#### Reminder: Multiple Linear Regression

Test-train data split to avoid overfitting

#### Multiple Linear Regression Model

$$Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + ... \beta_k x_k + \varepsilon$$

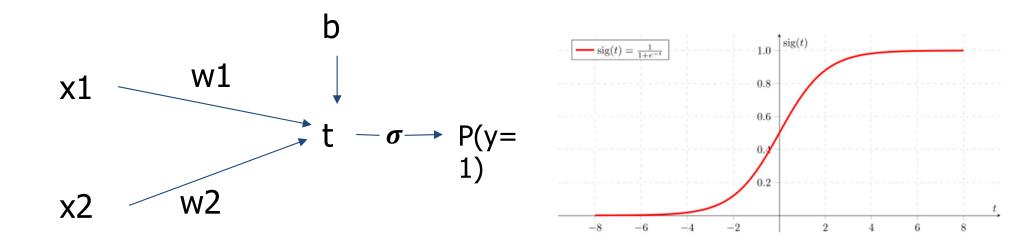
One can also use powers and products of other variables or even non-linear functions like  $exp(x_i)$  or  $log(x_i)$  instead of  $x_3, \ldots, x_k$ .

Example: the general two-variable quadratic regression has 6 constants:

$$Y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 (x_1)^2 + \beta_4 (x_2)^2 + \beta_5 (x_1 x_2) + \varepsilon$$

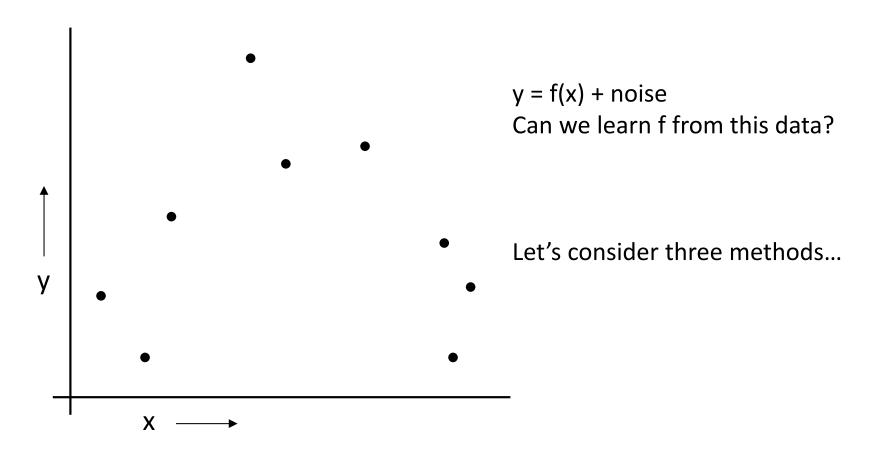
#### **Logistic Regression**

$$P(y=1) = \sigma(x1*w1 + x2*w2 + b)$$

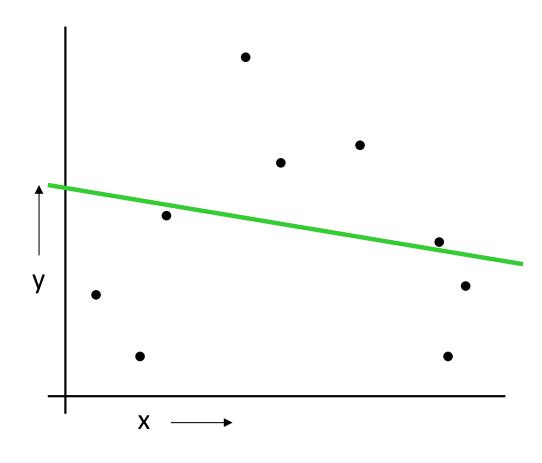


How to know when to stop adding new variables or model parameters in any data fitting algorithm such as multiple linear regression?

