





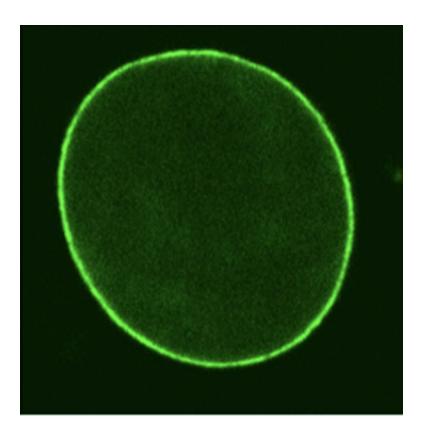
# Let's talk about the last lab!

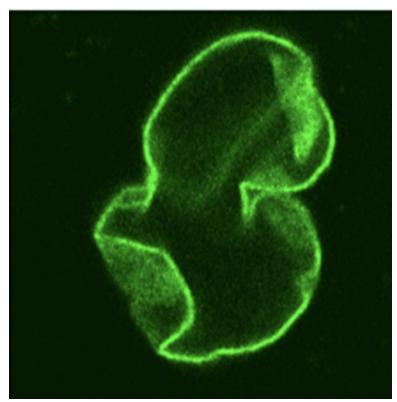
## Let's talk about the last lab!

# What circumstances made the model fit better? worse?

what does this even mean?







# **Progeria affects ~159 patients in the US**

we have a dataset of all American pediatric patients

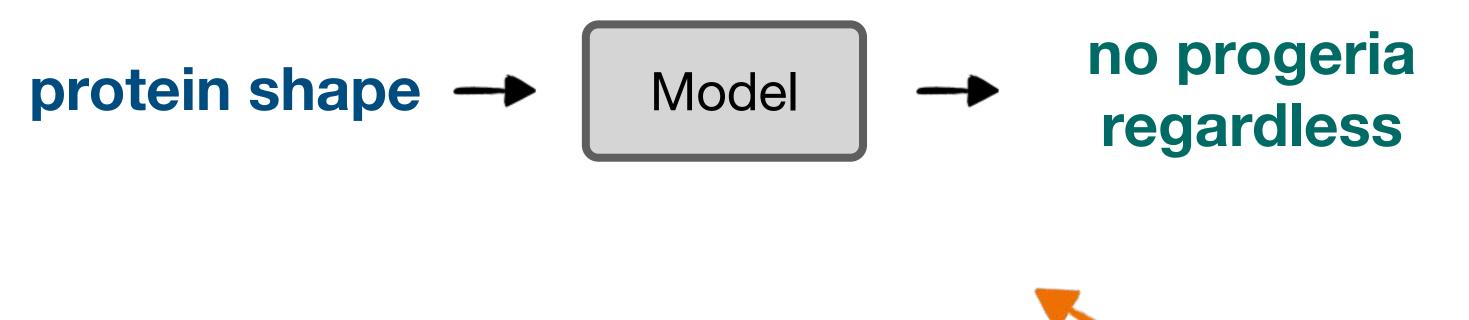
Q: If my model predicts with 99.99% accuracy, is it good enough?



we have a dataset of all American pediatric patients



#### **Progeria affects ~159 patients in the US**



we have a dataset of all American pediatric patients



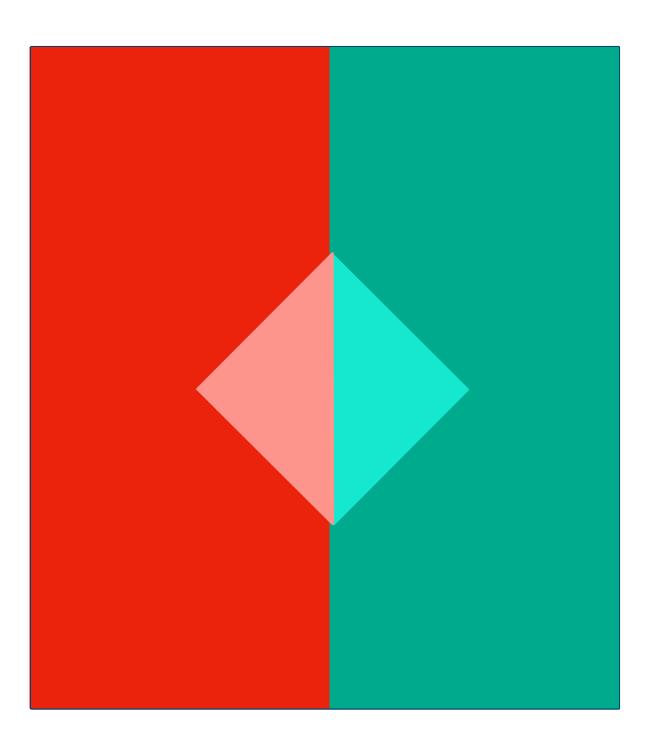
#### actually pretty accurate!

### **Progeria affects ~159 patients in the US**



Accuracy, Precision, and Recall

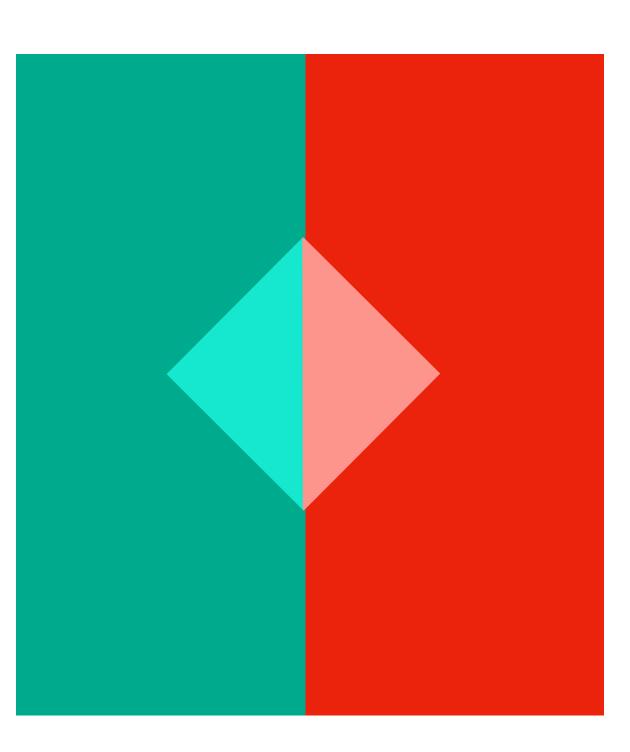




"Selection space"

Model selects positive and patient is positive



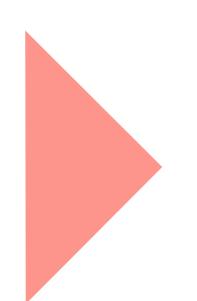


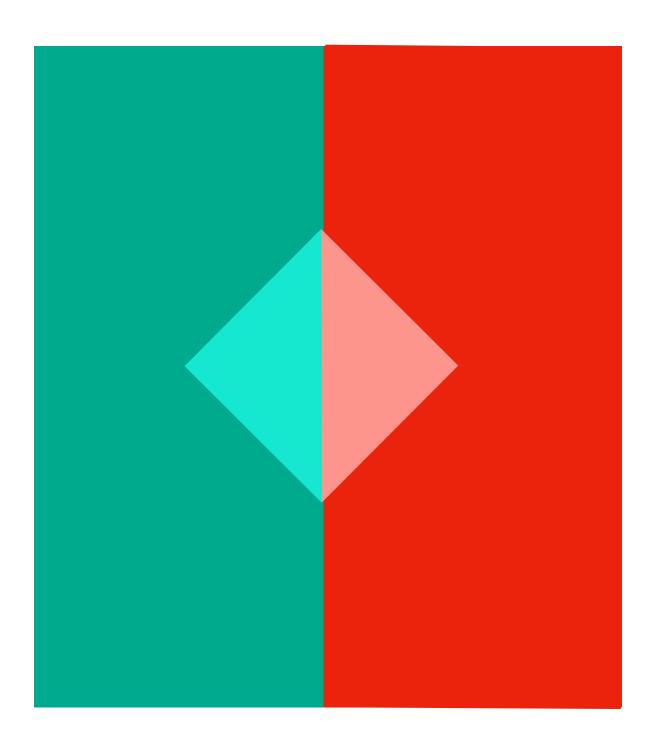
"Selection space"

Model selects **positive** and patient is **positive** 



Model selects **positive** and patient is **negative** 



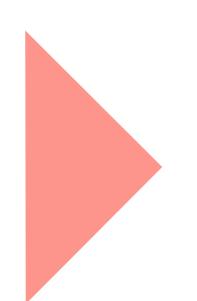


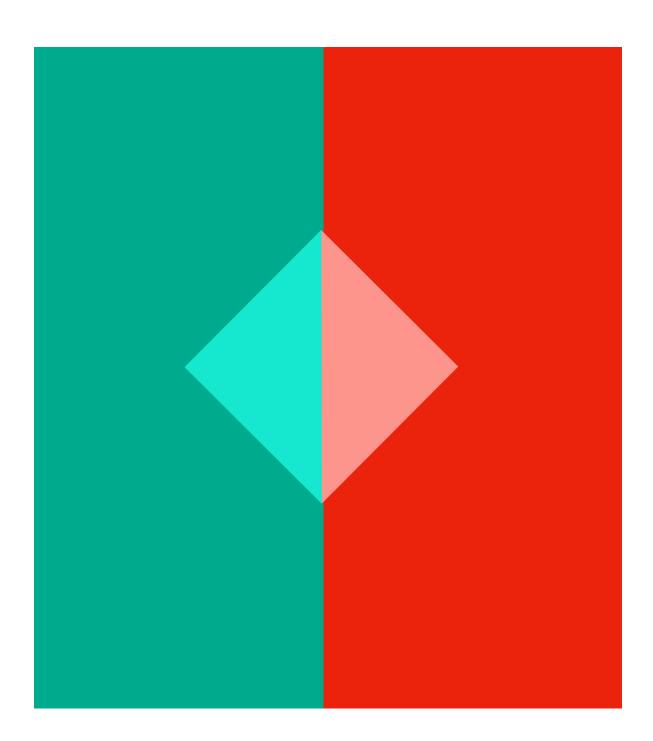
"Selection space"

Model selects positive and patient is positive



Model selects positive and patient is negative





"Selection space"



Model selects negative and patient is negative

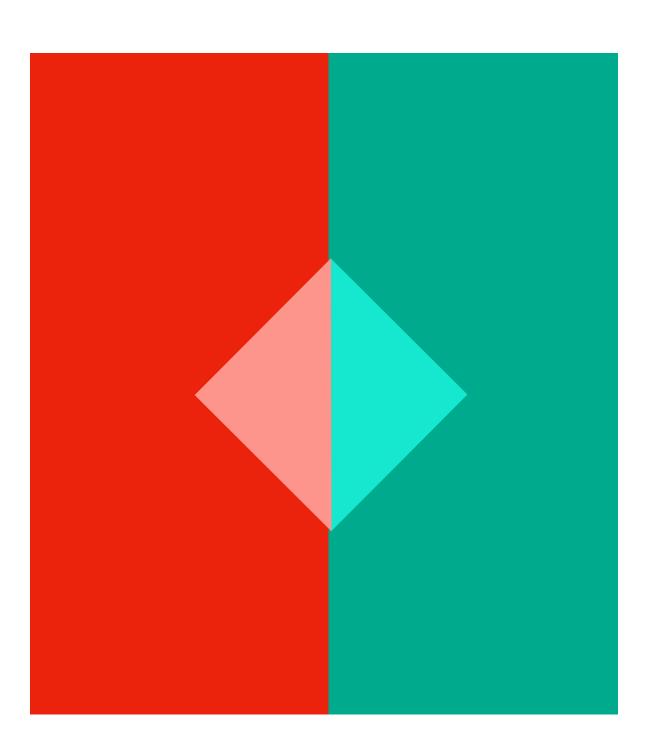
### TRUE POSITIVE

**TP: Model selects** positive and patient is **positive** 

### FALSE POSITIVE

FP: Model selects positive and patient is **negative** 



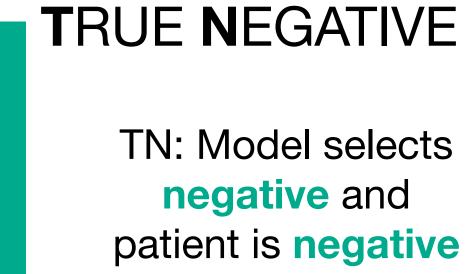


"Selection space"



#### FALSE NEGATIVE

FN: Model selects negative and patient is **positive** 





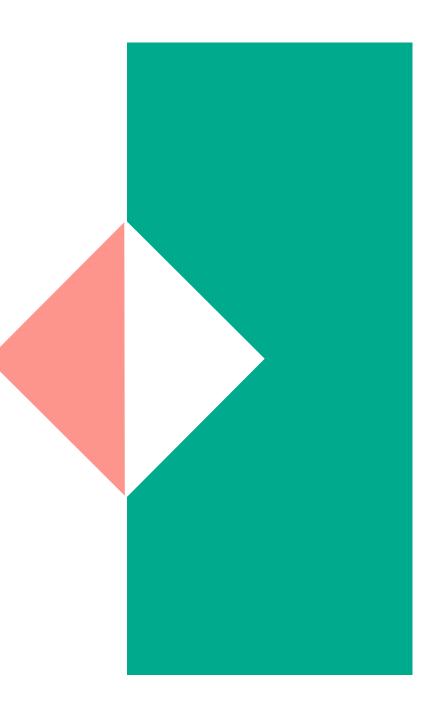




## Accuracy

#### Overall ability of model

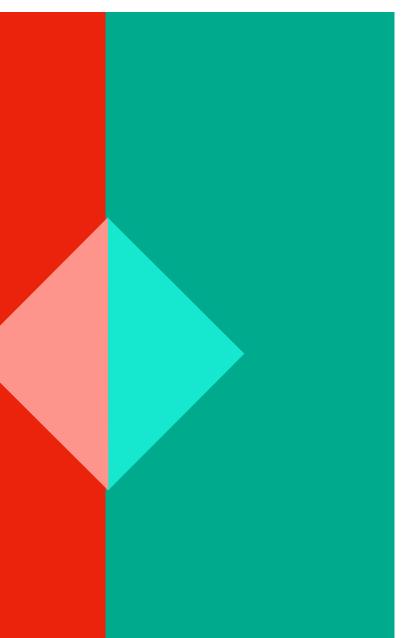




"Number of cases where we chose positive when patient is positive

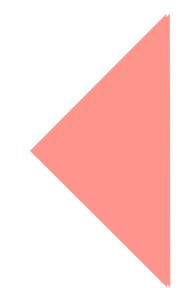
and

Number of cases where we chose negative when patient is negative"



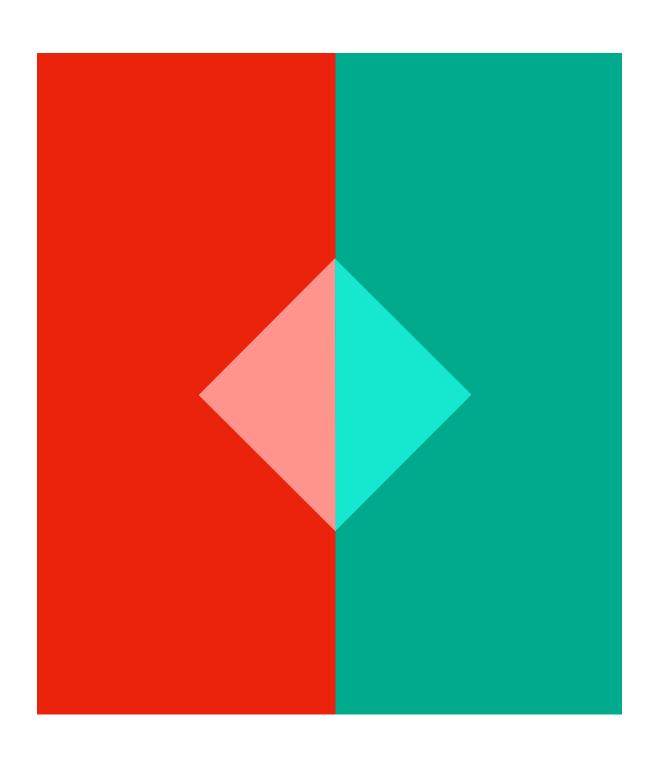
"Everything"

**TP: Model selects** positive and patient is **positive** 



FP: Model selects positive and patient is **negative** 





#### Accuracy

Overall ability of model





FN: Model selects negative and patient is **positive** 



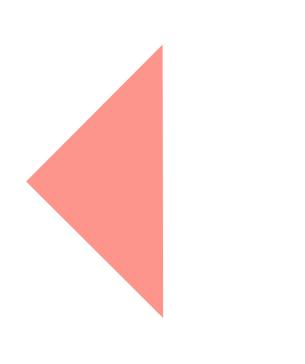
TN: Model selects negative and patient is **negative** 



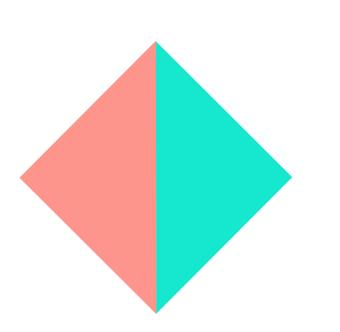


#### Precision

Accuracy of what we selected. Or amount of selection that's actually correct.

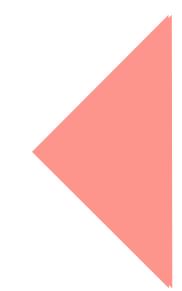


"Number of cases where we chose positive when patient is positive"



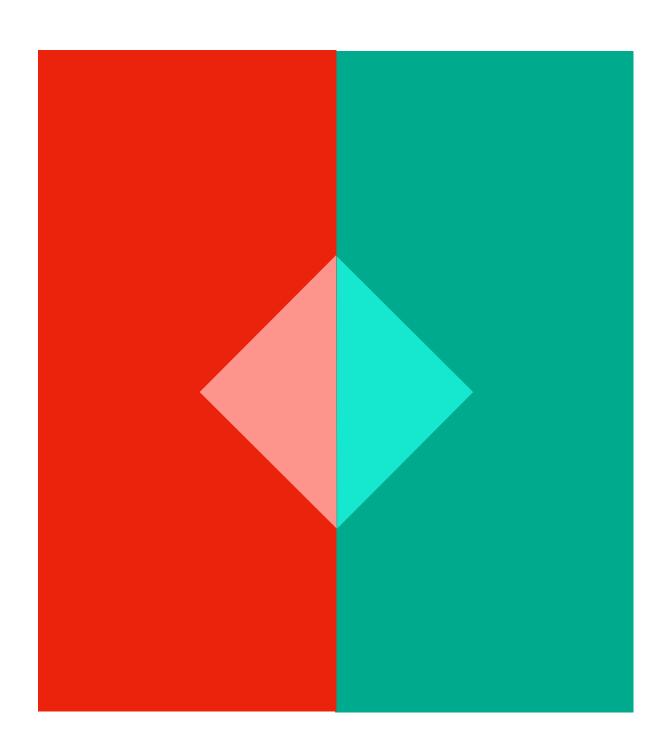
"All selected **positive** by the model"

**TP: Model selects** positive and patient is **positive** 



FP: Model selects positive and patient is **negative** 





"Selection space"

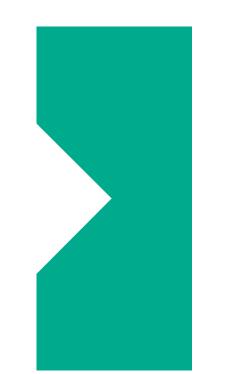
**Precision** Amount of selection that's actually correct.

