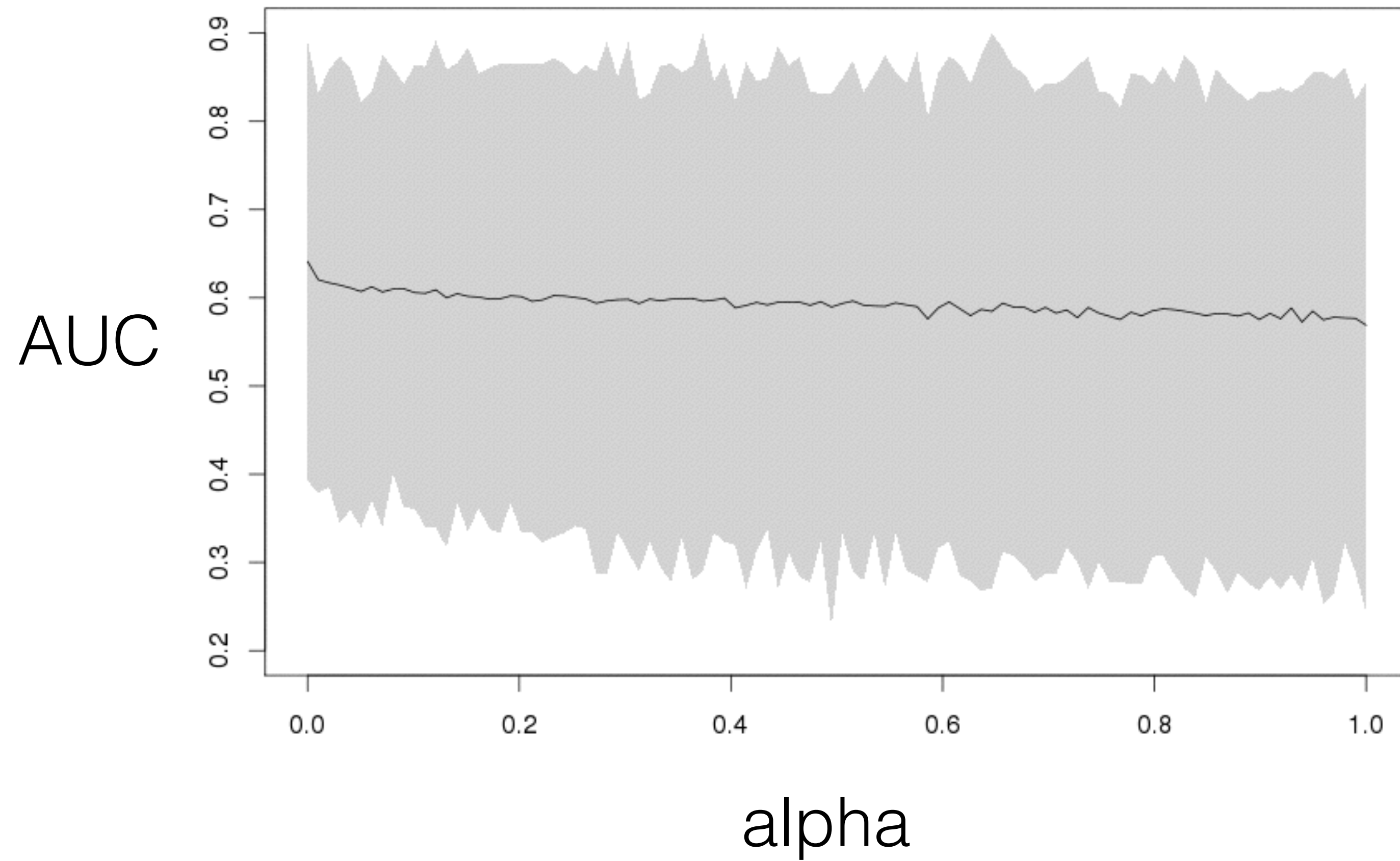
# Cross validation is almost useless to me



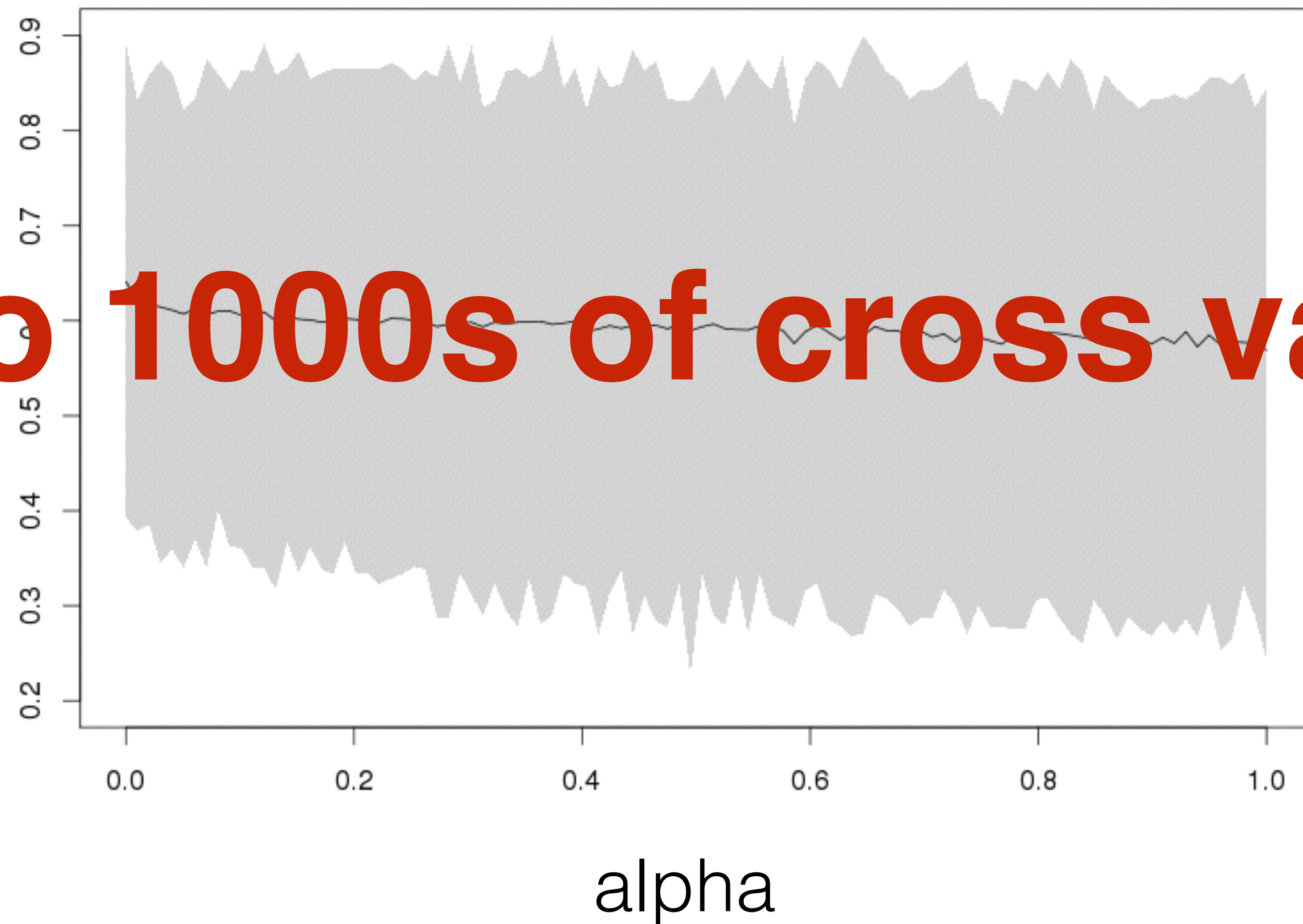
I spent actual time interpreting plots like these.....