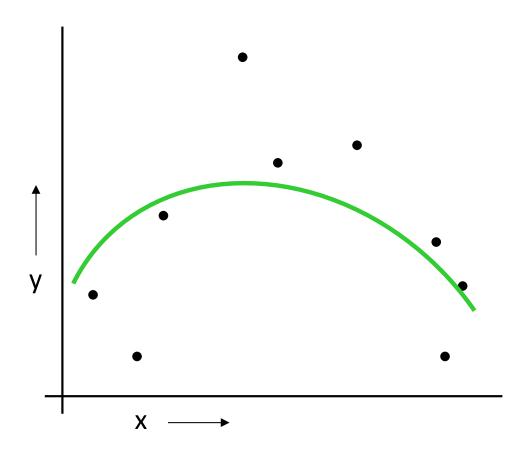
#### A Regression Problem



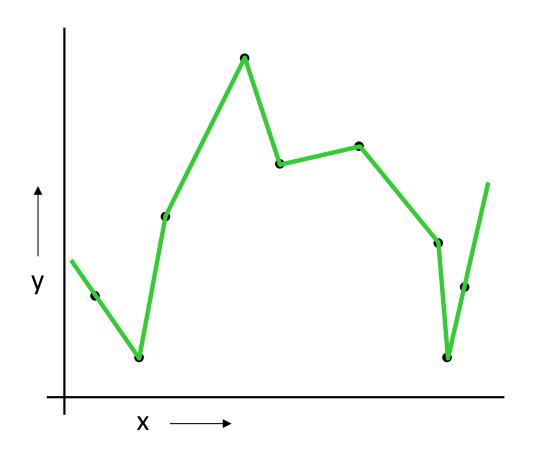
#### Single Variable Linear Regression



#### 2-variable Linear Regression with x and x<sup>2</sup>

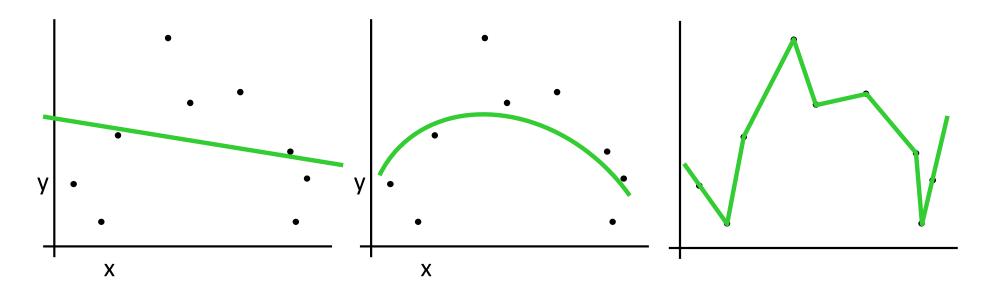


#### Join-the-dots



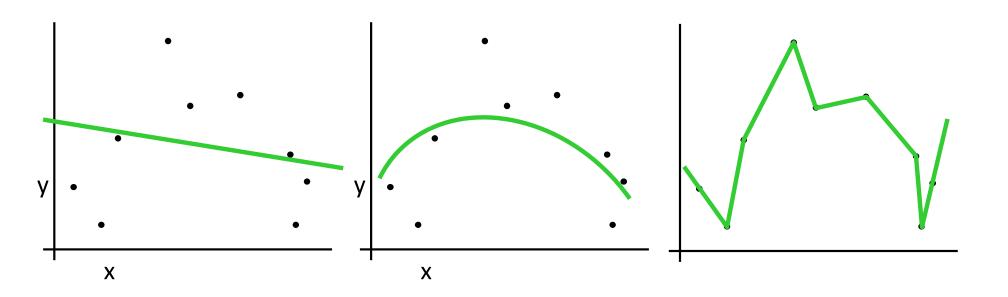
Also known as piecewise linear nonparametric regression if that makes you feel better