#### Accuracy

Overall ability of model



FN: Model selects negative and patient is **positive** 



TN: Model selects negative and patient is **negative** 

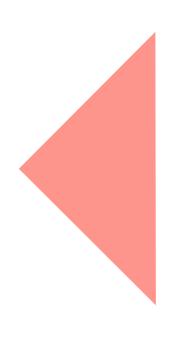
18





#### Recall

Accuracy of what we should select. Or amount of what needs to be selected that is selected

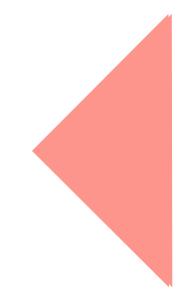


#### "Number of cases where **we** chose positive when patient is positive"

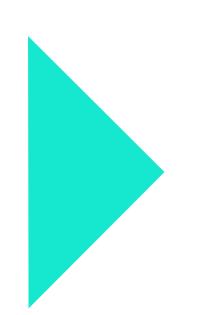
#### "All cases that the patients are positive"

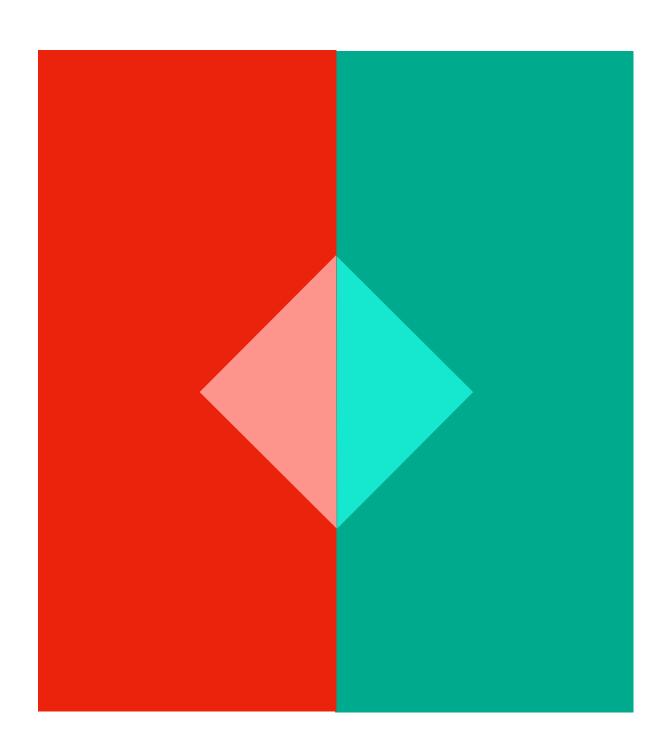


**TP: Model selects positive** and patient is **positive** 



FP: Model selects positive and patient is **negative** 





"Selection space"

Accuracy

Overall ability of model

Amount of selection that's actually correct.



FN: Model selects negative and patient is **positive** 



TN: Model selects negative and patient is **negative** 

#### Precision

#### Recall

Amount of what needs to be selected that is selected









#### FALSE POSITIVE



Overall ability of model

Amount of selection that's actually correct.



#### FALSE NEGATIVE



#### TRUE NEGATIVE

Precision

#### Recall

Amount of what needs to be selected that is selected



		Predicted condition		Sources: [6][7][8][9][10][11][12][13][14] view · talk · edit	
	Total population = P + N	Positive (PP)	Negative (PN)	Informedness, bookmaker informedness (BM) = TPR + TNR – 1	$\frac{\text{Prevalence threshold (PT)}}{=\frac{\sqrt{\text{TPR} \times \text{FPR}} - \text{FPR}}{\text{TPR} - \text{FPR}}}$
Actual condition	Positive (P)	<b>True positive (TP),</b> hit	False negative (FN), type II error, miss, underestimation	True positive rate (TPR), recall, sensitivity (SEN), probability of detection, hit rate, power $=\frac{TP}{P}=1-FNR$	False negative rate (FNR), miss rate = $\frac{FN}{P}$ = 1 – TPR
	Negative (N)	False positive (FP), type I error, false alarm, overestimation	True negative (TN), correct rejection	False positive rate (FPR), probability of false alarm, fall-out $= \frac{FP}{N} = 1 - TNR$	True negative rate (TNR), specificity (SPC), selectivity $=\frac{TN}{N} = 1 - FPR$
	$\frac{\text{Prevalence}}{=\frac{P}{P+N}}$	Positive predictive value (PPV), precision $= \frac{TP}{PP} = 1 - FDR$	False omission rate (FOR) = $\frac{FN}{PN}$ = 1 – NPV	Positive likelihood ratio (LR+) = TPR FPR	Negative likelihood ratio (LR–) = $\frac{FNR}{TNR}$
	$\frac{\text{Accuracy (ACC)}}{= \frac{\text{TP} + \text{TN}}{\text{P} + \text{N}}}$	False discovery rate (FDR) = $\frac{FP}{PP}$ = 1 – PPV	Negative predictive value (NPV) = $\frac{TN}{PN}$ = 1 – FOR	<mark>Markedness (MK)</mark> , deltaP (Δp) = PPV + NPV – 1	Diagnostic odds ratio (DOR) = $\frac{LR+}{LR-}$
	Balanced accuracy (BA) = TPR + TNR 2	$F_{1} \text{ score}$ $= \frac{2PPV \times TPR}{PPV + TPR} = \frac{2TP}{2TP + FP + FN}$	Fowlkes–Mallows index (FM) = $\sqrt{PPV \times TPR}$	Matthews correlation coefficient(MCC) $=\sqrt{TPR \times TNR \times PPV \times NPV}$ $-\sqrt{FNR \times FPR \times FOR \times FDR}$	Threat score (TS), critical success index (CSI), Jaccard index = $\frac{TP}{TP + FN + FP}$

https://en.wikipedia.org/wiki/Precision\_and\_recall

#### Accuracy

Overall ability of model



#### Precision

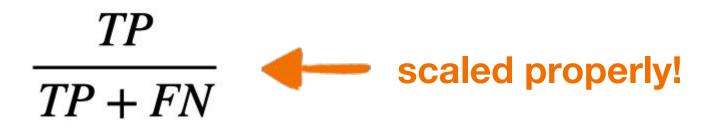
Amount of selection that's actually correct.

 $\frac{TP}{TP + FP}$ 

#### Recall

rı hav

Amount of what needs to be selected that is selected

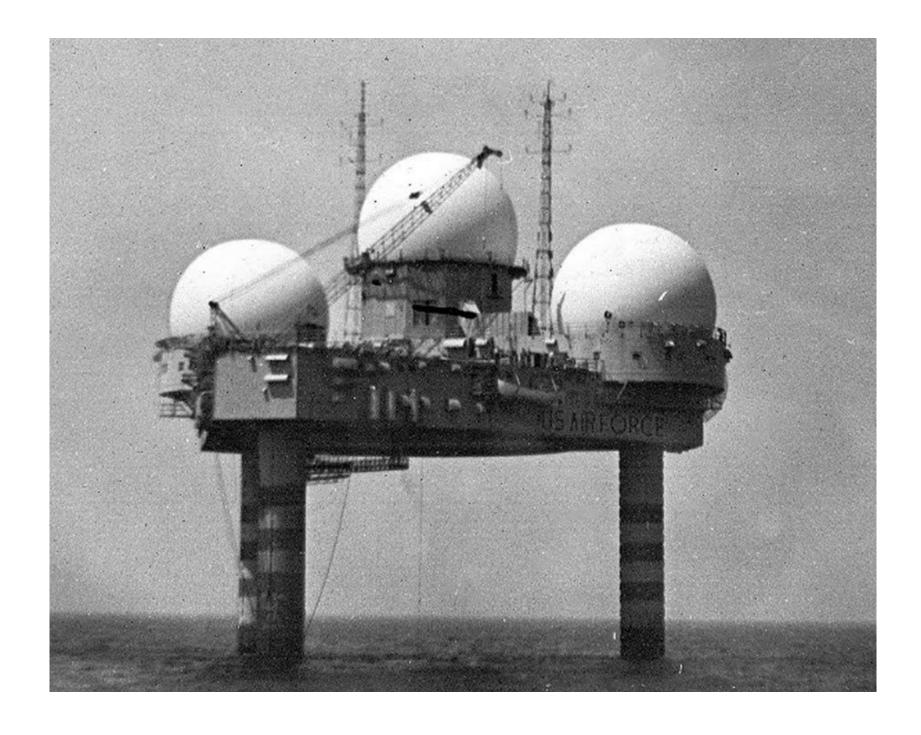


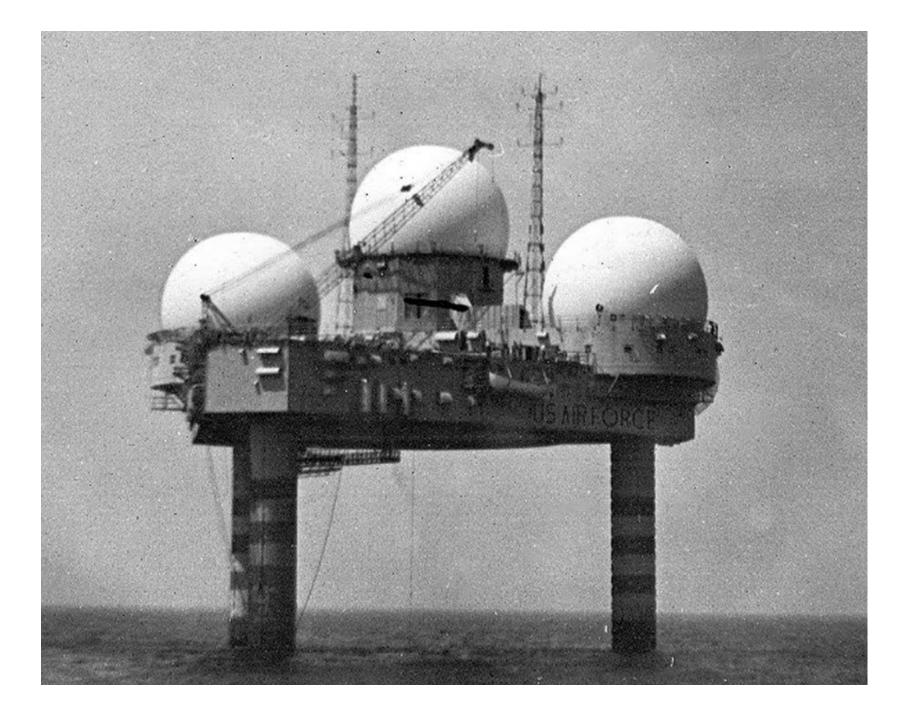


#### Progeria affects ~159 patients in the US

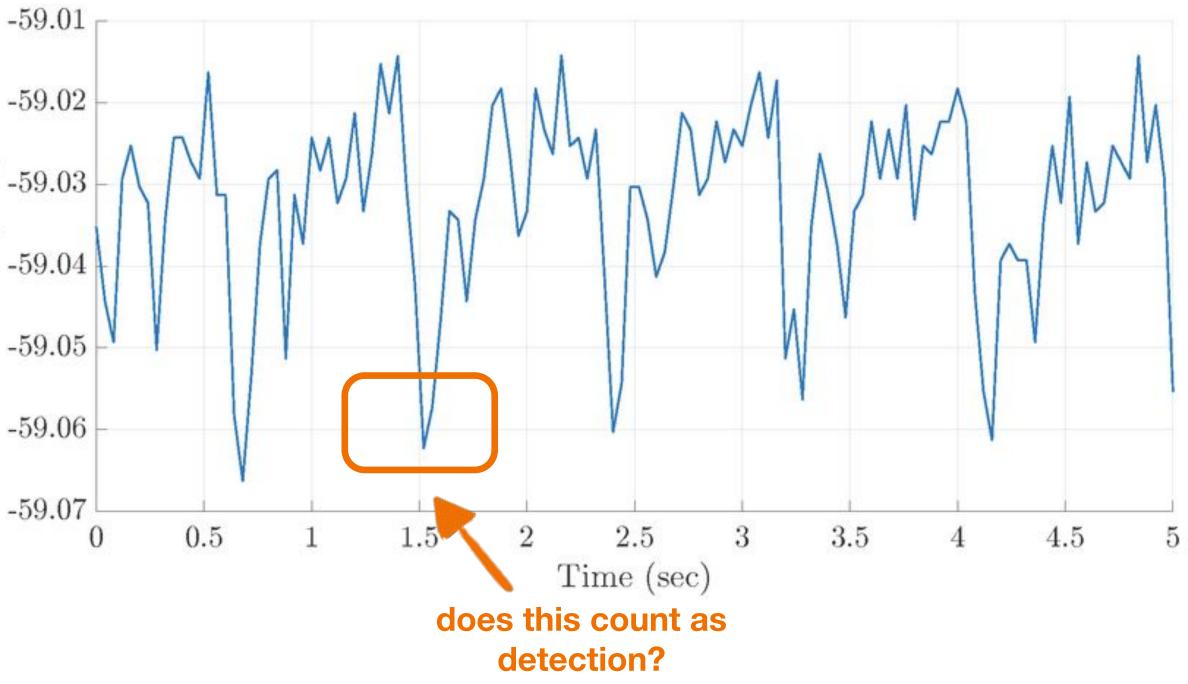
we have a dataset of all American pediatric patients

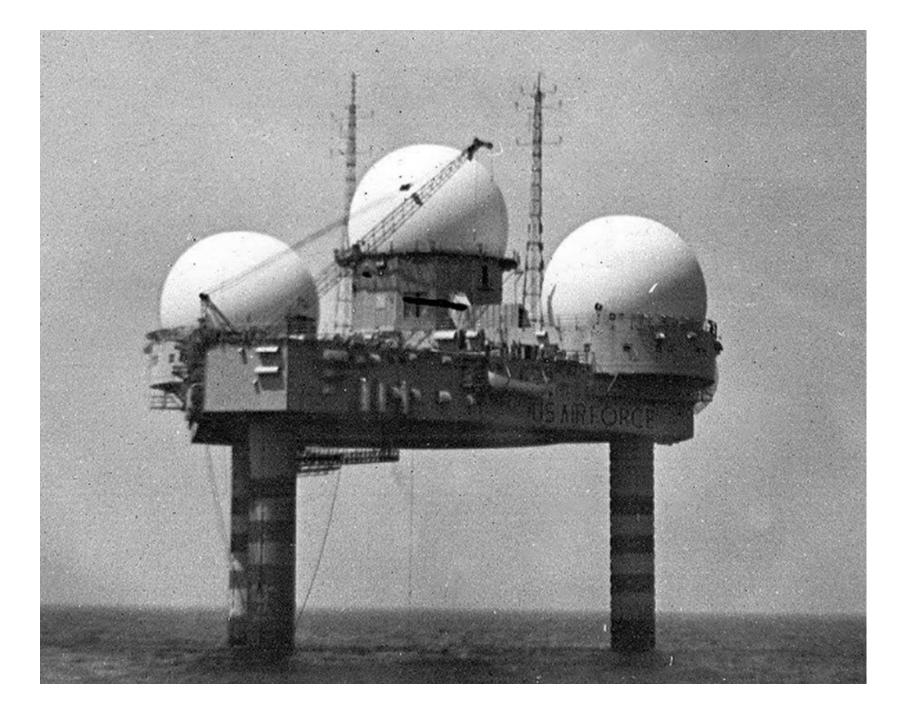


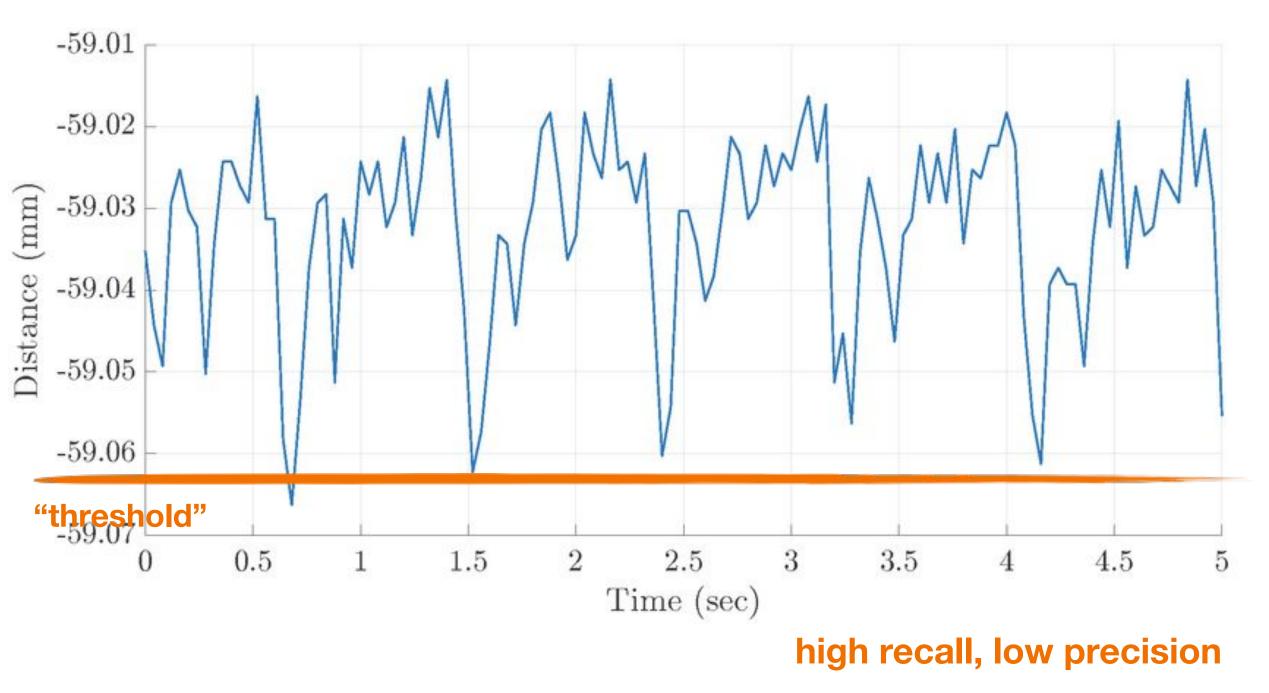


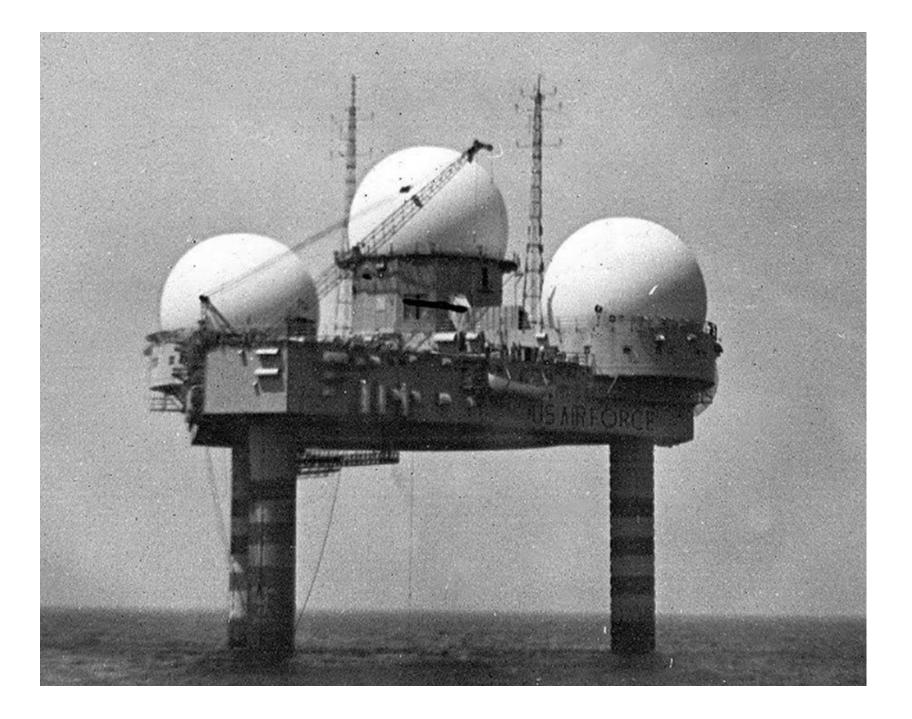


Distance (mm)