# Solution: resampling

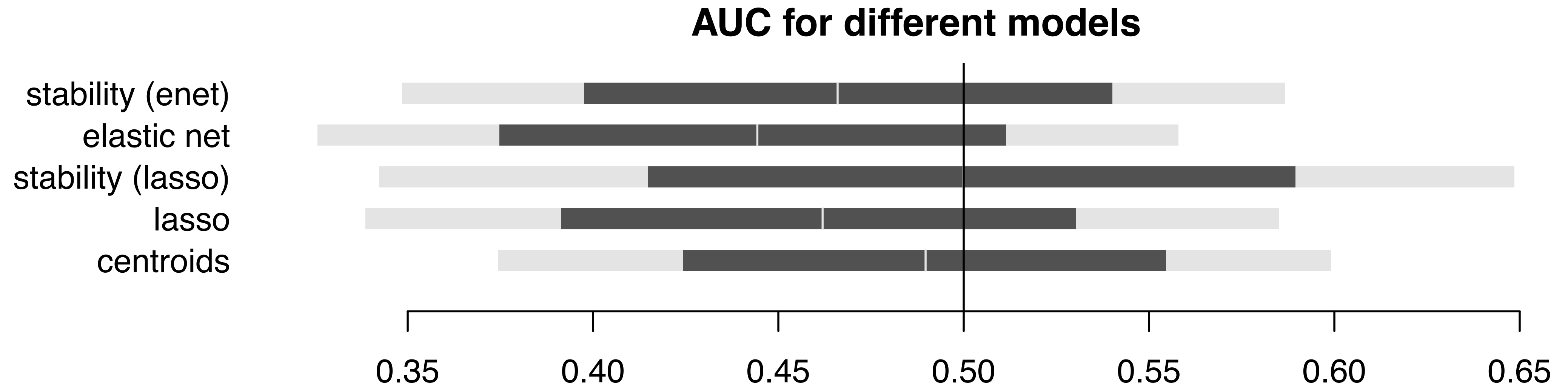


# Solution: resampling

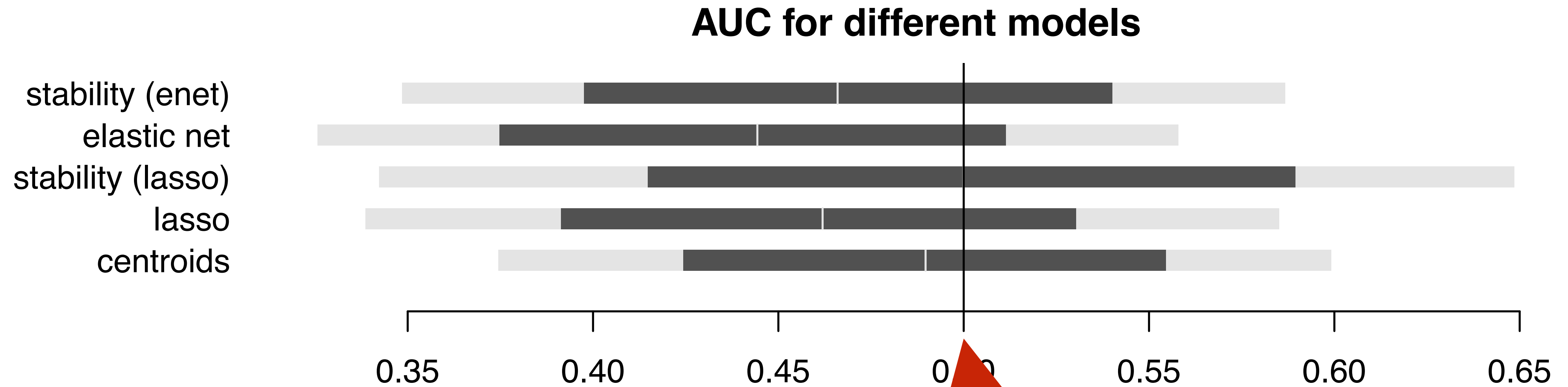


**Simply do 1000s of cross validations**

# Another confusing thing



# Another confusing thing



**The line for random guess**

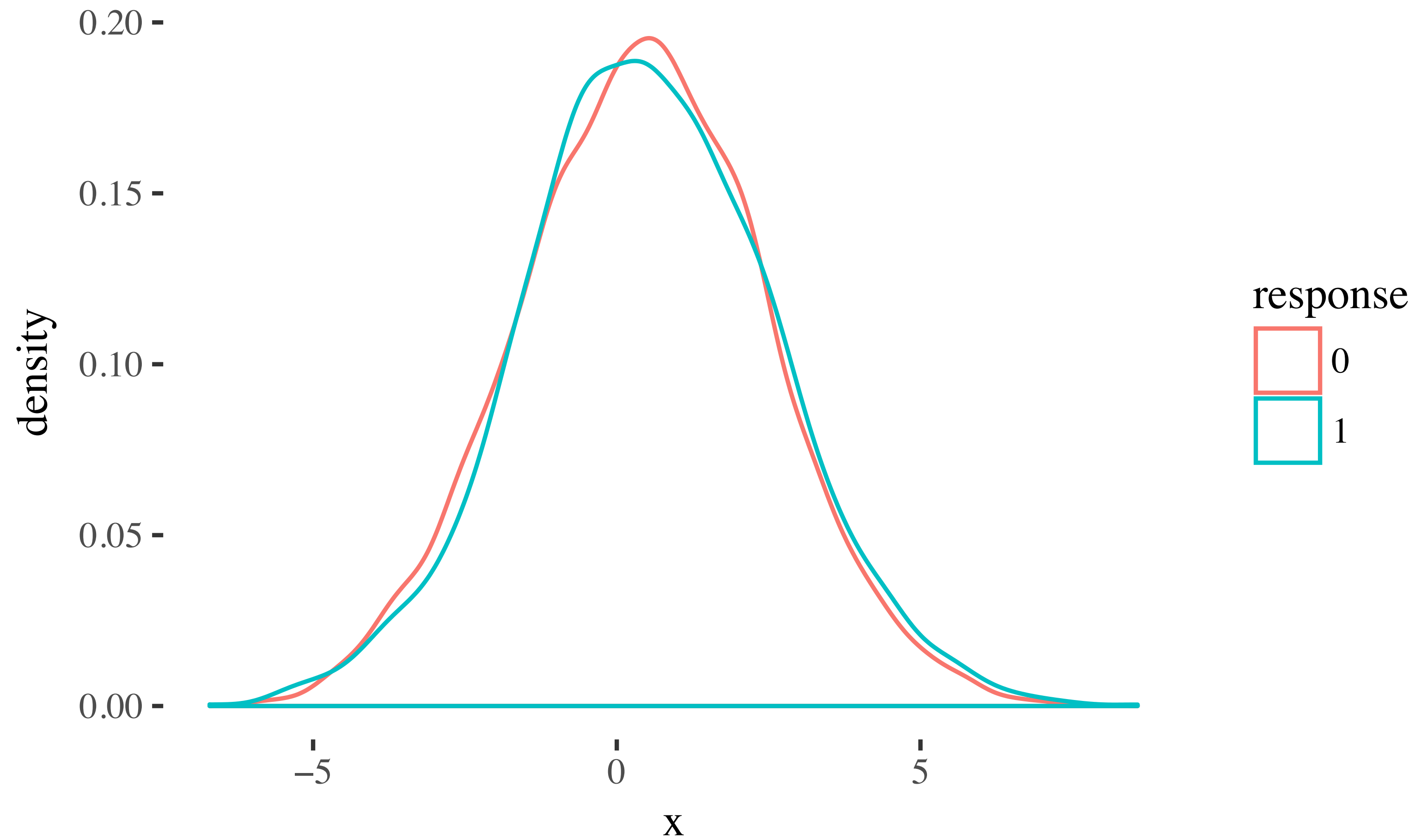
# 2 ways to get AUROC $< .5$

- A. You made a mistake calculating AUC
- B. There is something v. strange with the data