#### Which is best?



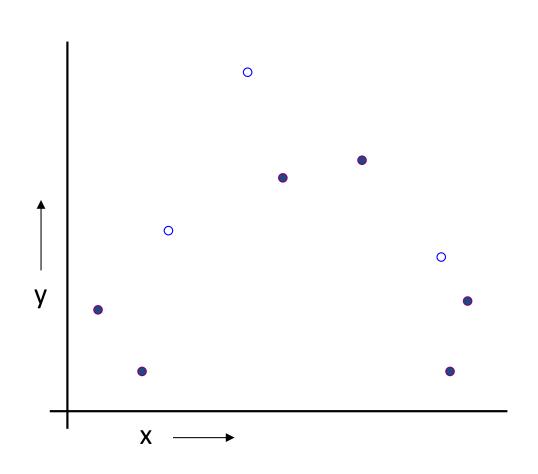
Why not choose the method with the best fit to the data?

#### What do we really want?

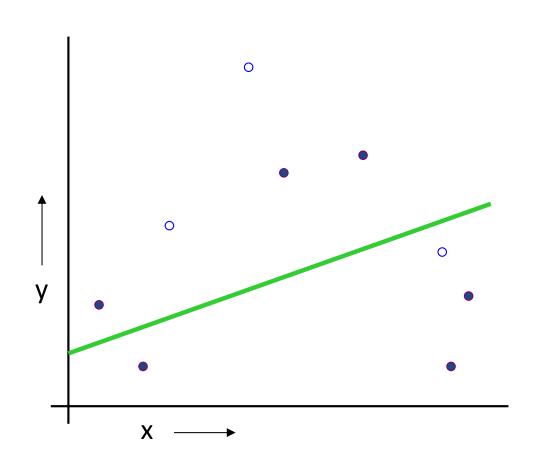


Why not choose the method with the best fit to the data?

"How well are you going to predict future data drawn from the same distribution?"

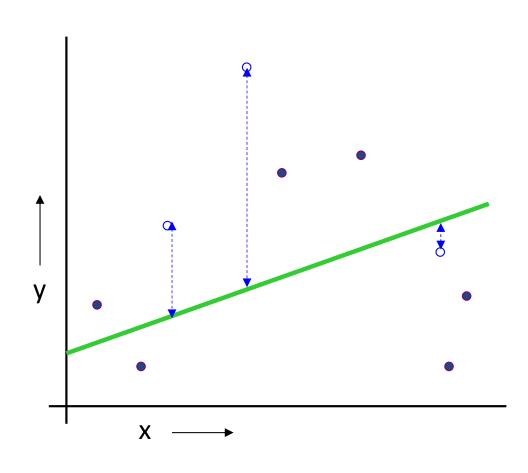


Randomly choose
 of the data to
 in a test set
 The remainder is a training set



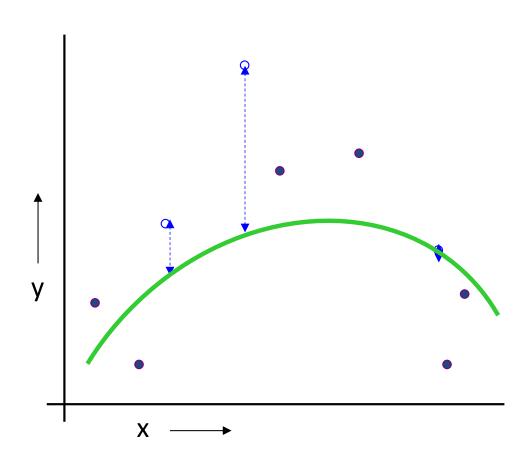
(Linear regression example)

- 1. Randomly choose 30% of the data to be in a test set
- 2. The remainder is a training set
- 3. Perform your regression on the training set