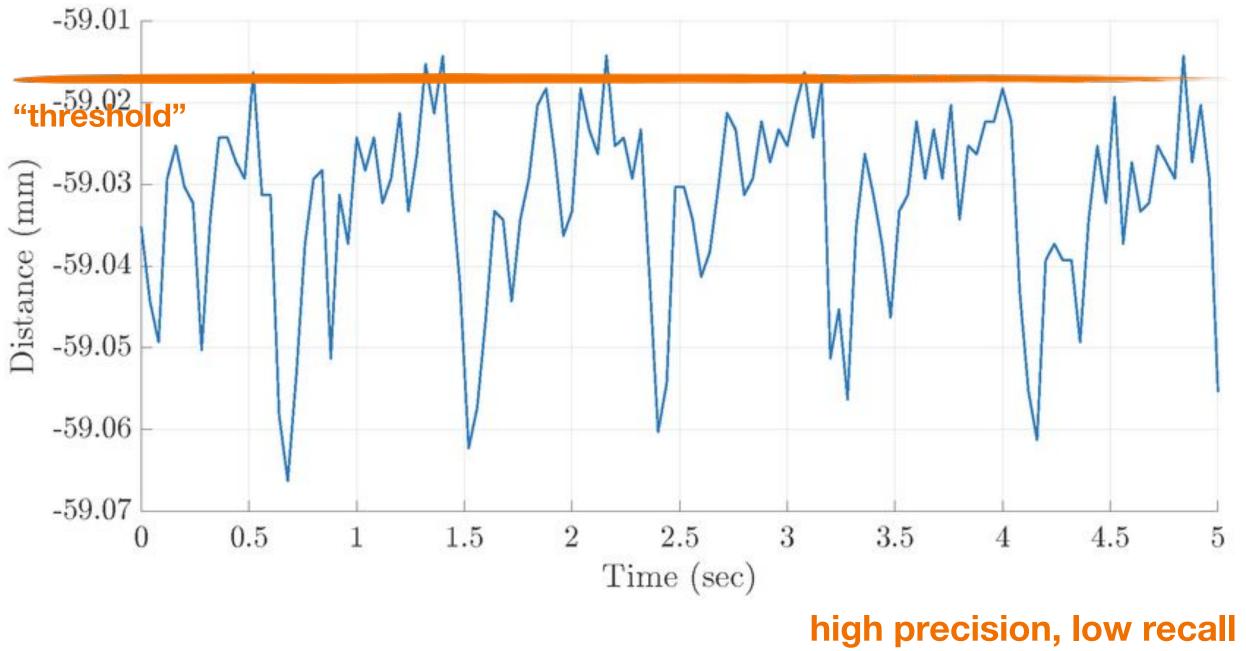








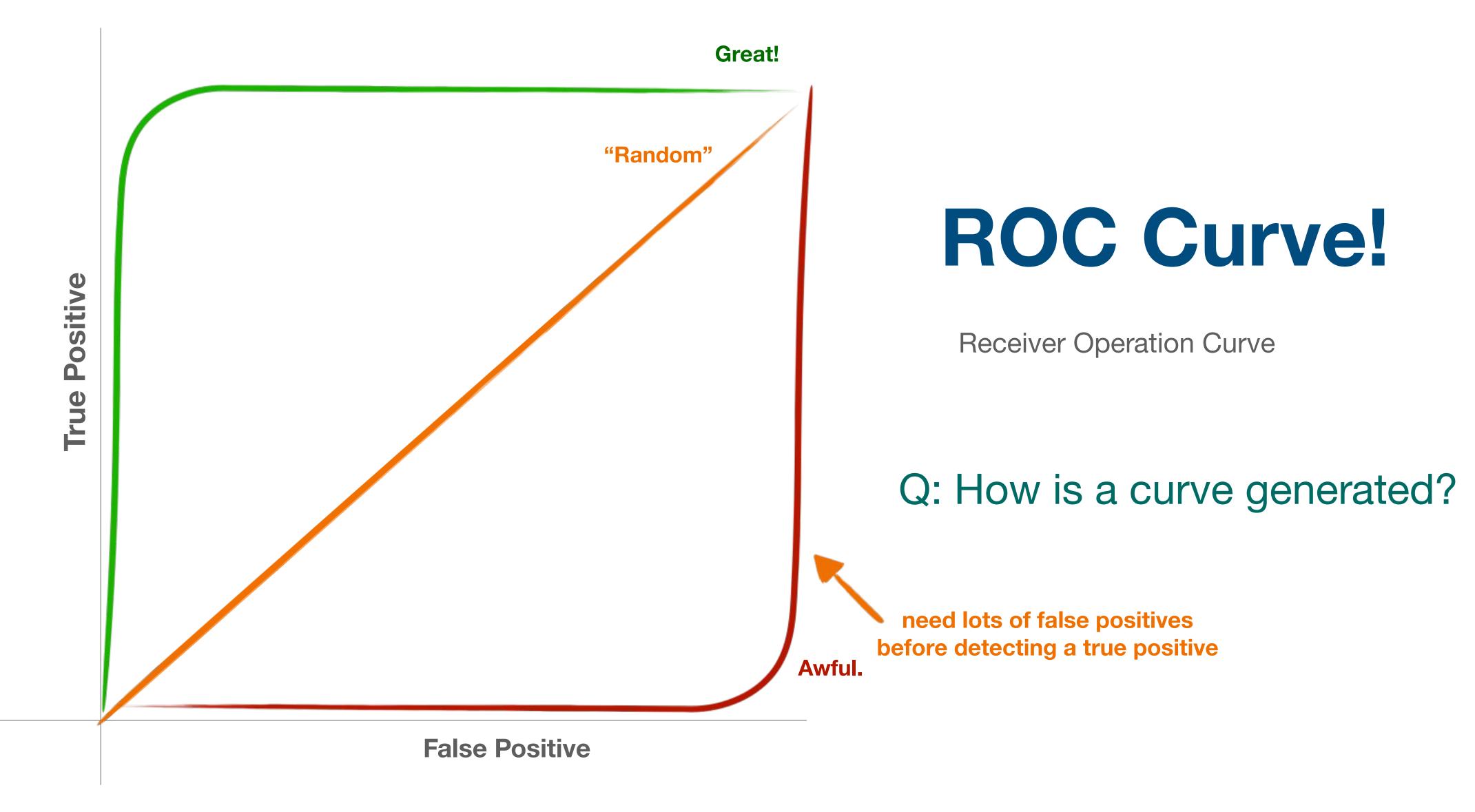
Distance (mm)



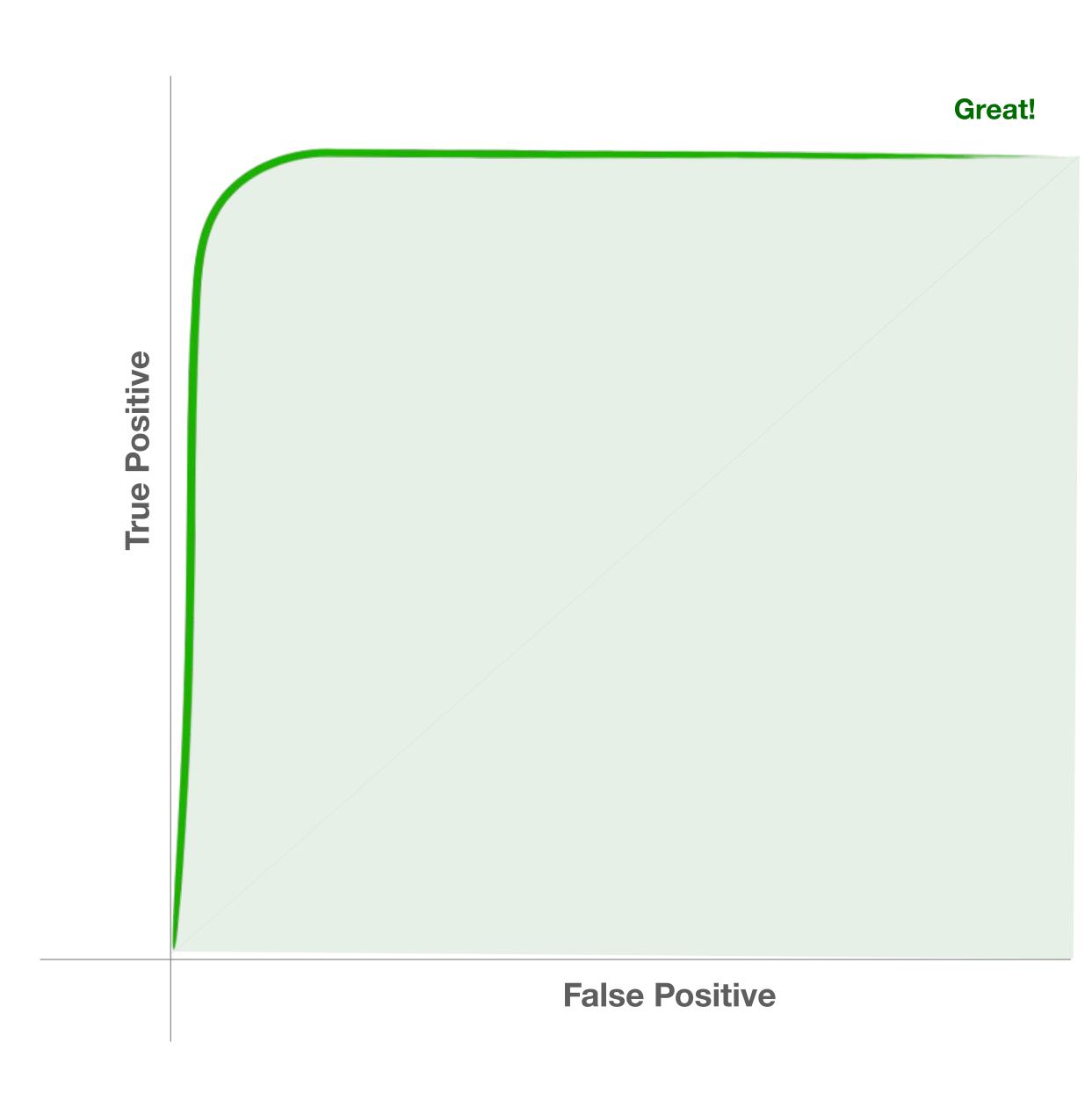
# quantifying "threshold"

# quantifying "threshold"

**ROC Curve!** 



■ ROC Curve quantify the amount of "error"/noise that is necessary for a classifier to make a good prediction



# AUC area under [the ROC] curve

Q: how do you compare these points



Precision	
	Recall

Self-test

# **Precision-recall AUC**

Q: When do we really need it? Q: what would it look like?