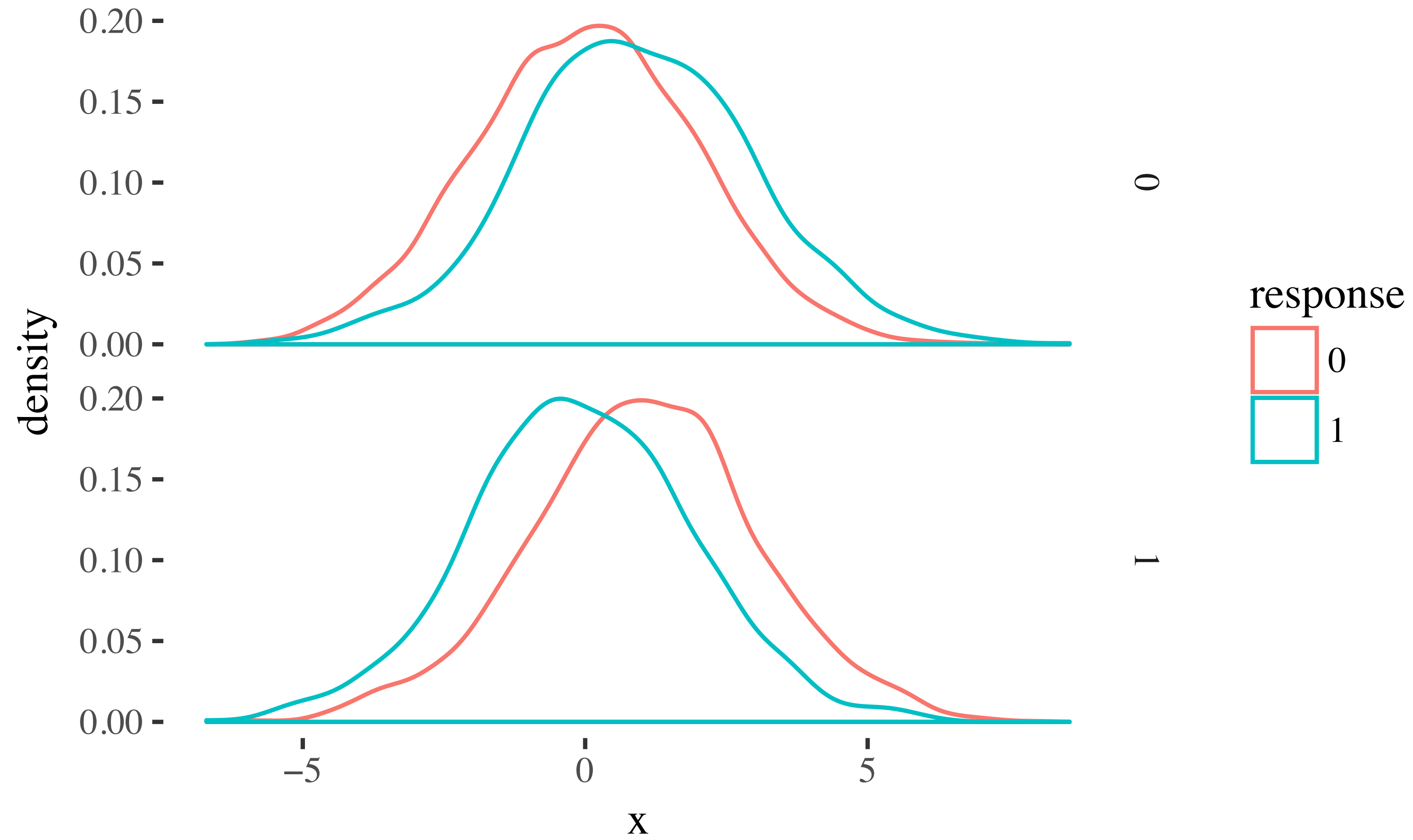
# A simulated paradox

- One “gene,”  $x$
- Response 1 or 0
- Two strata: 1 and 0
- If stratum == response,  $x \sim N(1, \text{variance})$
- Else,  $x \sim N(0, \text{variance})$

# A simulated paradox



# A simulated paradox

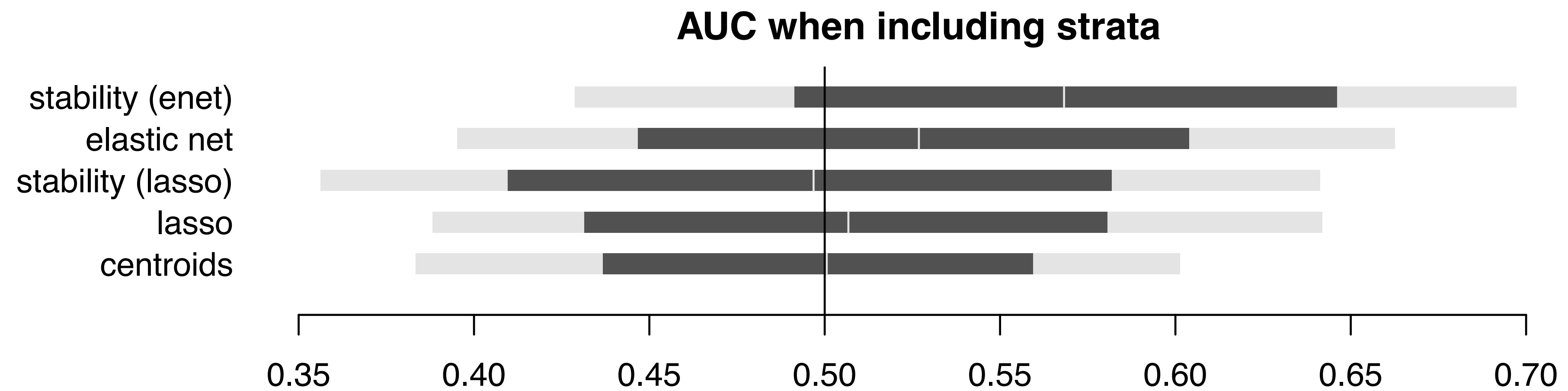




“You have to stratify.”