Especially for unbalanced datasets

# what makes models fit better

more data balanced data normalized data quality data

## more data balanced data normalized data

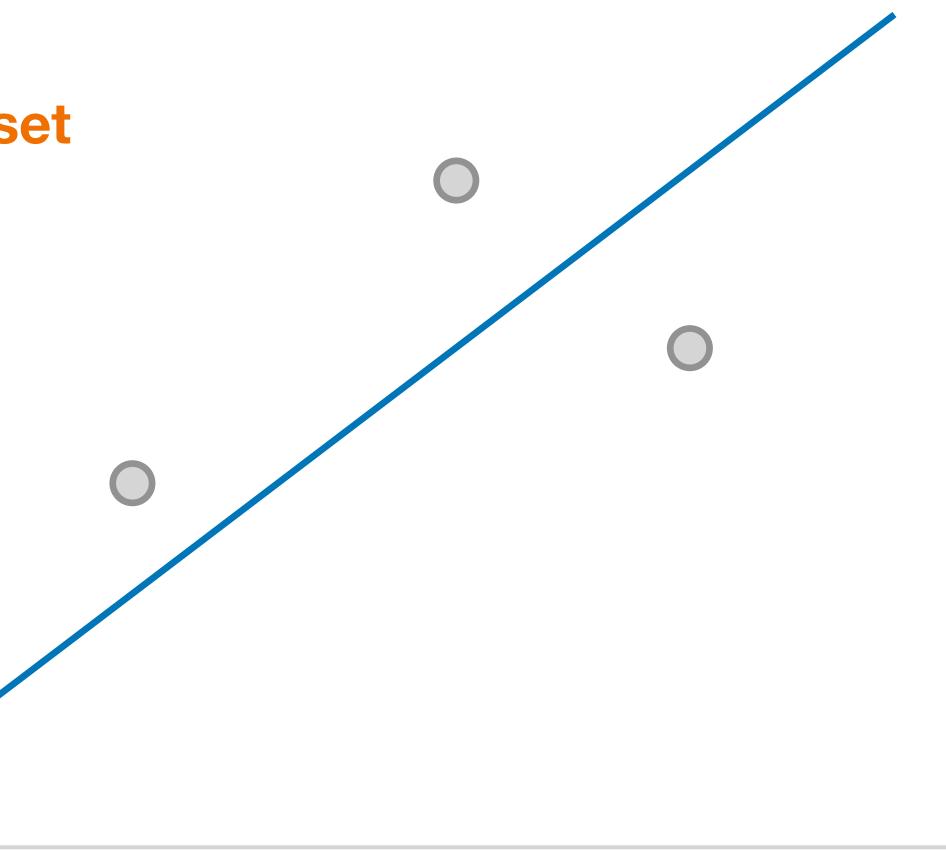
#### quality data

# more databalanced datanormalized dataquality data

## Quality

## let's say we have a simpler wine dataset

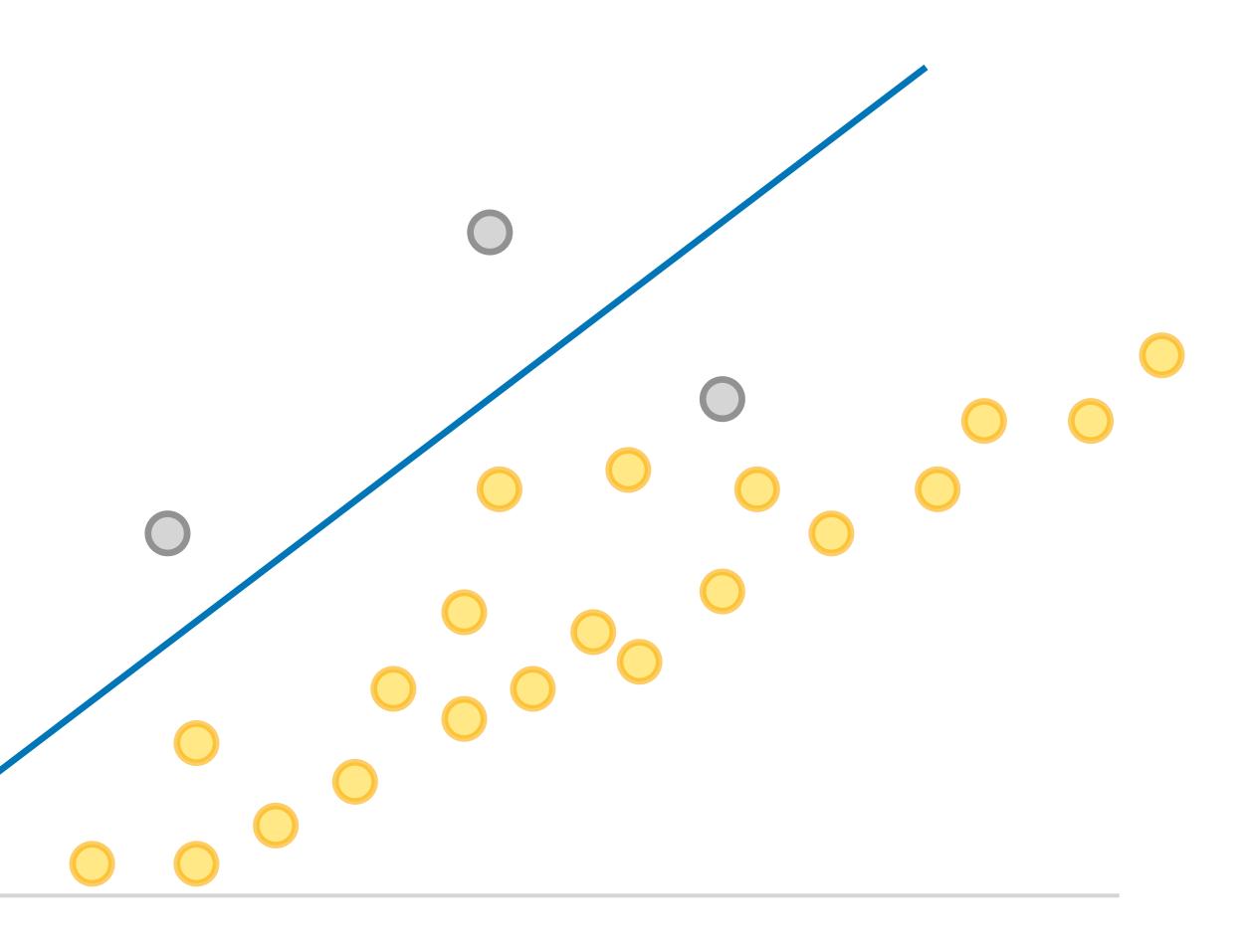
Quality on the y axis





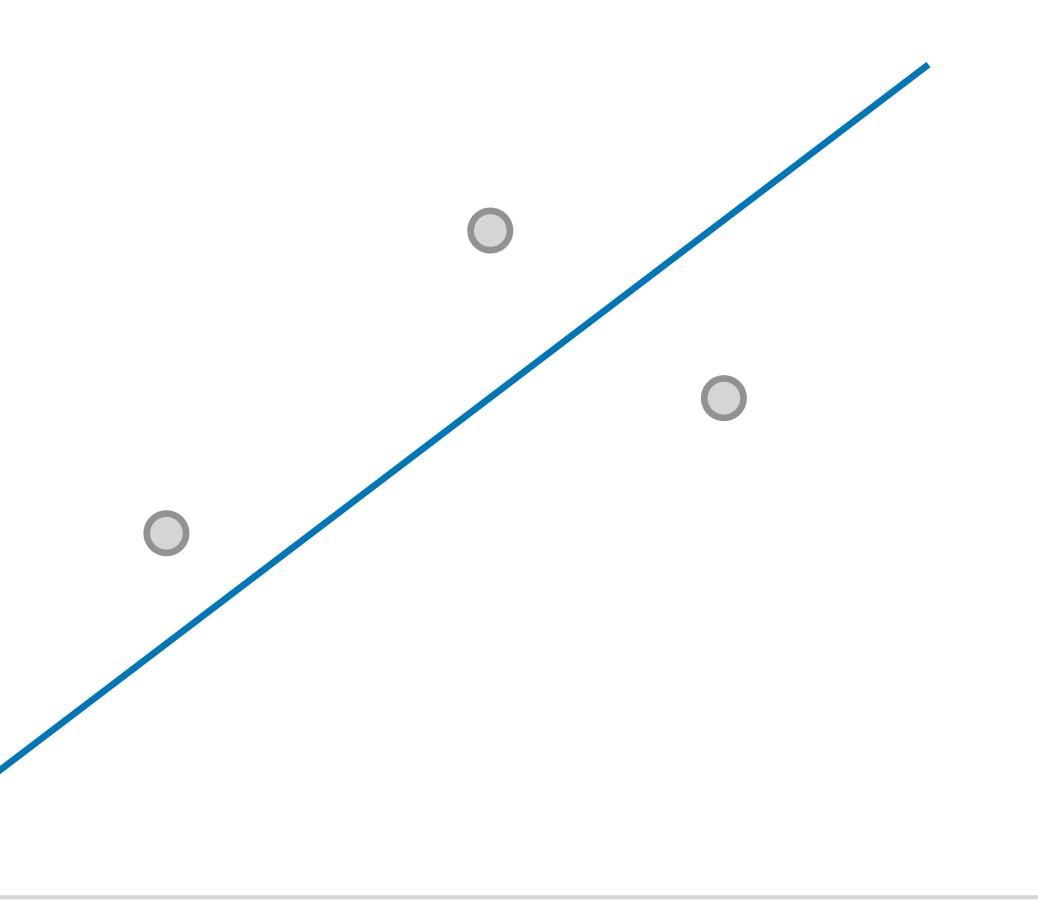
#### Quality

Quality on the y axis



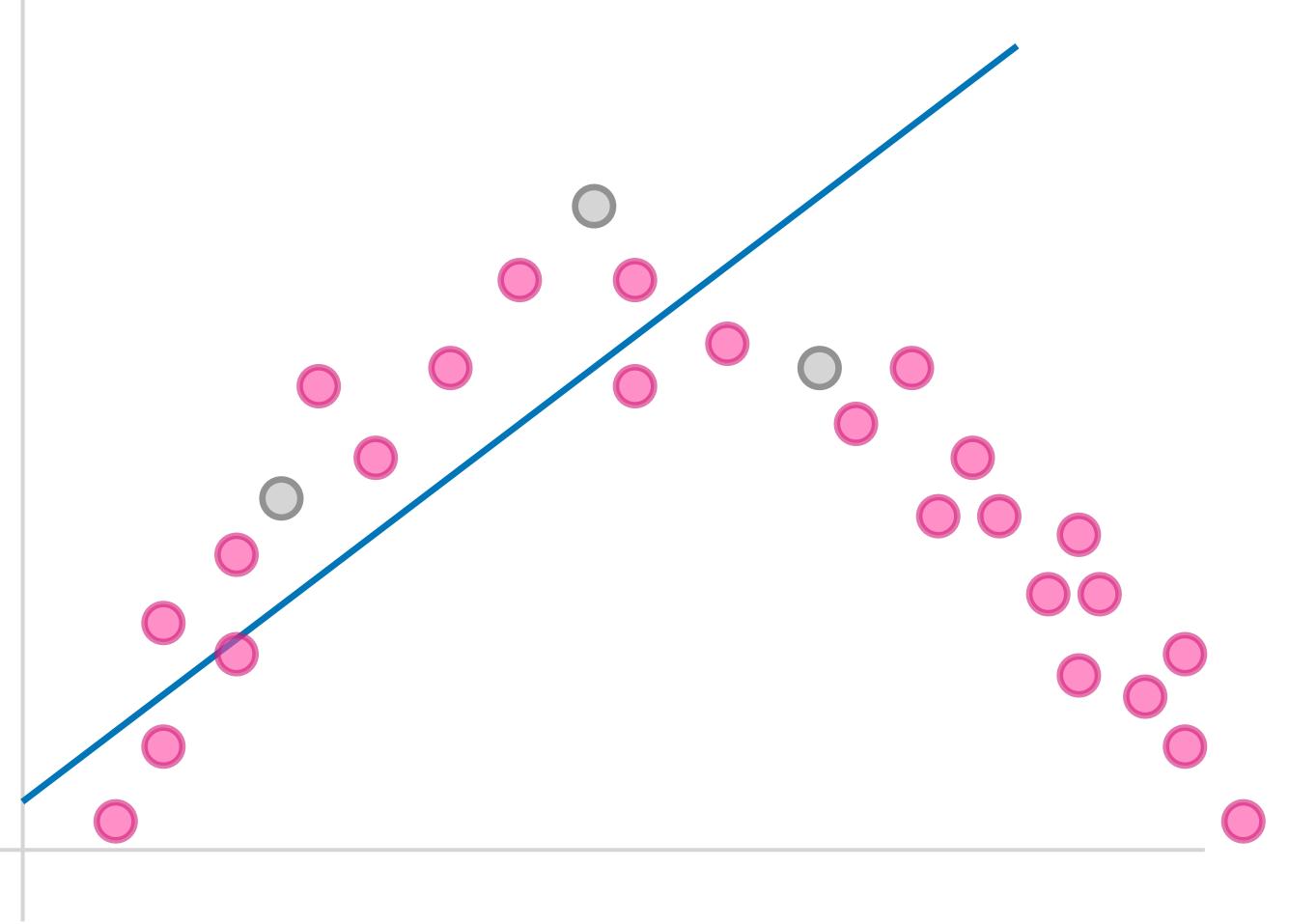
#### Quality

Quality on the y axis



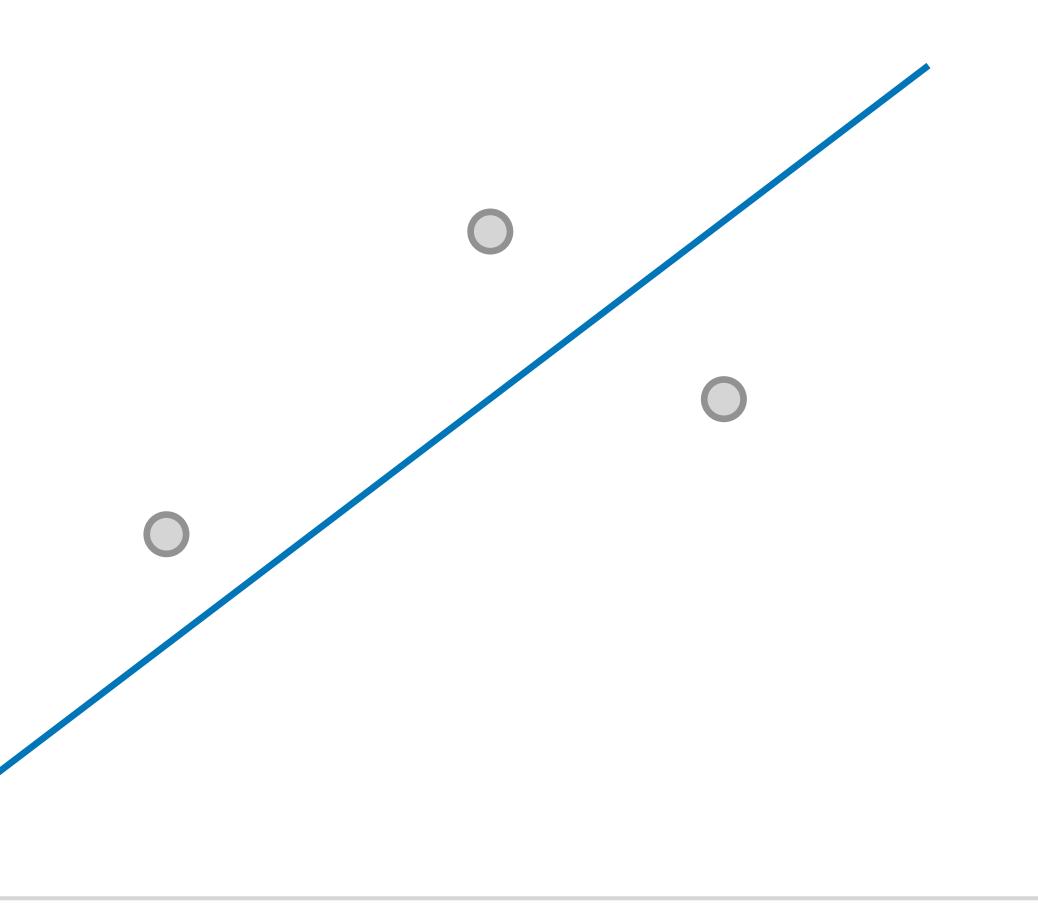
### Quality

Quality on the y axis



#### Quality

Quality on the y axis

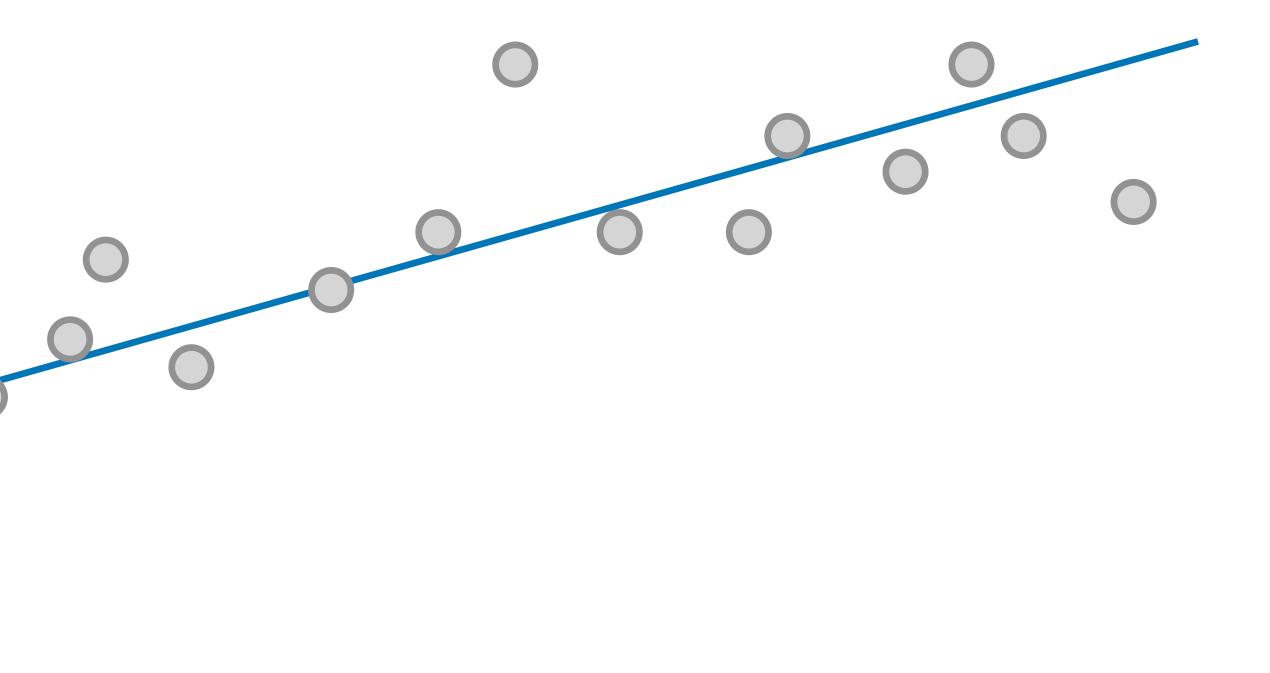




Quality on the y axis

Acidity on the x axis

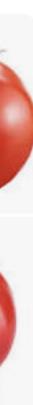




#### use more data, get more accurate results

## more databalanced datanormalized dataquality data







Amber Colored



Angora Super Sweet



Black Ethiopian



Burbank Slicing



Dona



Sophie's Choice



White Bush



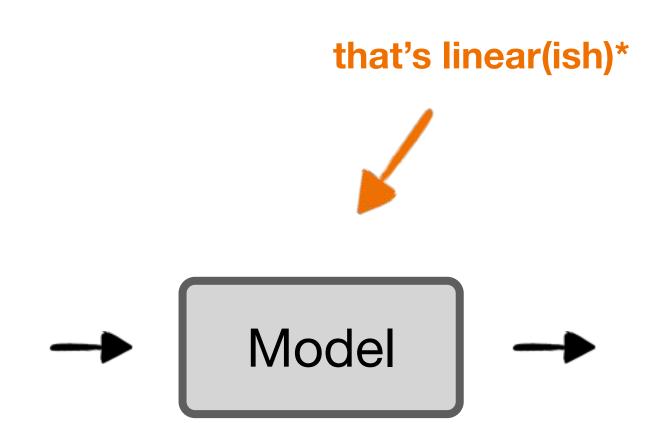
Ace 55





## more databalanced datanormalized dataquality data

## a reminder...

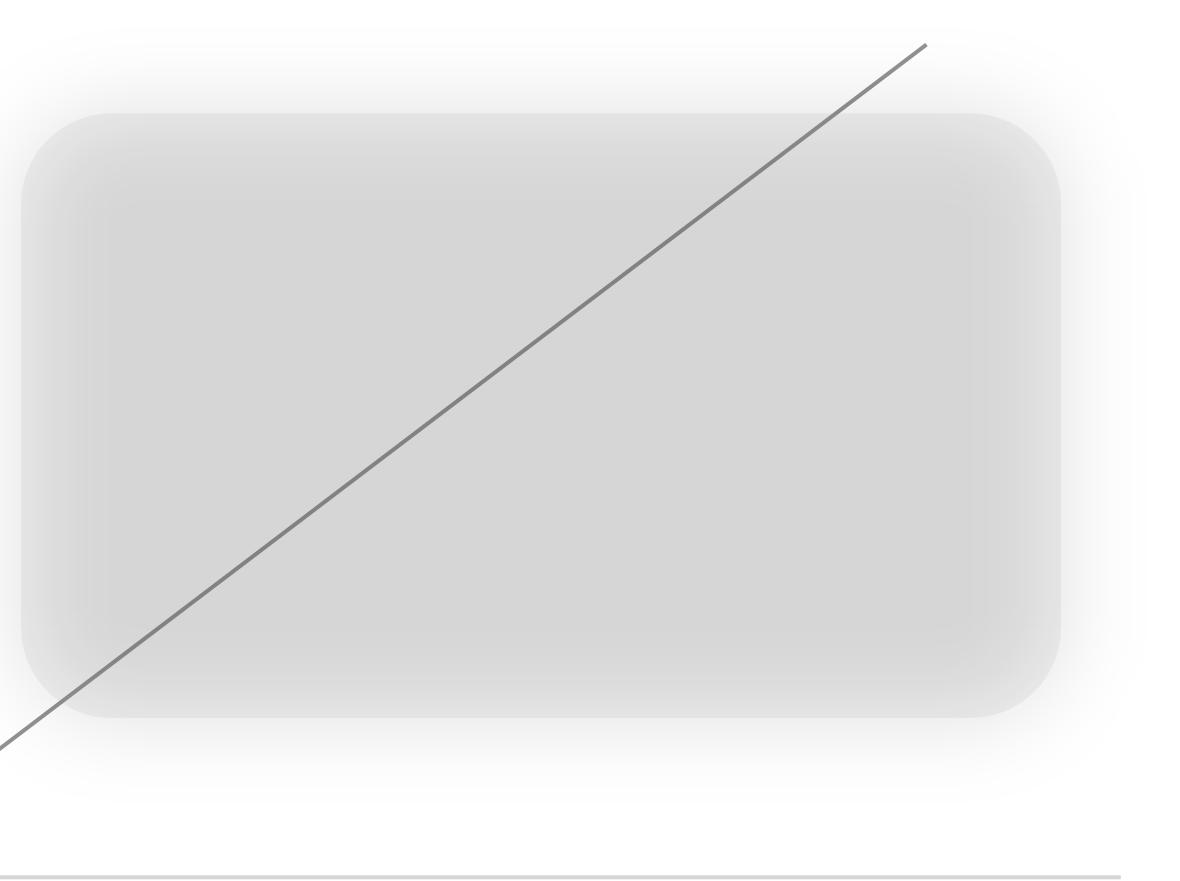


## models are

## functions

Quality

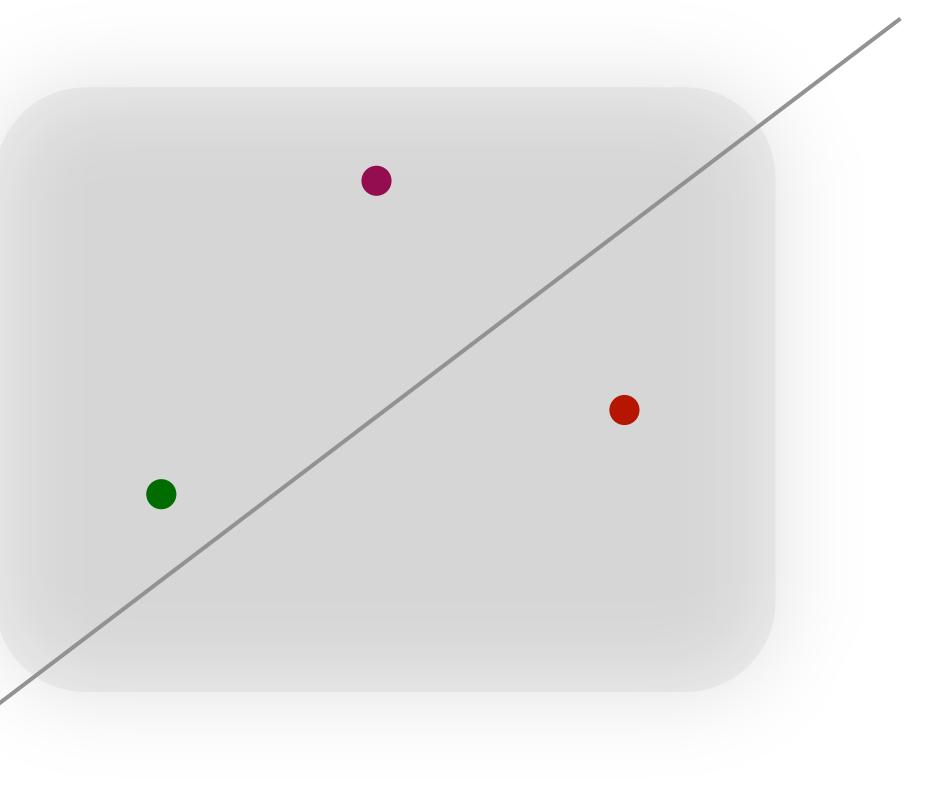
## that's linear(ish)\*



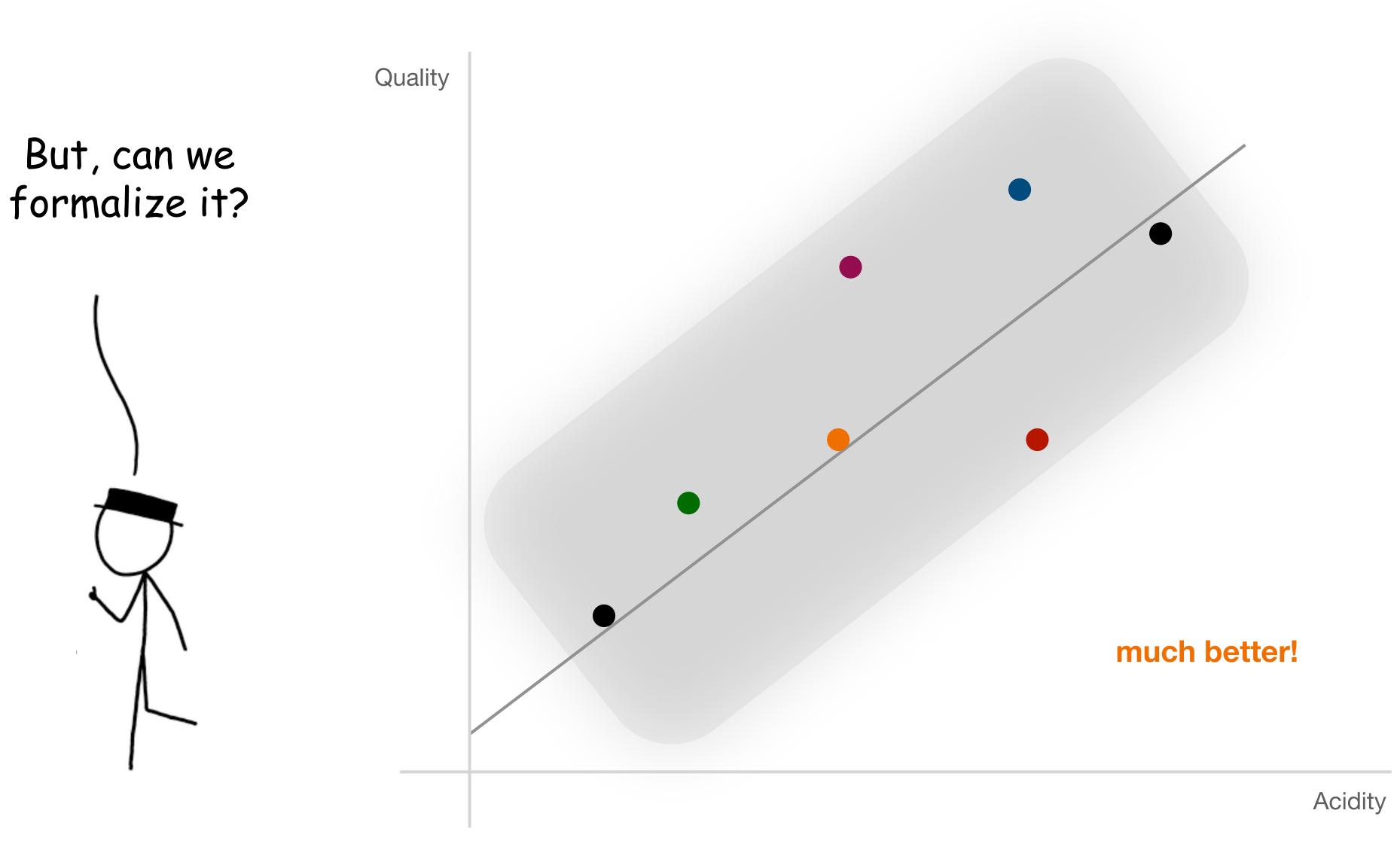
Acidity

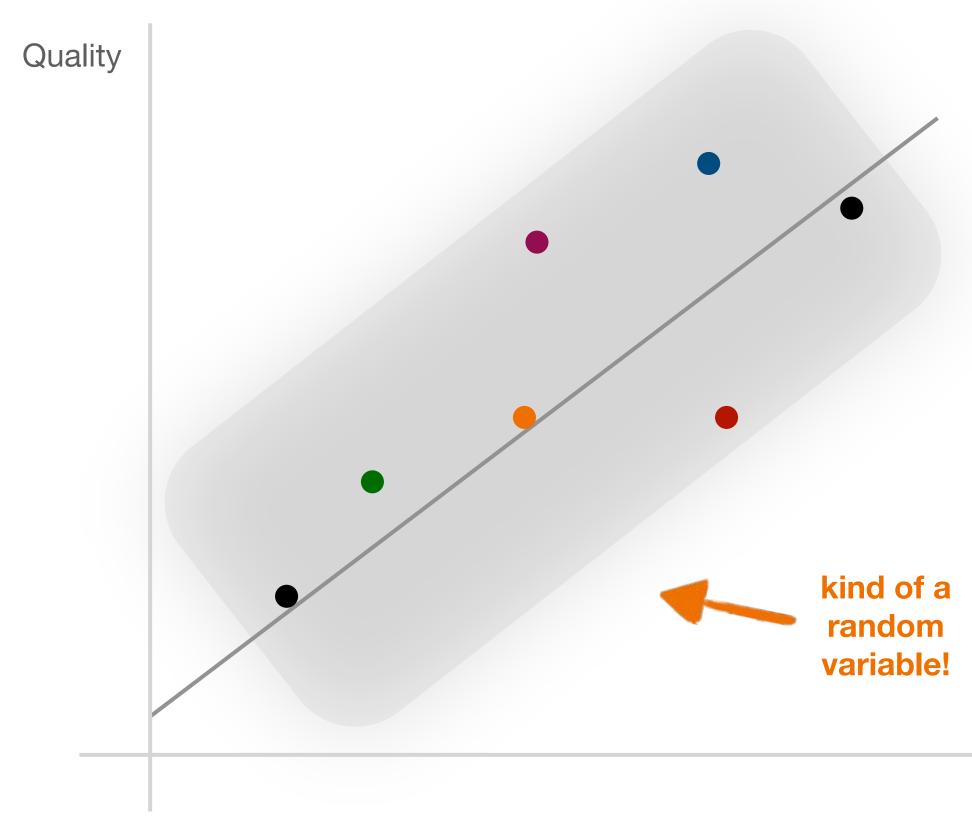


#### **??? still not super well described**

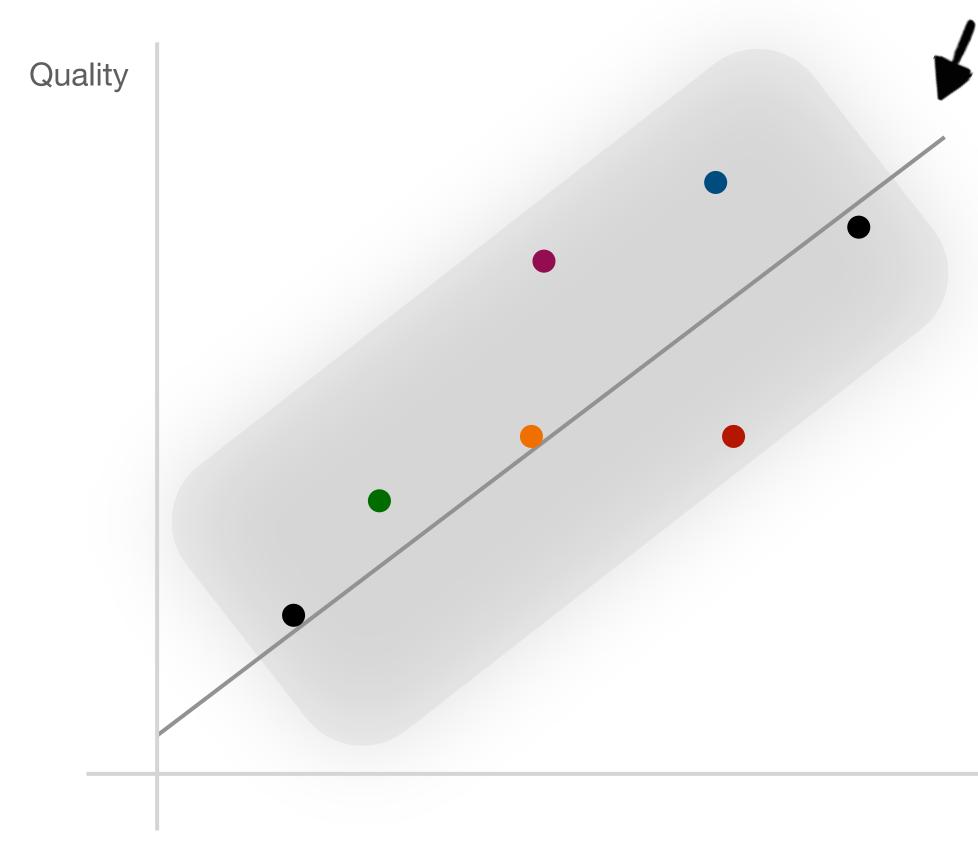