–Eiliv Lund to myself, like two-and-a-half years ago

# Including stratum gives expected null behavior

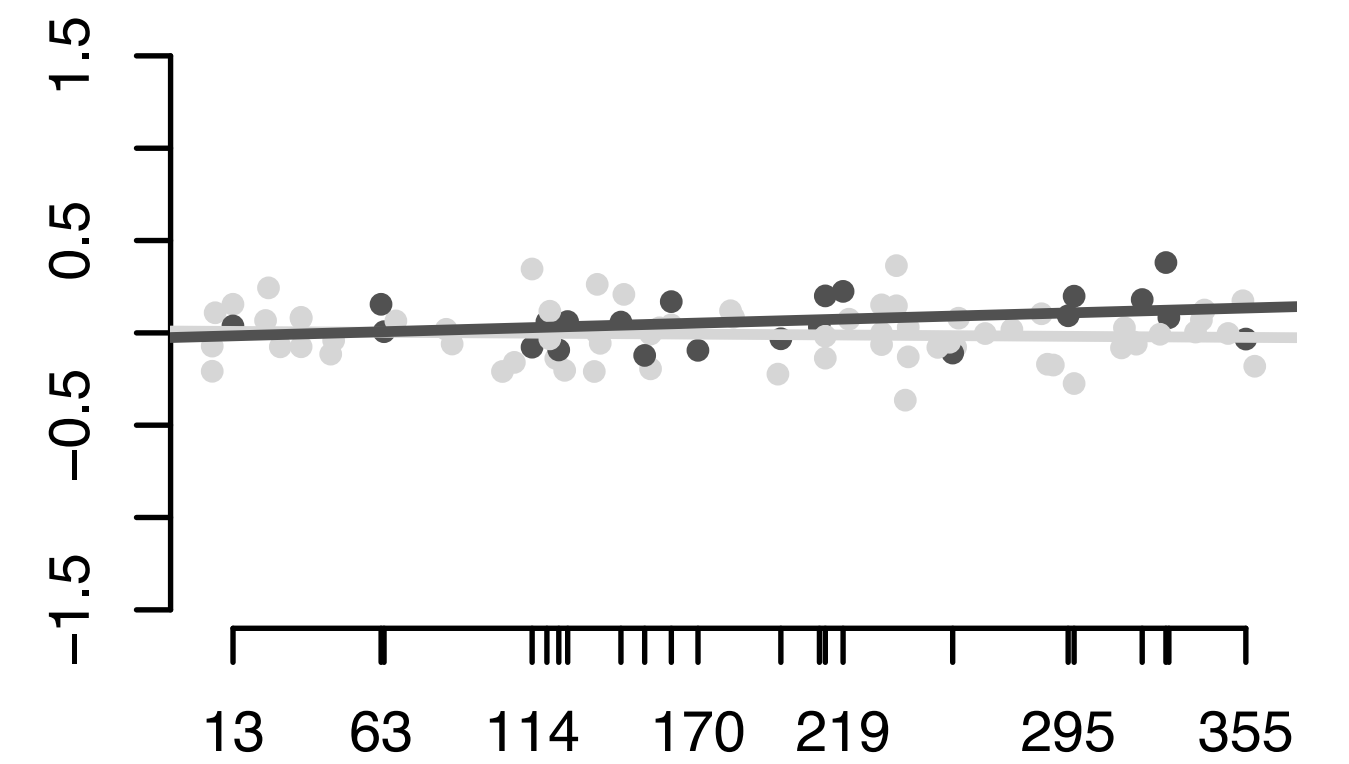
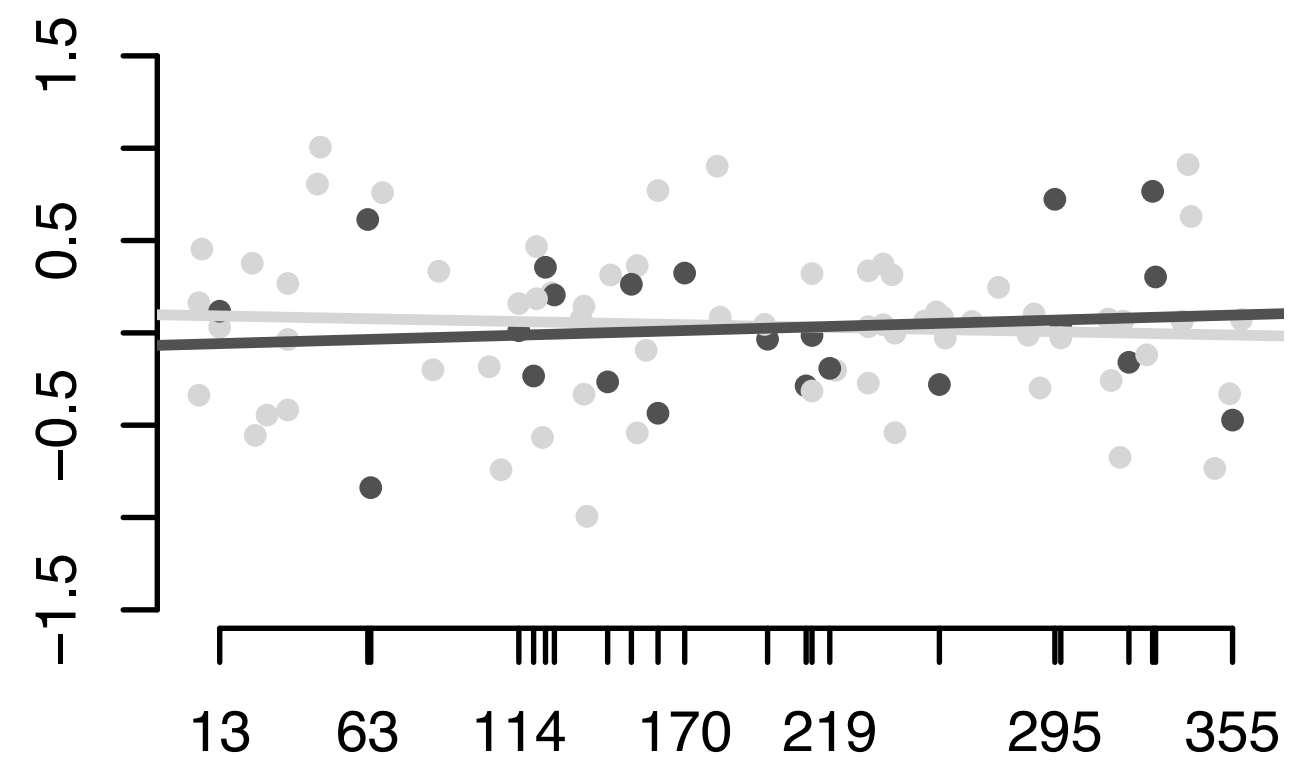