Acidity





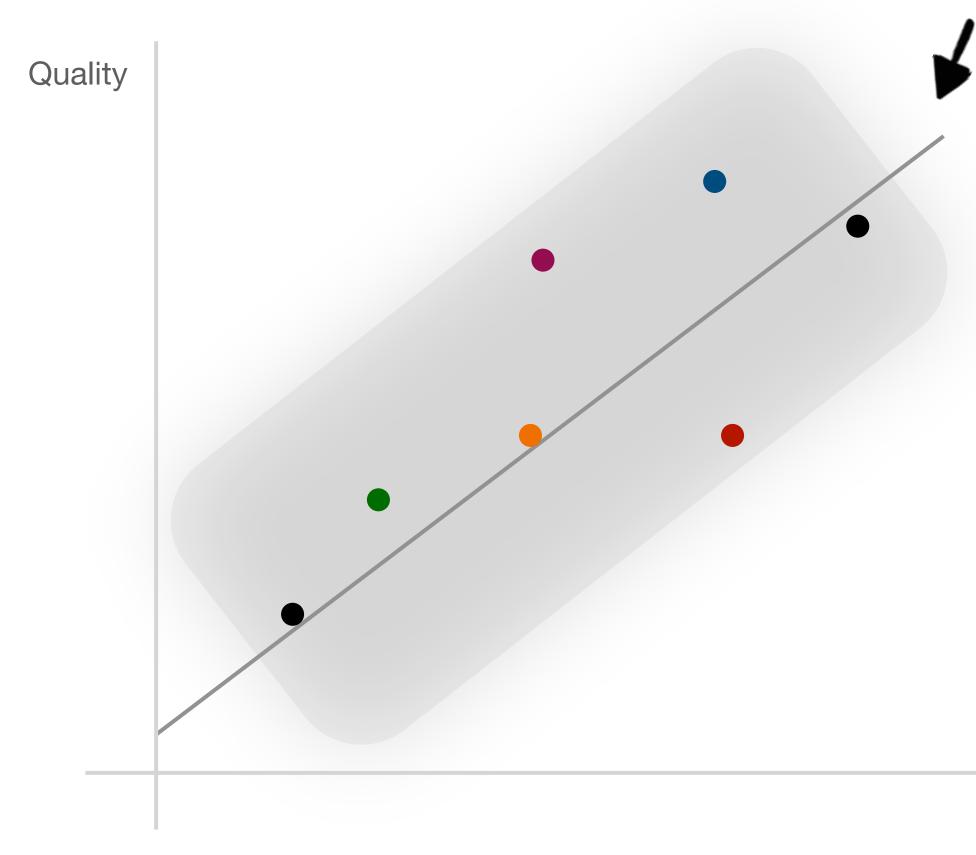
Acidity

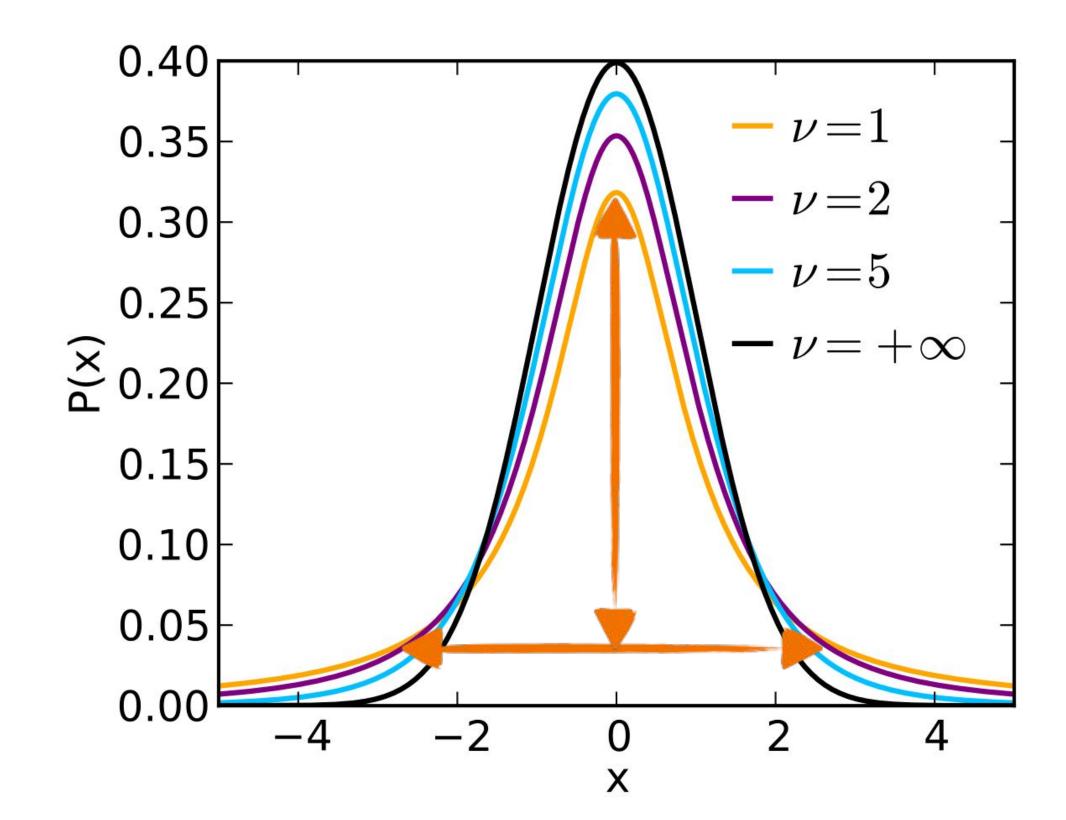


## wait... this is a t-test!

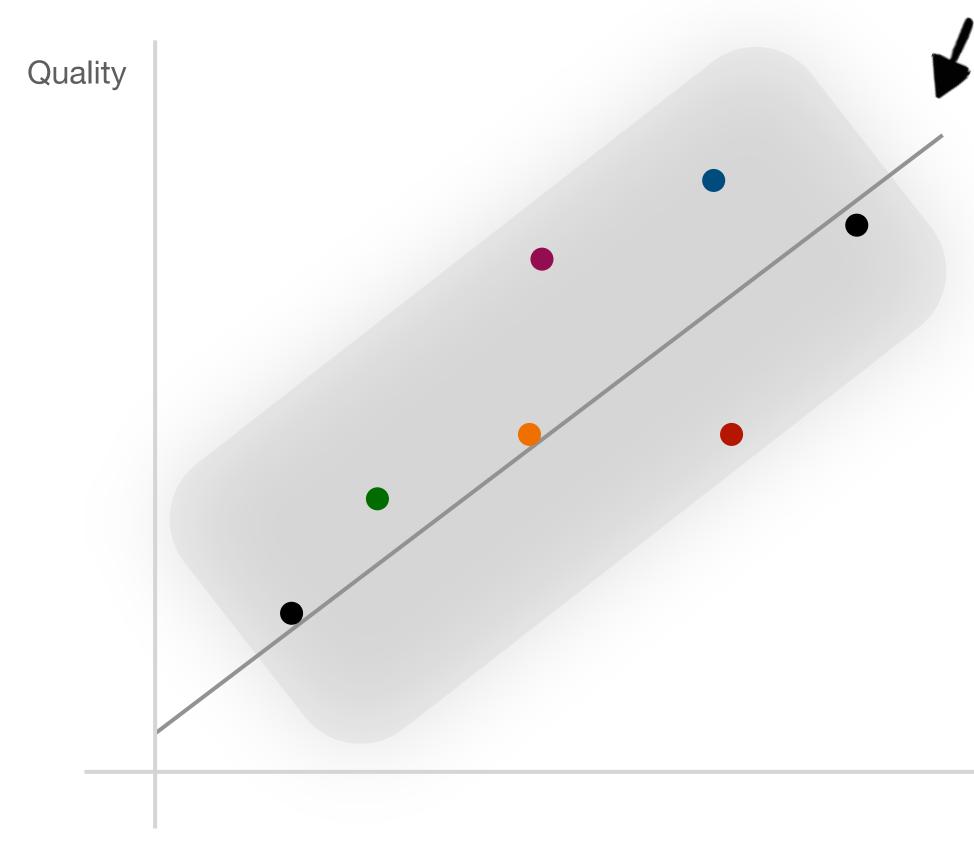
Acidity

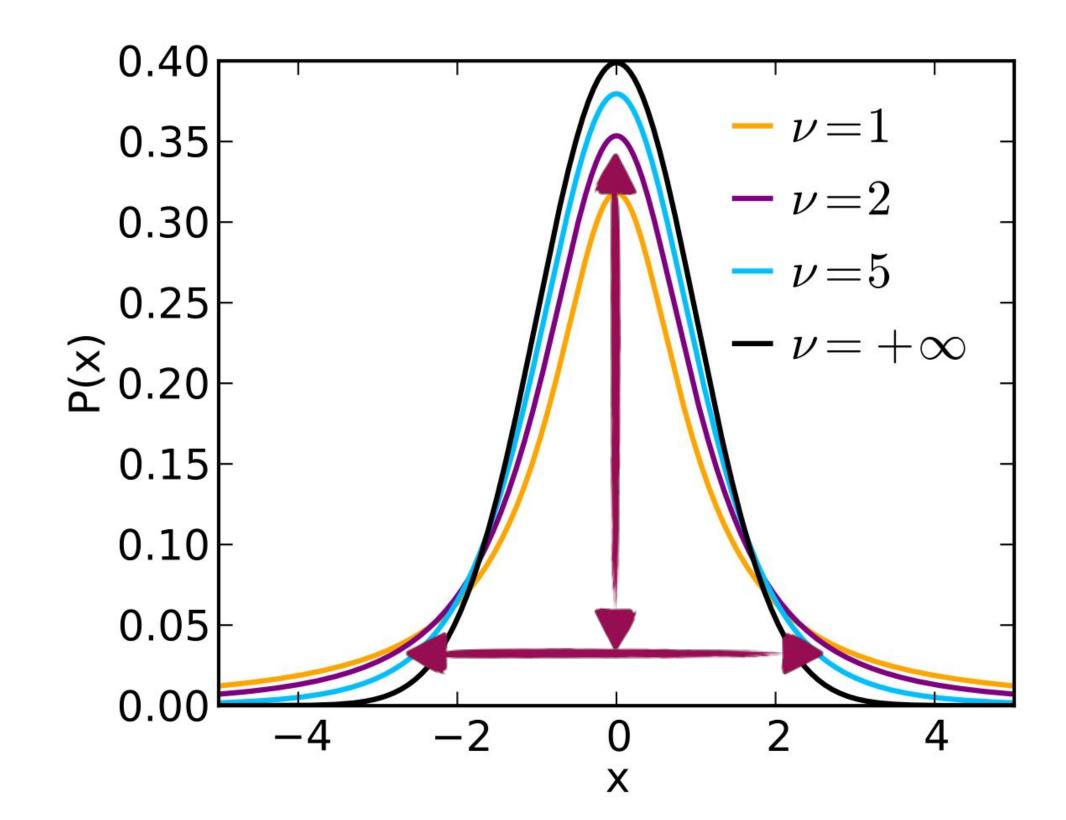
5 2



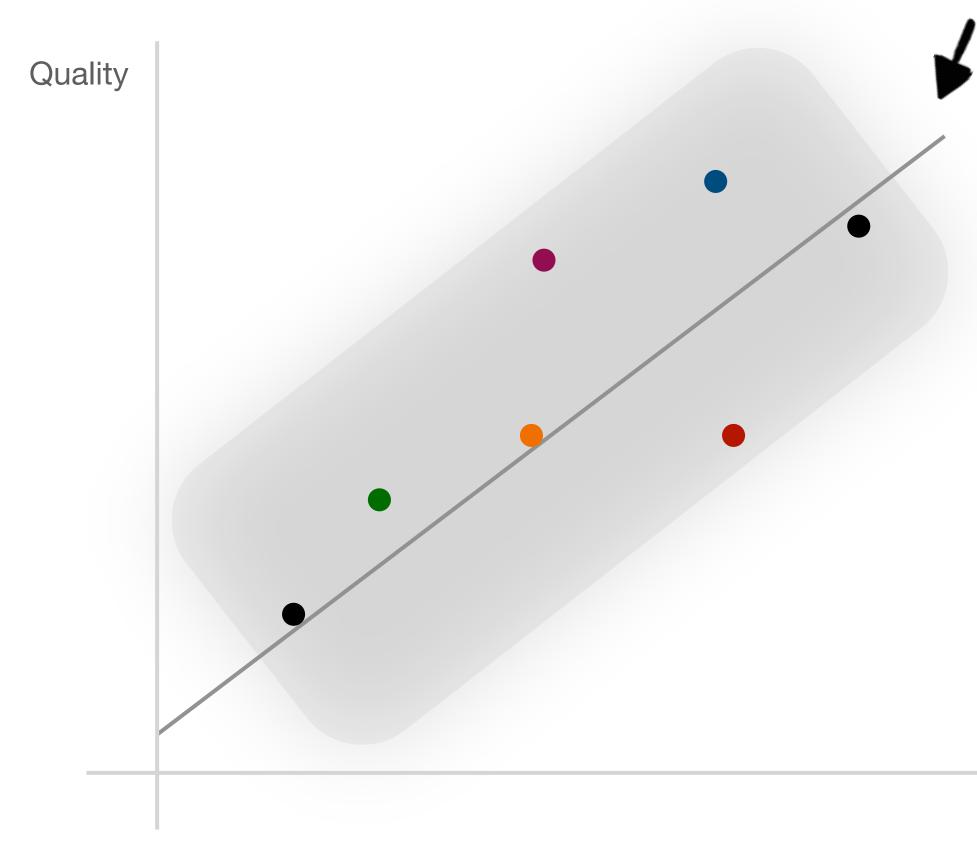


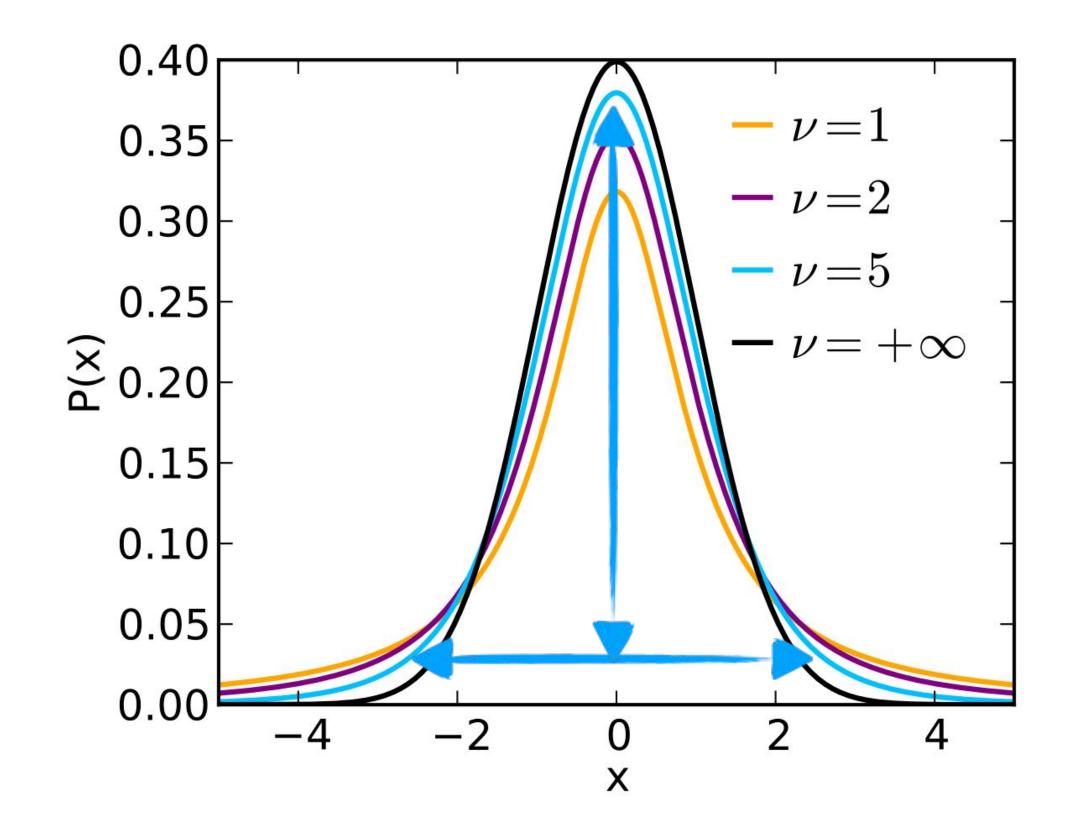




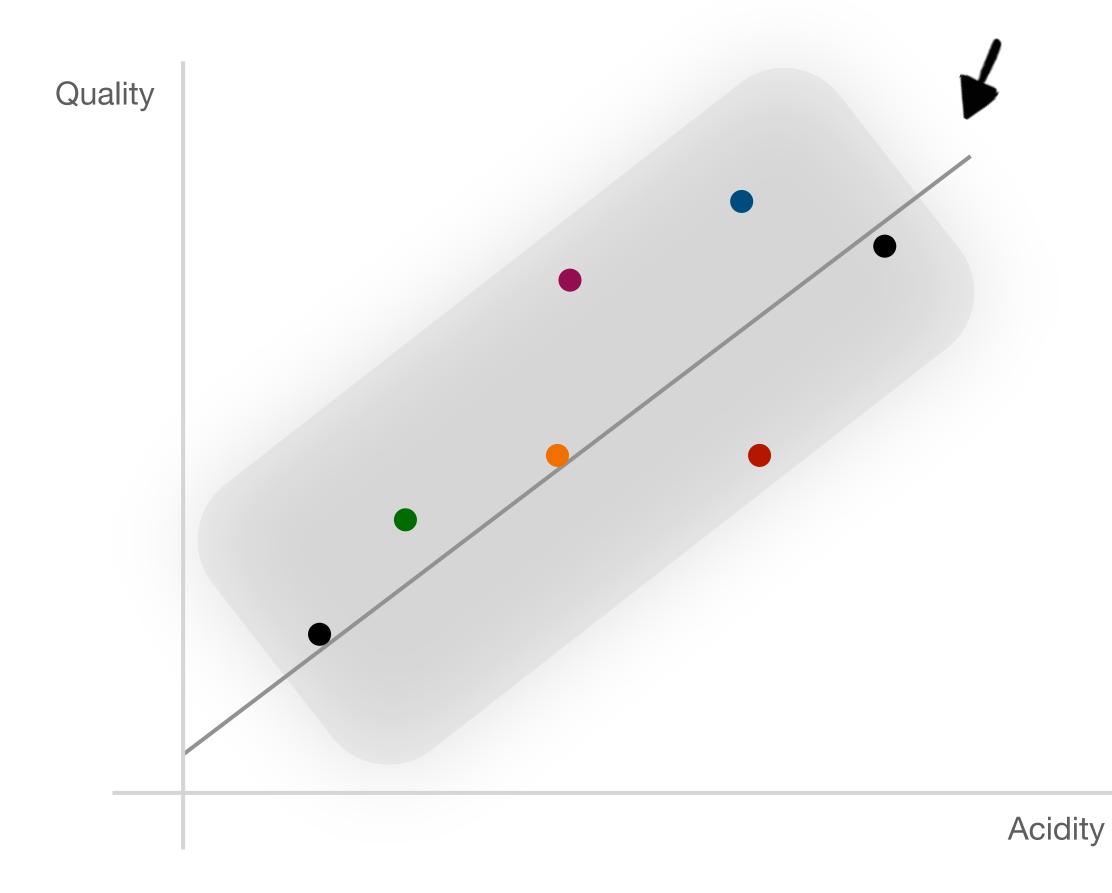




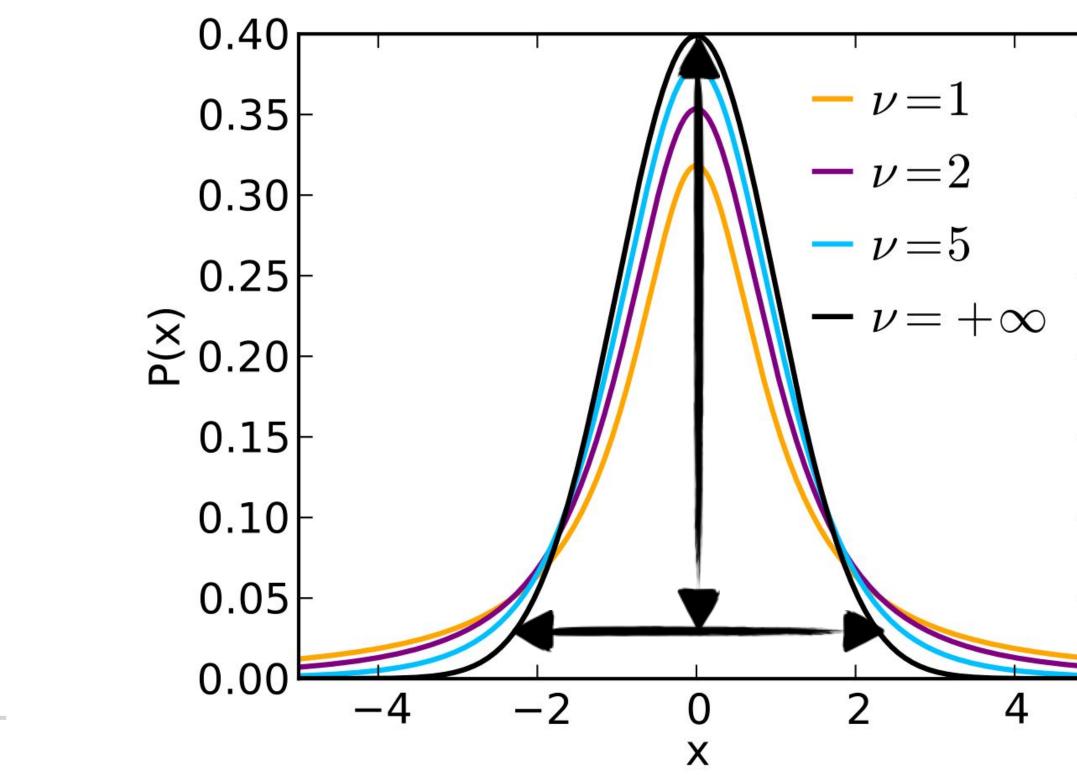




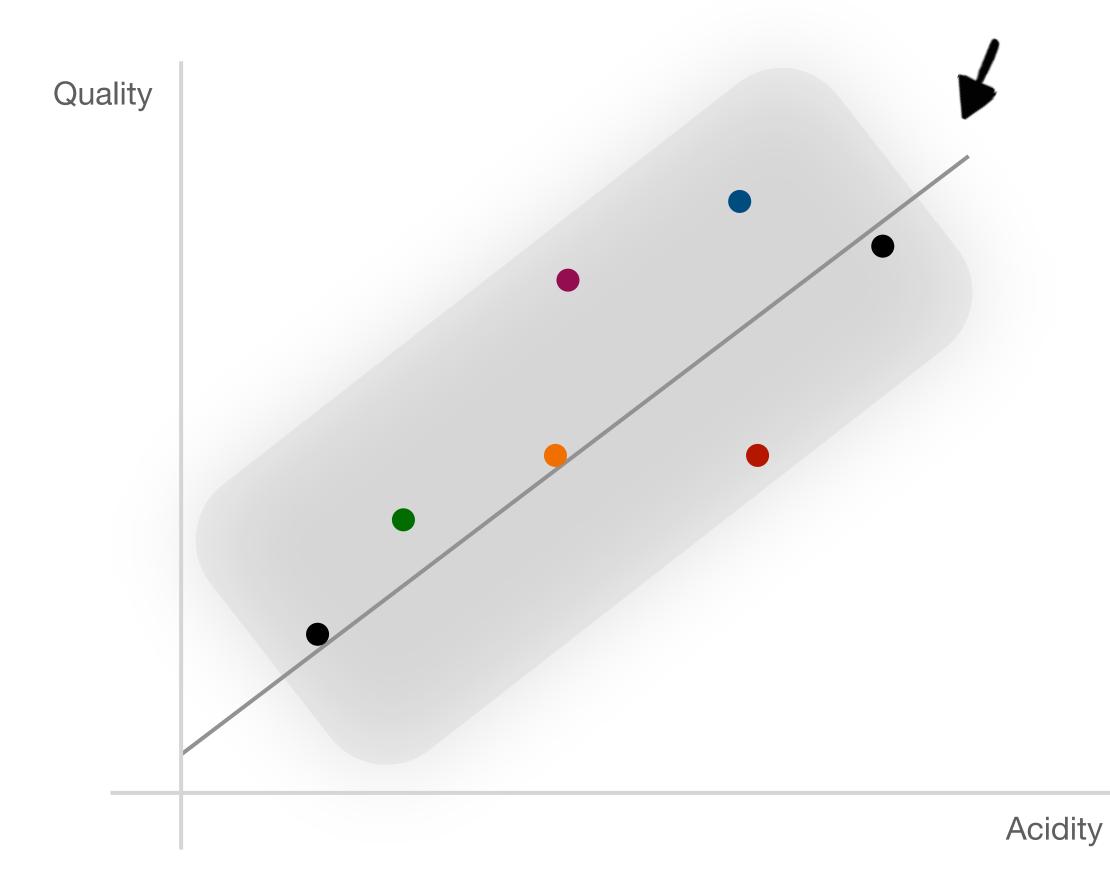




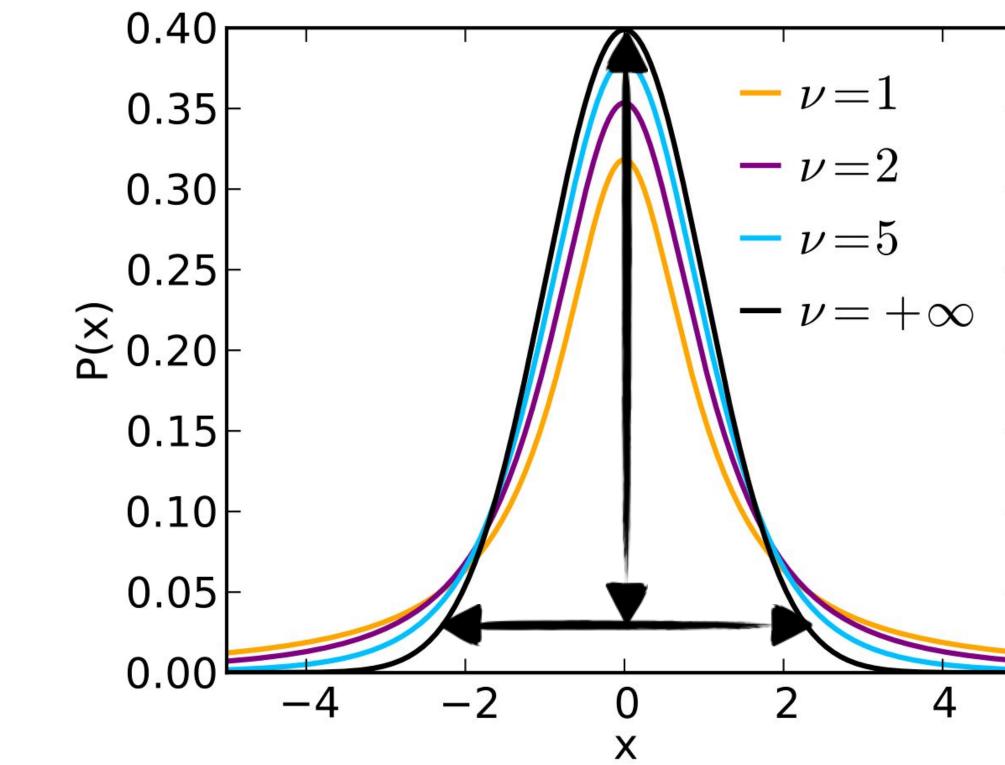
■ increased degrees-of-freedom increases the probability of the population equaling sample







increased degrees-of-freedom increases the probability of the population equaling sample
more data, better line