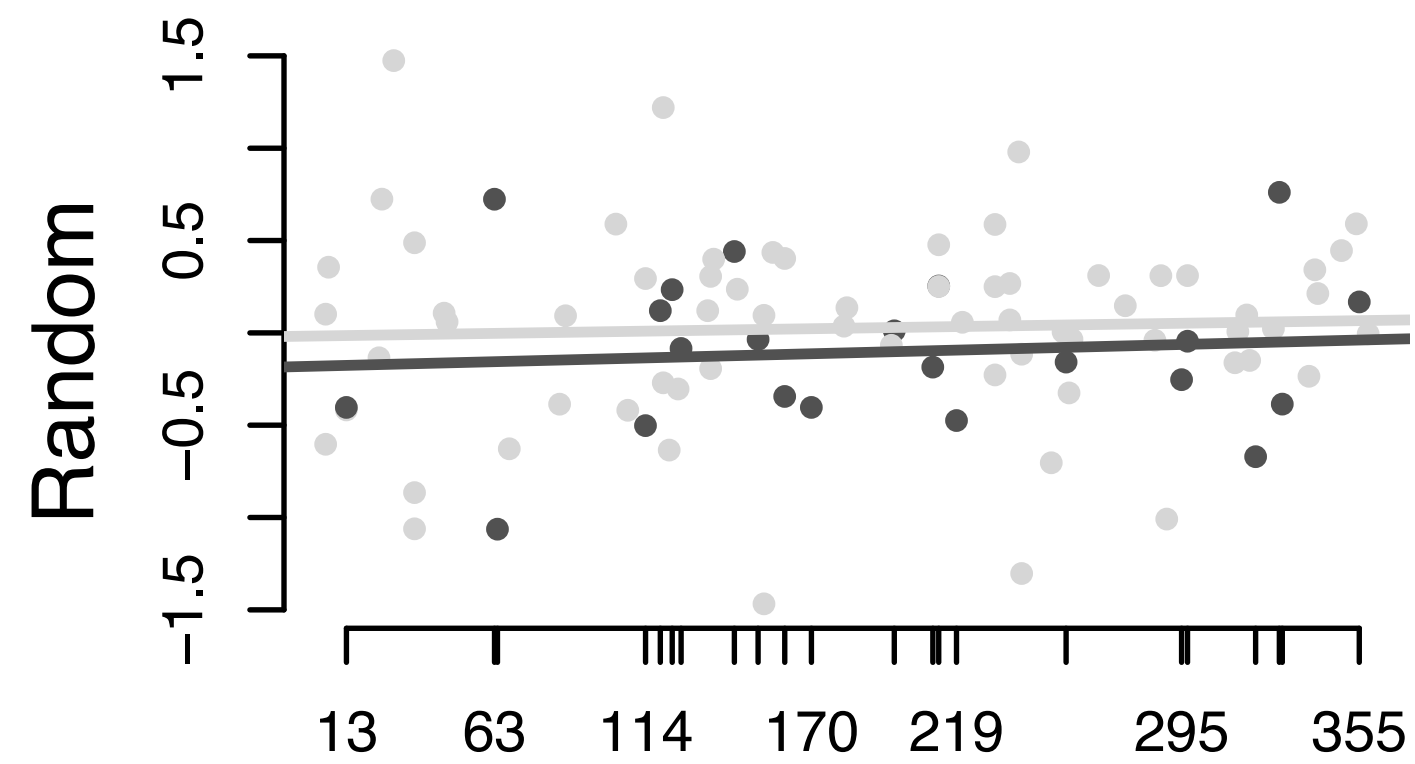
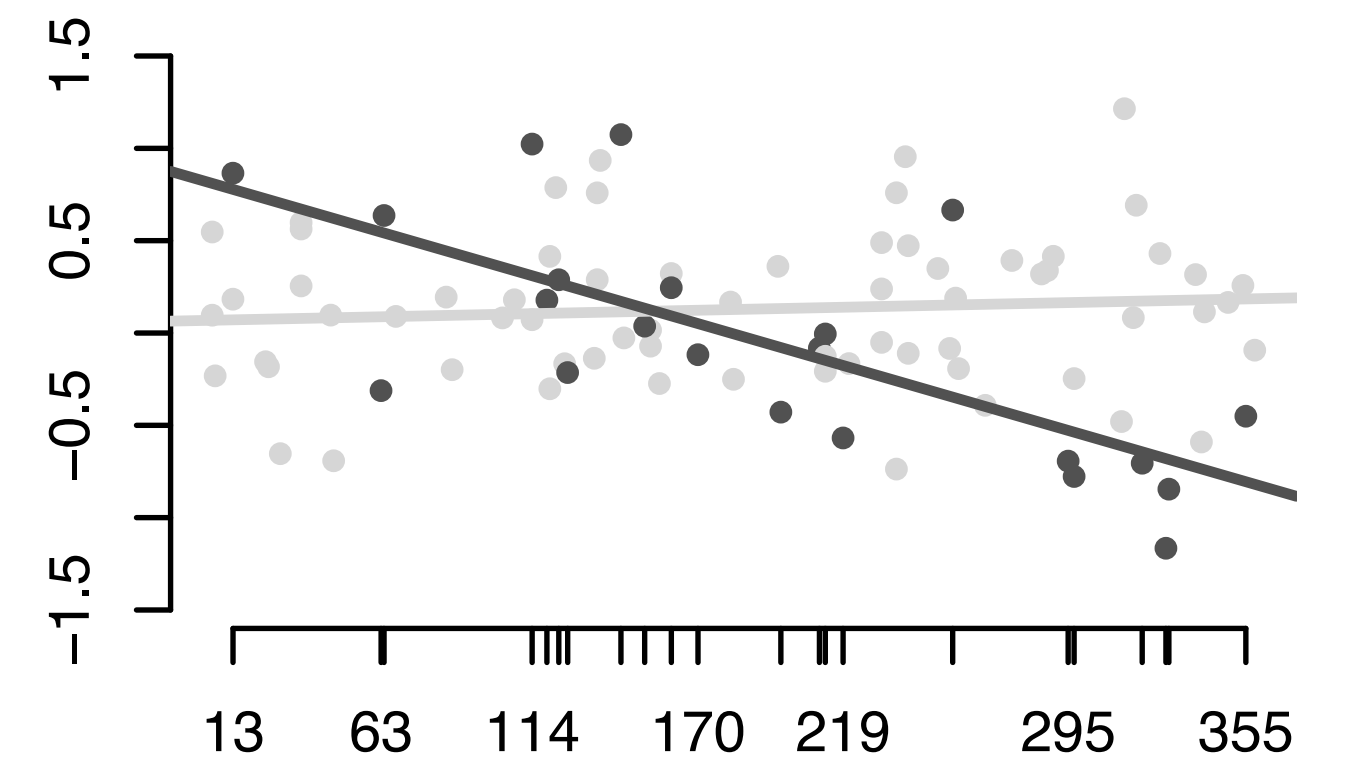