## balanced data normalized data quality data

## Let's think about logistic functions!

White

## Let's think about logistic functions!



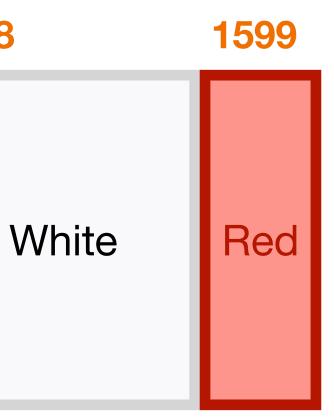
#### in an ideal world ...but no



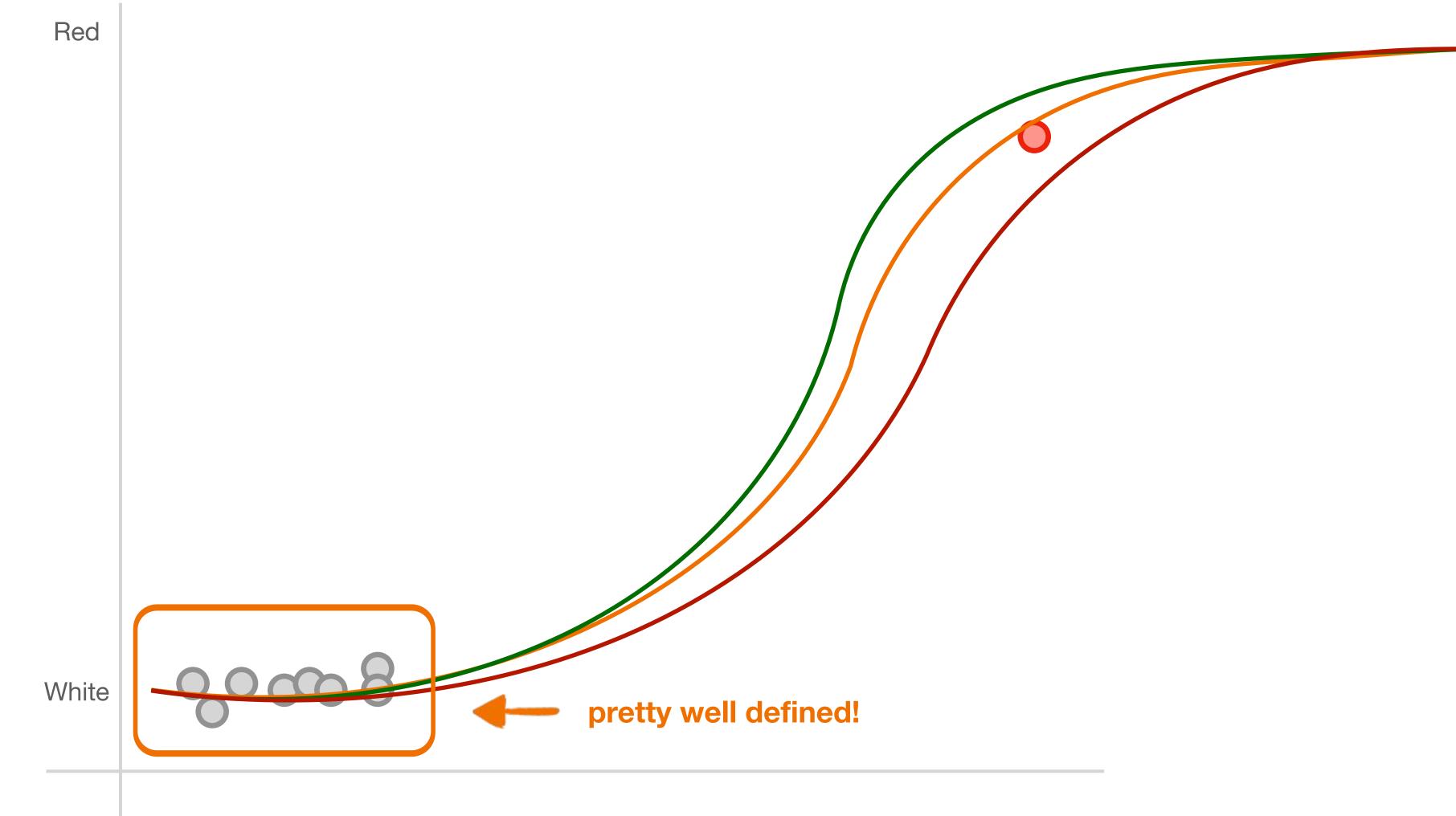


# What happens when we fit this dataset entirely?

## Let's think about logistic functions!



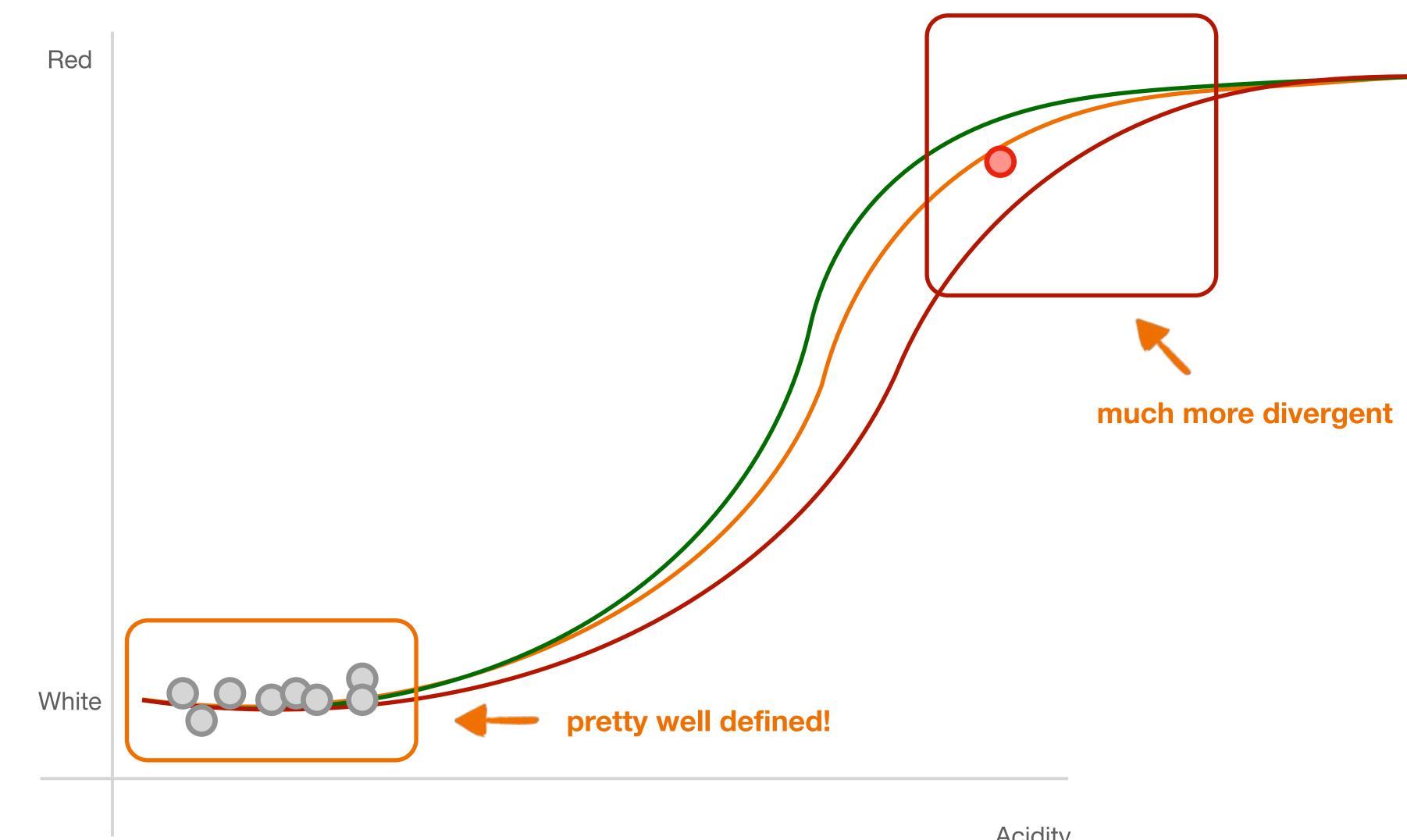




## Let's think about logistic functions!

Acidity





### Let's think about logistic functions!

Acidity

#### balanced data, more accurate results

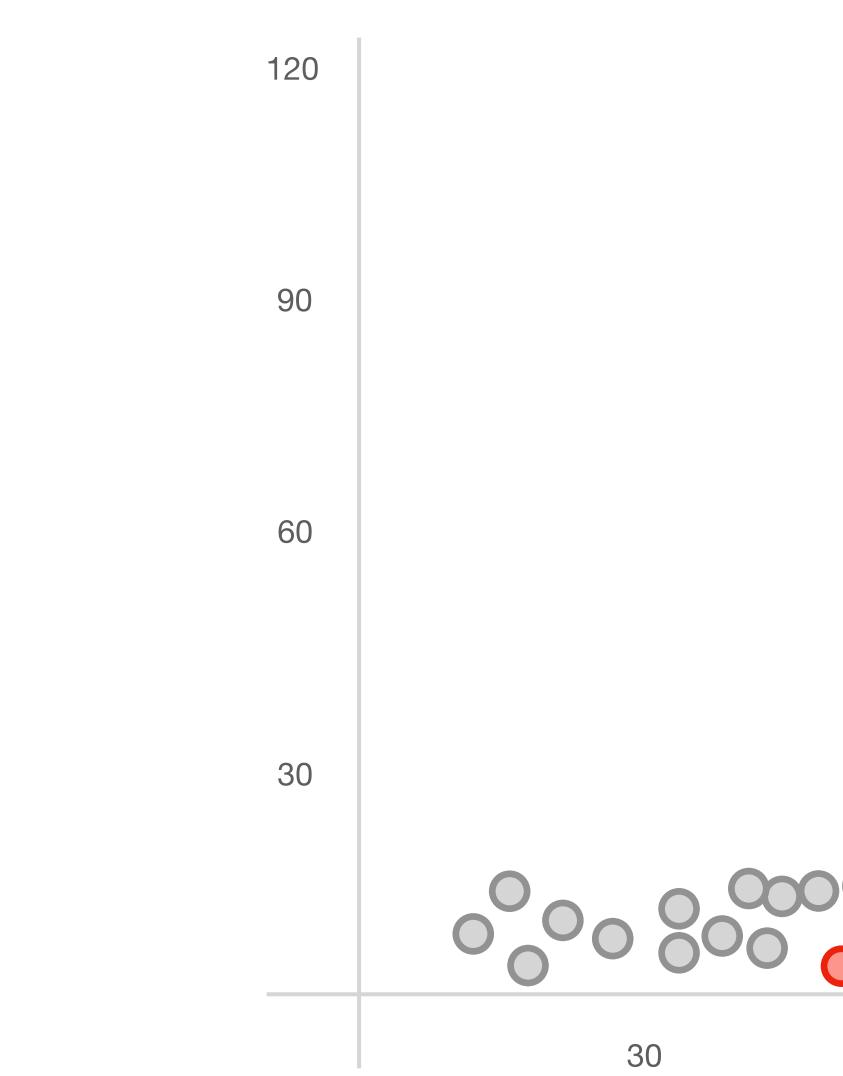


## balanced data normalized data quality data

## more data balanced data

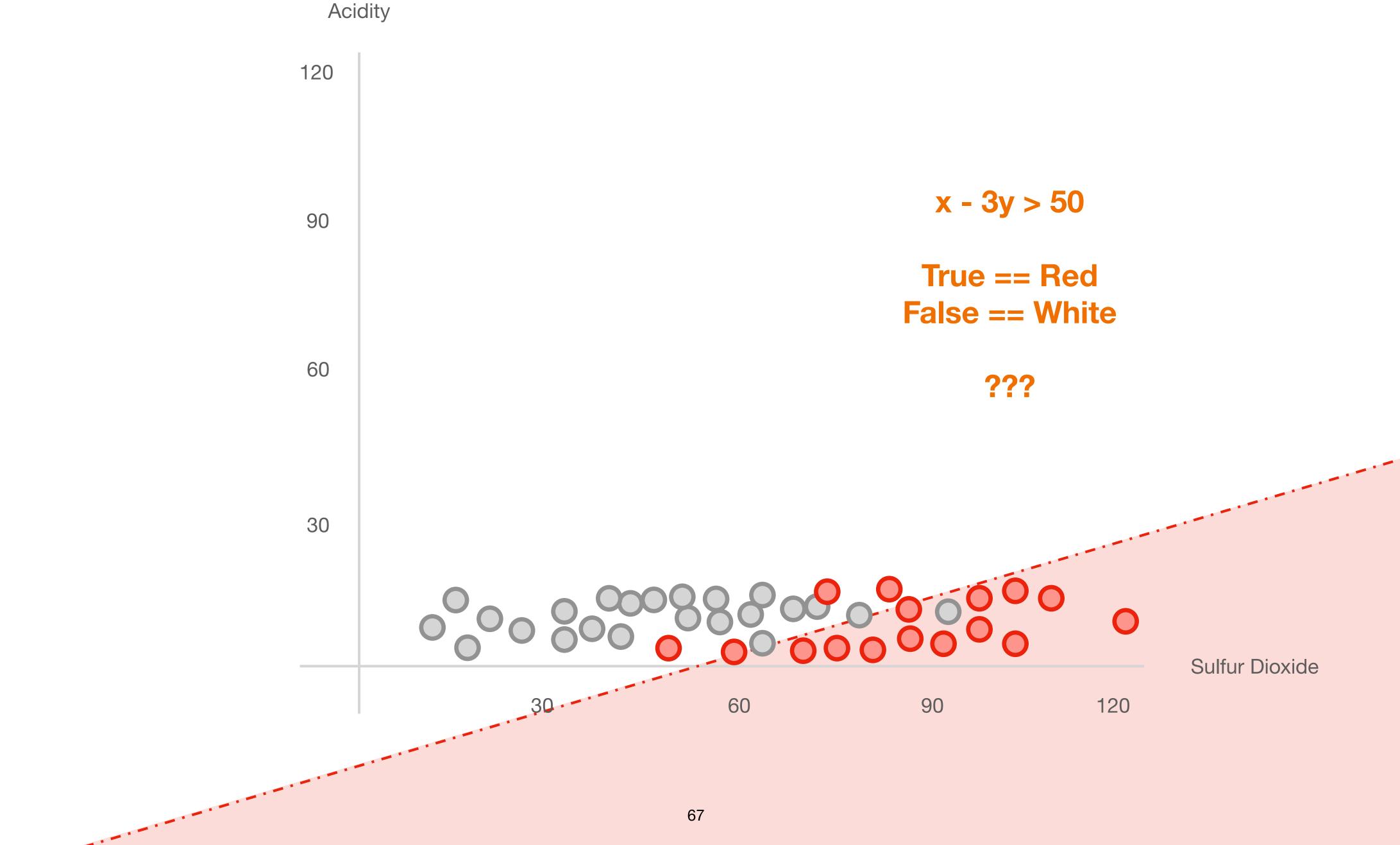
## normalized data

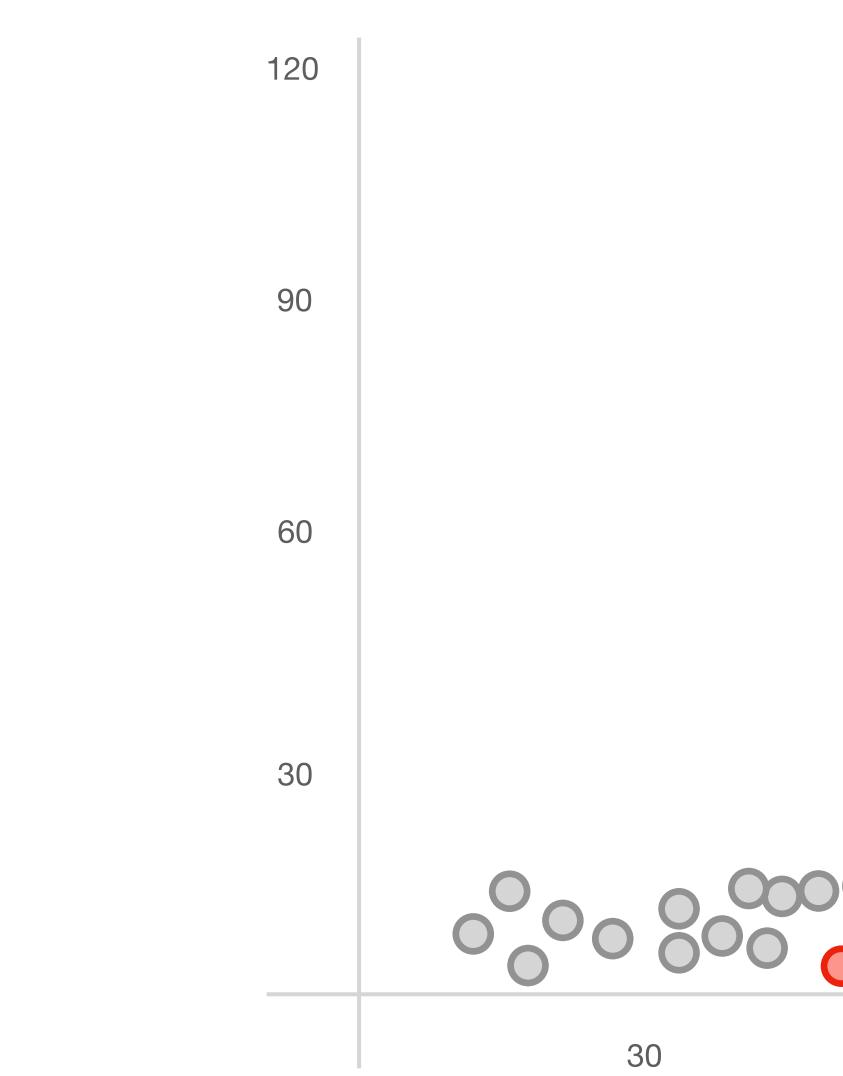
## quality data



Acidity

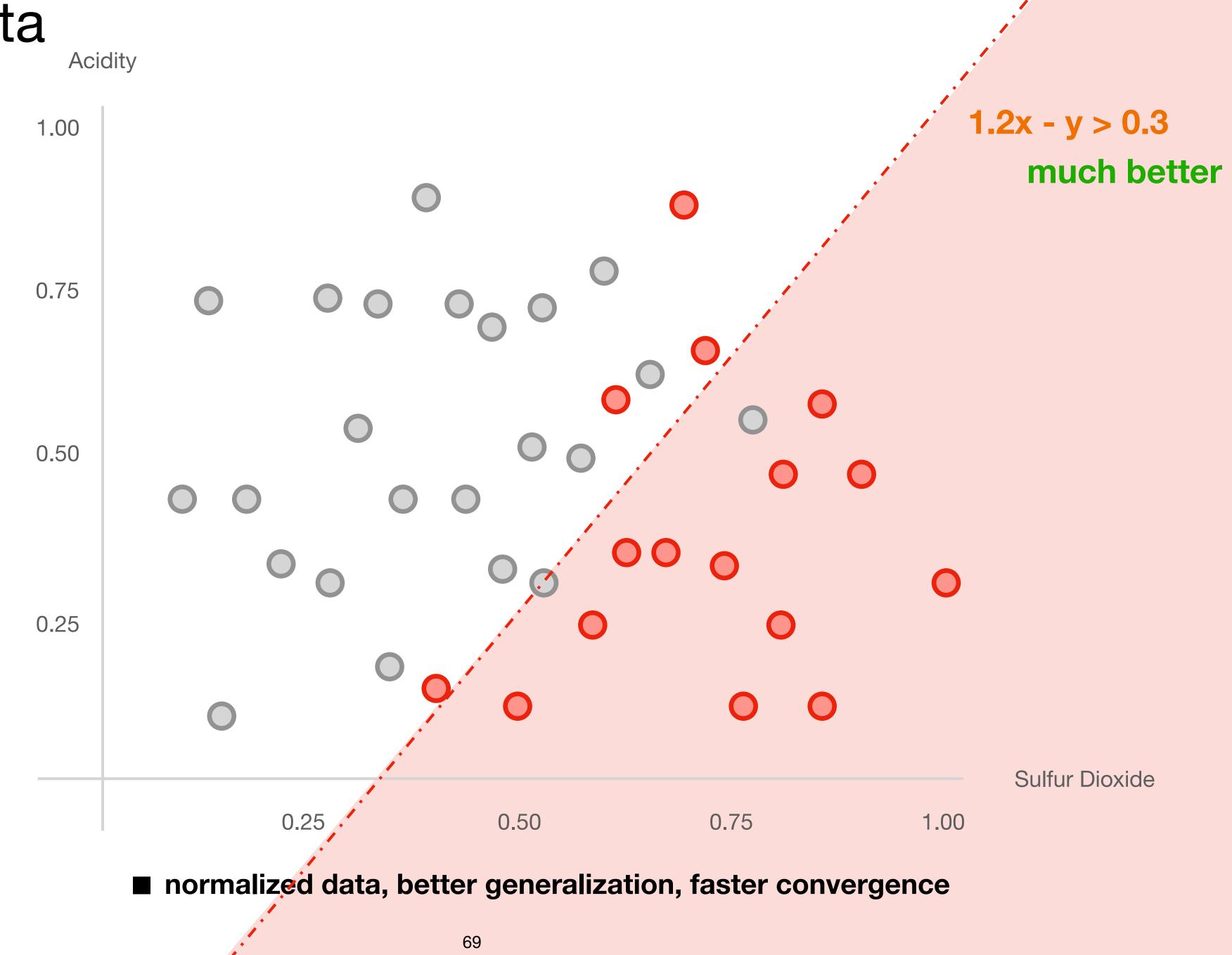
Sulfur Dioxide

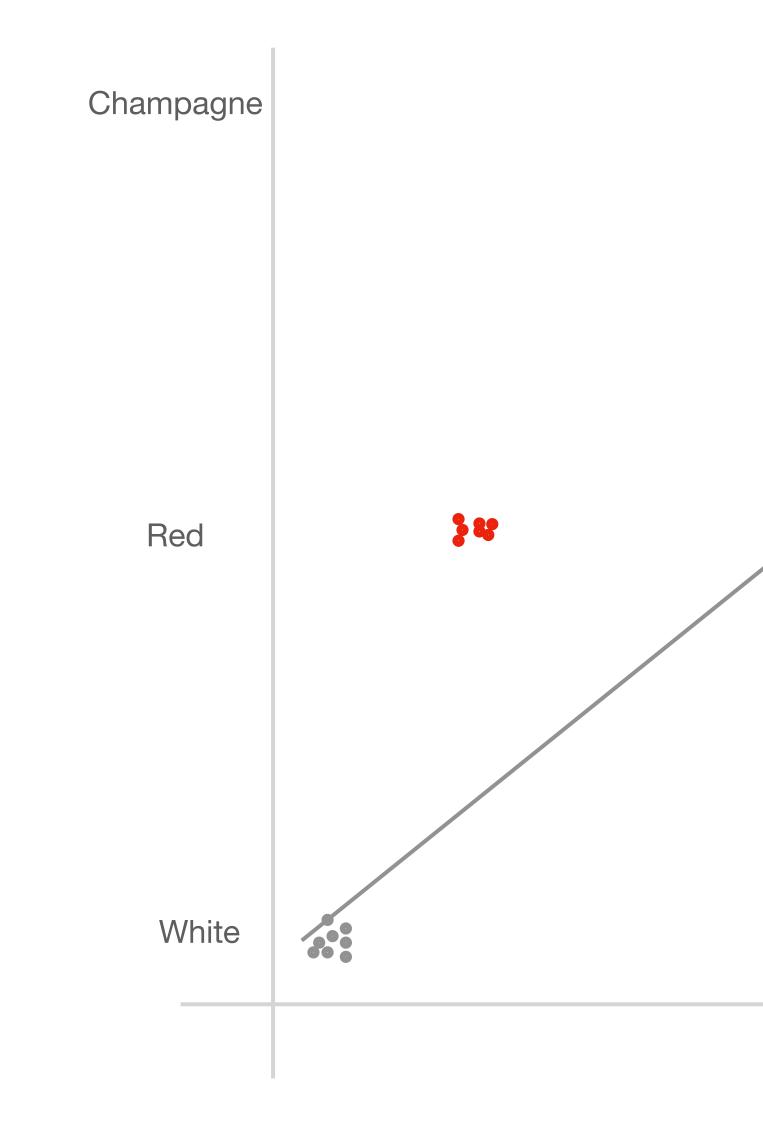




Acidity

68

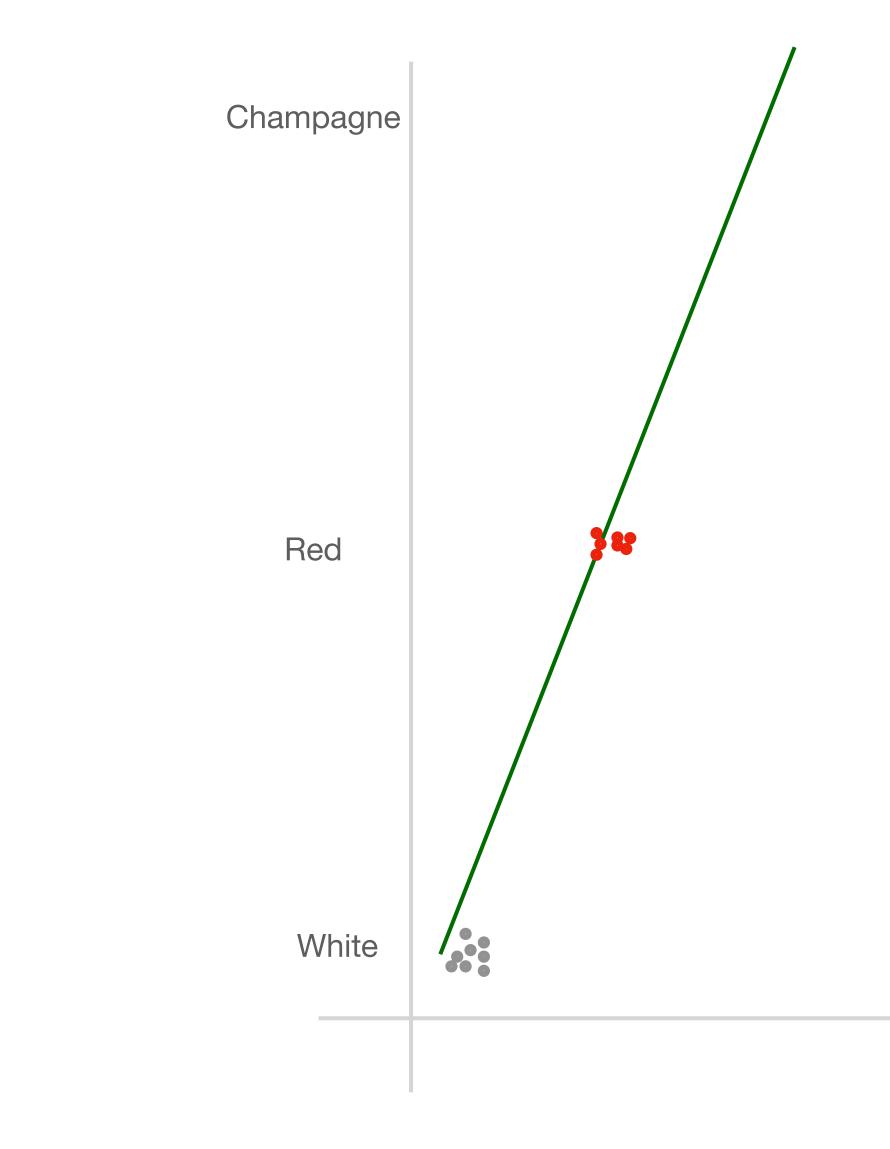




Acidity

**???** how to fit a line





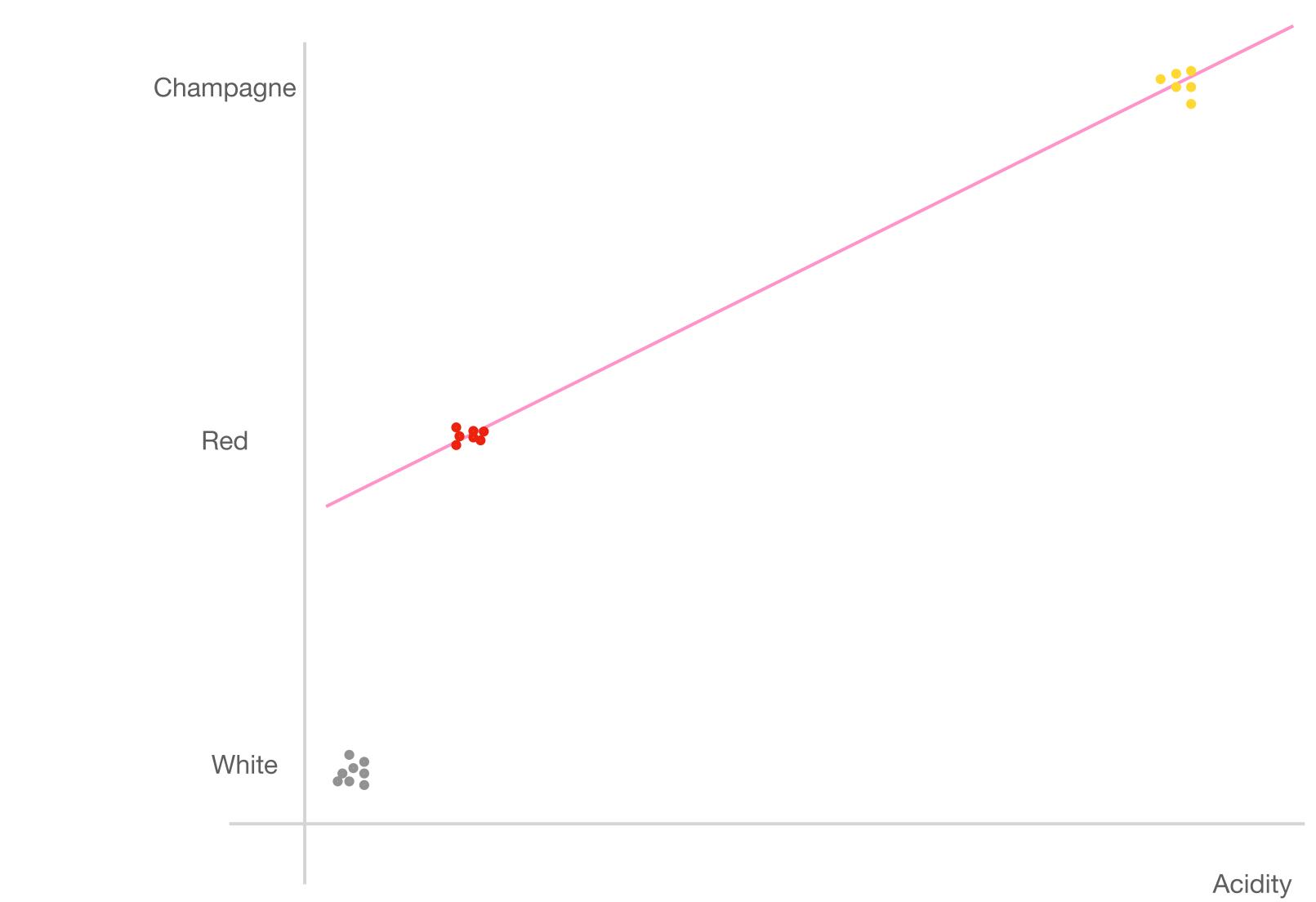


Acidity

**???** how to fit a line

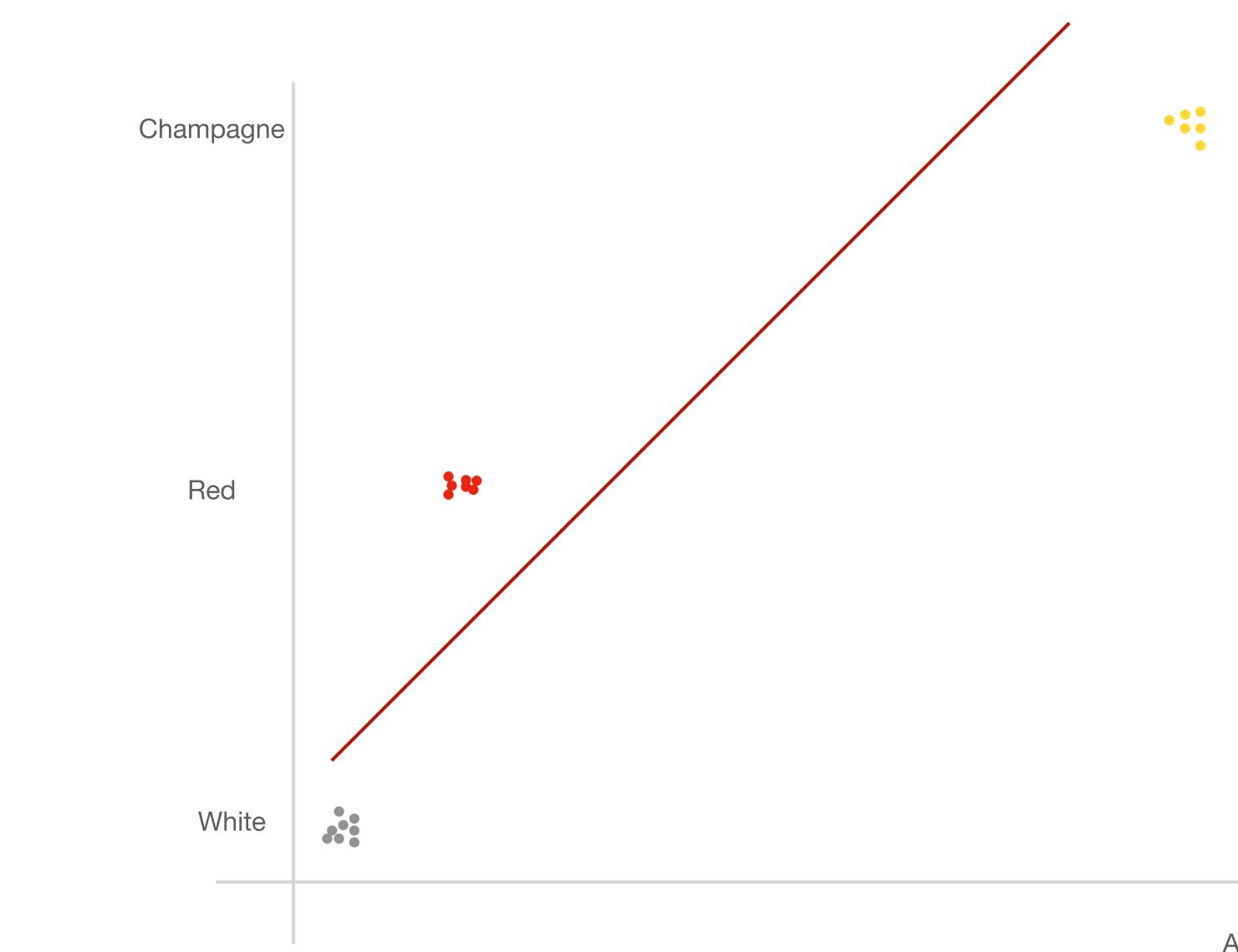
7 1





**???** how to fit a line

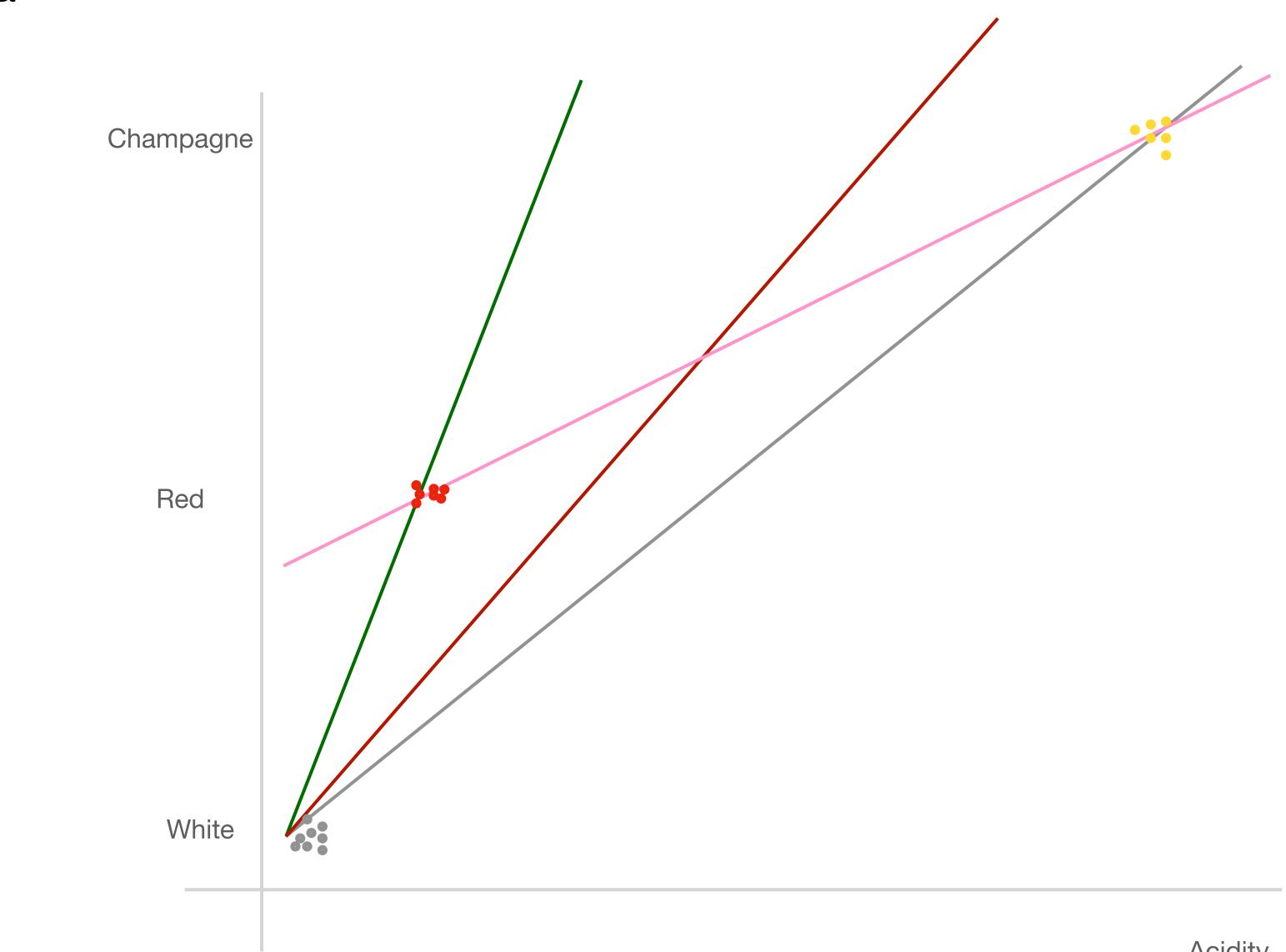




Acidity

**???** how to fit a line



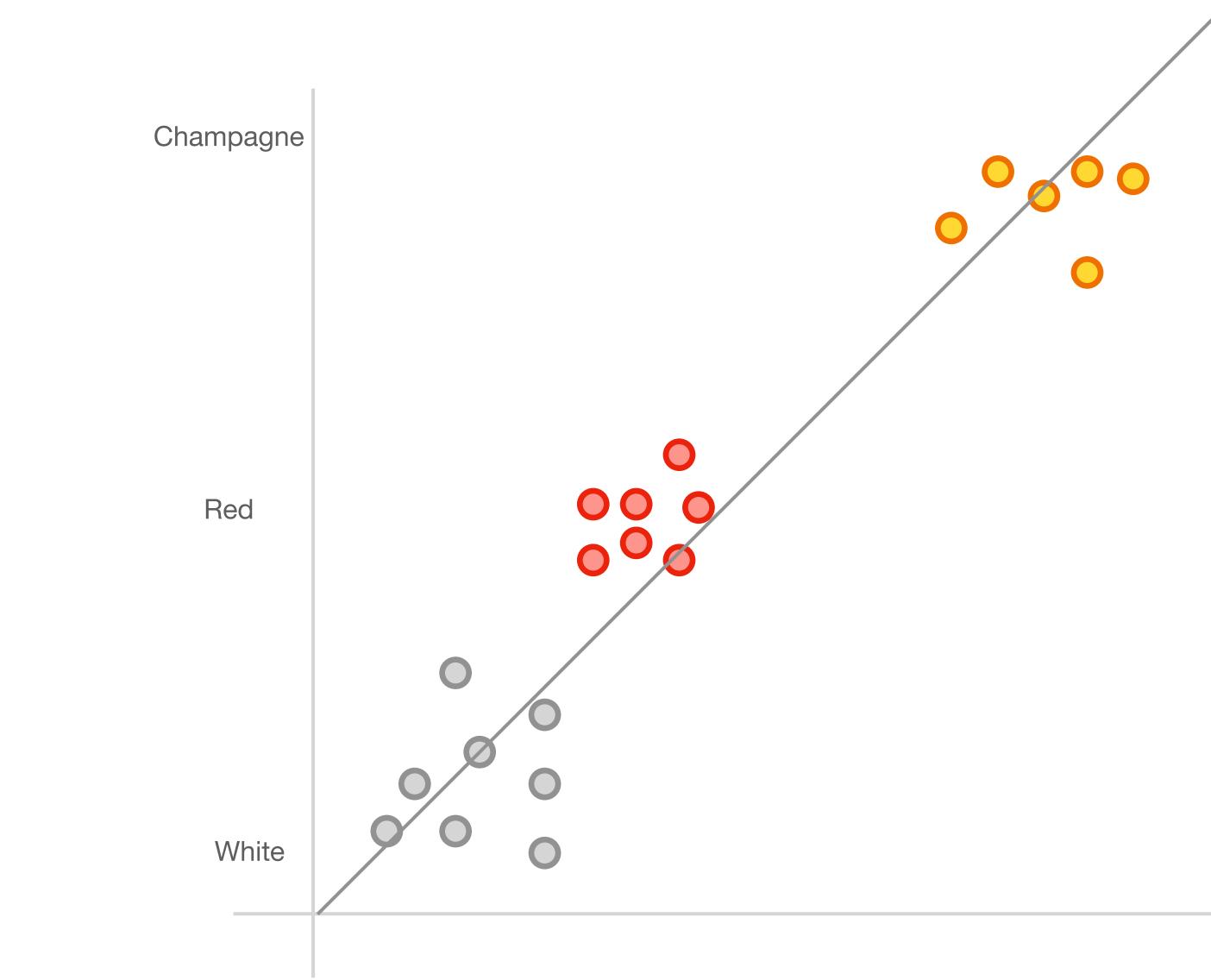




Acidity

#### non-normalized data is hard to fit

4



Acidity

#### ensure all features are internally normalized (same order of mag.)

5

## more data balanced data

## normalized data

## quality data

## more data balanced data normalized data

## quality data



Image credit: Passionned Group

## more data balanced data normalized data

## quality data

Missing Data

# Missing Data

## Missing completely at random

## Missing at random

## Missing not at random



# Missing Data

#### remove

## Use mean/most often

## regression

more data

normalized data

let's clean some data!