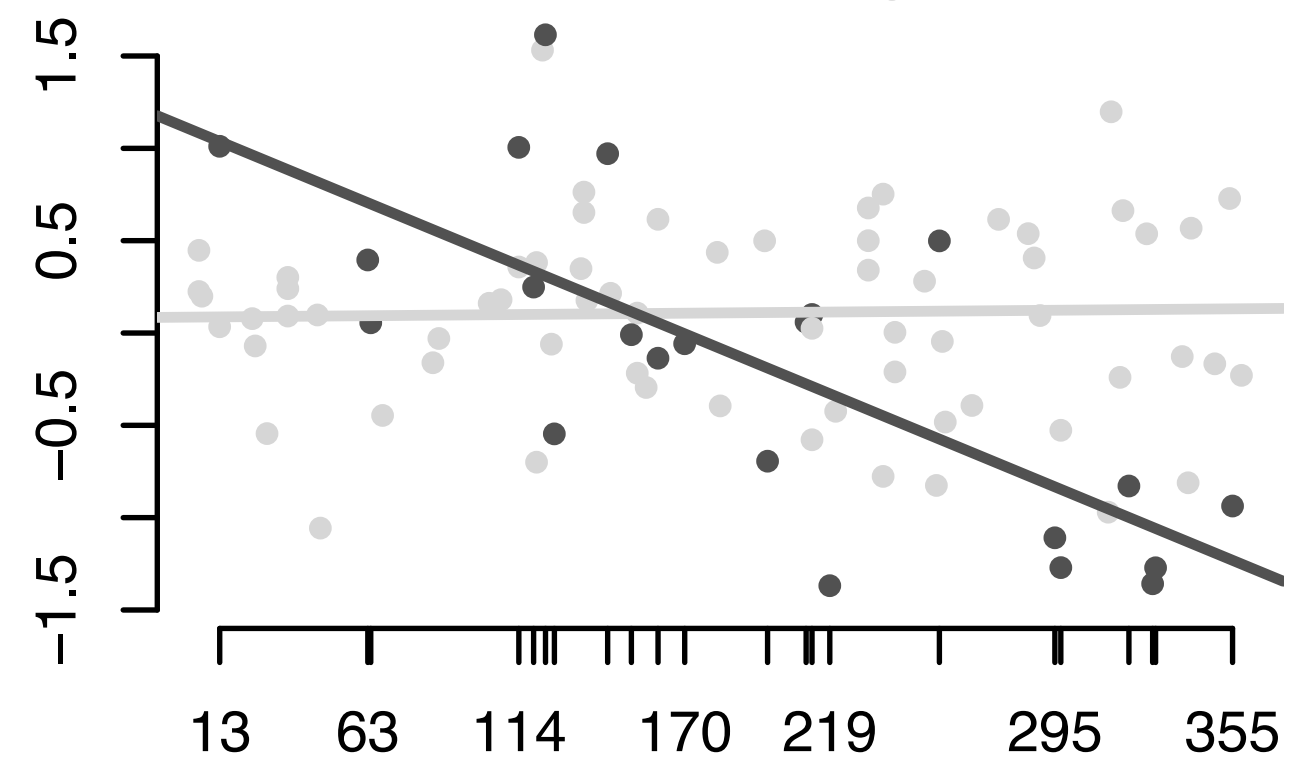
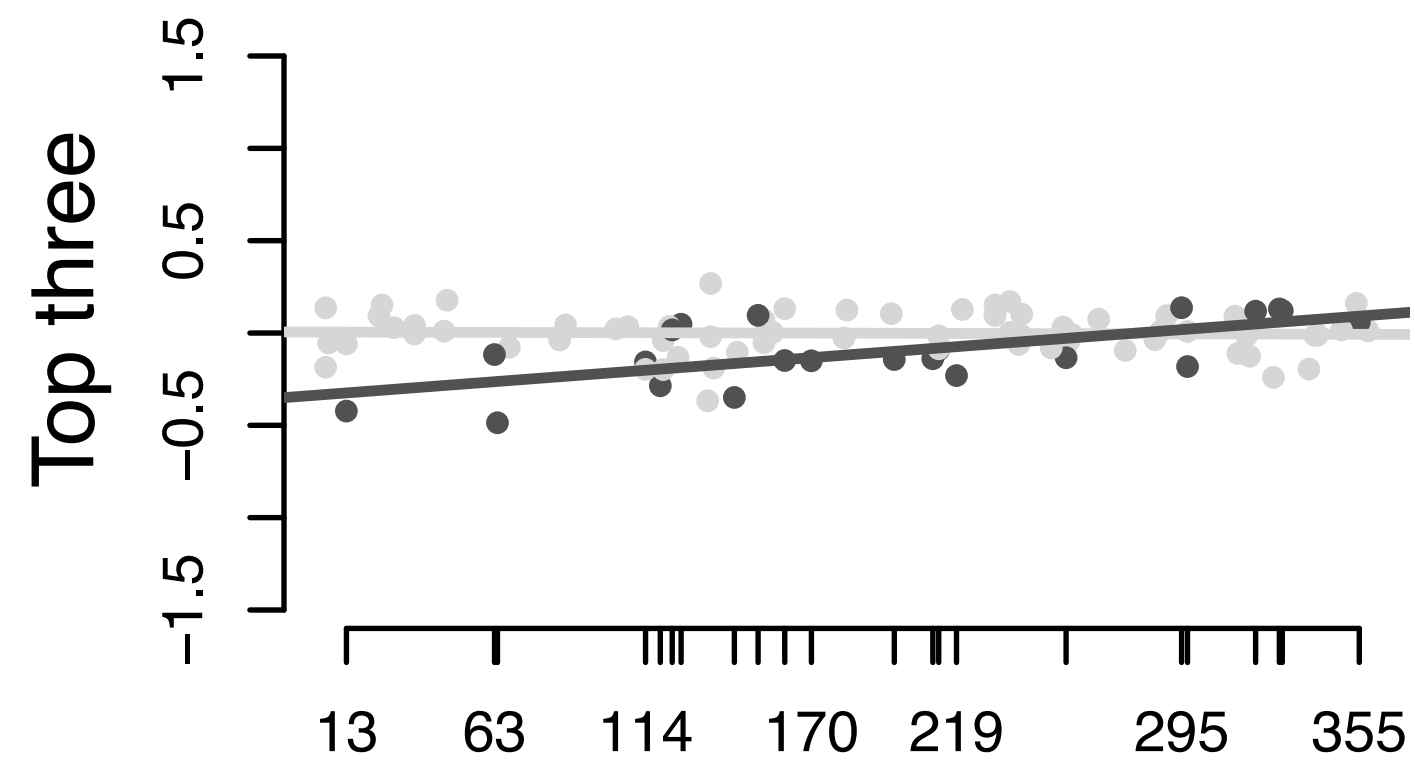


# Introducing some bias: focus on a likely subspace

- In high dimensions, bias is your friend
- Theory: there is something going on in the gene expression as we get closer to diagnosis
- Rank by linear model:

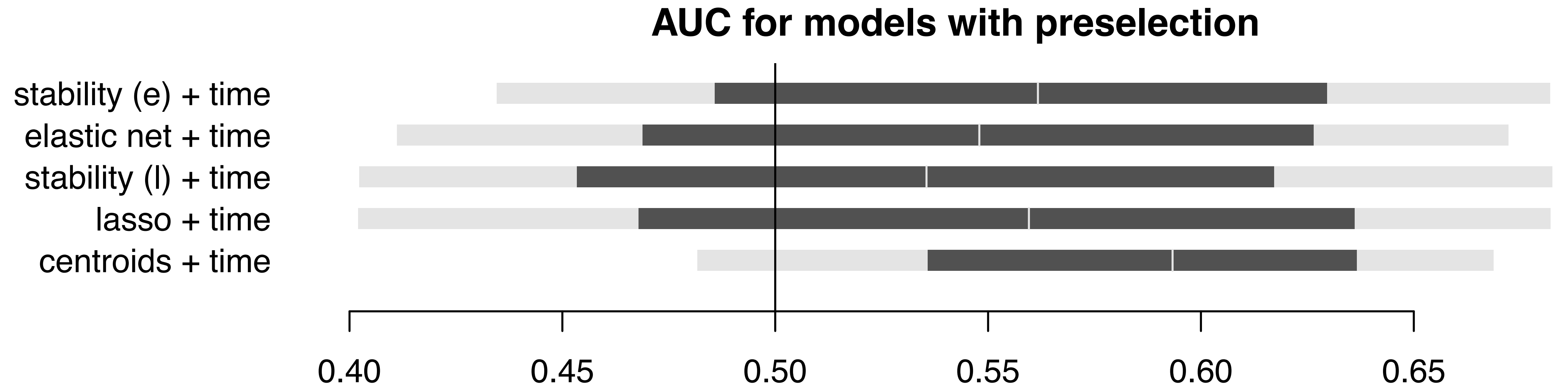
$$\text{expression} = \beta_0 + \beta_1 \text{time} + \beta_2 \text{metastasis} + \beta_3 \text{time} \times \text{metastasis} + \text{error}$$

# log(fold change) as linear function of time-to-diagnosis

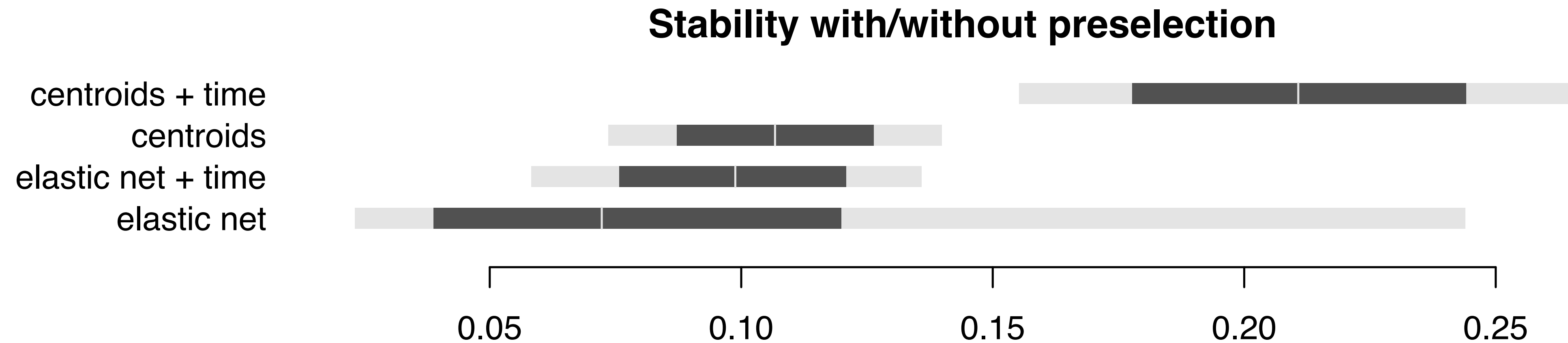


● metastasized ● non-metastasized

# Improved predictions



# Lower variance/higher stability



Stability = set overlap between predictive genes across two resmaplings

# Lessons/perspectives

- Cross validation can actually be super high in variance, be careful
- But be especially careful of holdout set validation
- Remember Simpson's paradox, watch your strata
- Be critical of Signatures

# Lessons/perspectives

- OTOH: There seems to be some weak signal here



# These are my advisers

- **Lars Ailo Bongo**, BDPS group, University of Tromsø
- **Etienne Birmelé**, MAP5, Université Paris Descartes
- **Eiliv Lund**, Department of Community Medicine, University of Tromsø



# Thank you!

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Slides available online at `3inar.github.io/talks/`