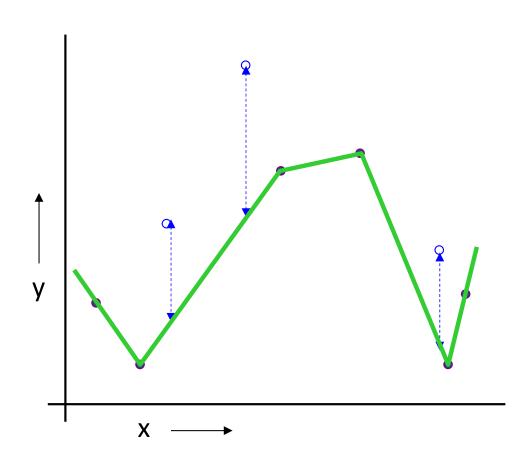
(Linear regression example) Mean Squared Error = 2.4

- 1. Randomly choose 30% of the data to be in a test set
- 2. The remainder is a training set
- 3. Perform your regression on the training set
- 4. Estimate your future performance with the test set



(Quadratic regression example) Mean Squared Error = 0.9

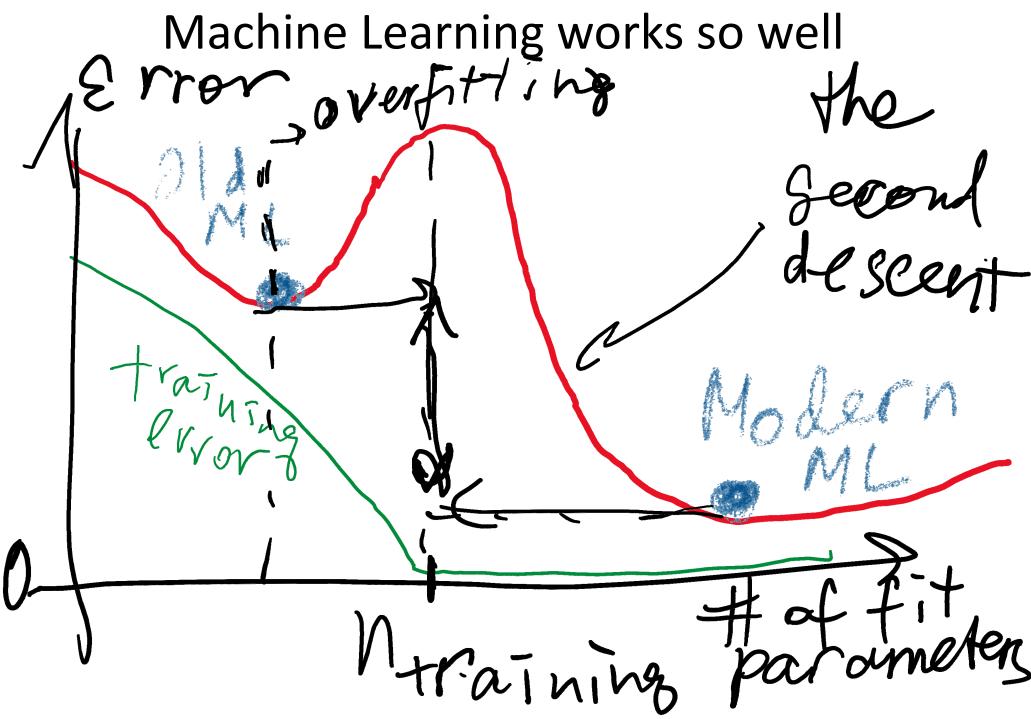
- 1. Randomly choose 30% of the data to be in a test set
- 2. The remainder is a training set
- 3. Perform your regression on the training set
- 4. Estimate your future performance with the test set



(Join the dots example)
Mean Squared Error = 2.2

- 1. Randomly choose 30% of the data to be in a test set
- 2. The remainder is a training set
- 3. Perform your regression on the training set
- 4. Estimate your future performance with the test set

Double descend- the main reason modern



#### R<sup>2</sup> and Adjusted R<sup>2</sup>

#### The coefficient of multiple determination R<sup>2</sup>

$$R^2 = \frac{SS_R}{SS_T} = 1 - \frac{SS_E}{SS_T}$$

The adjusted  $R^2$  is

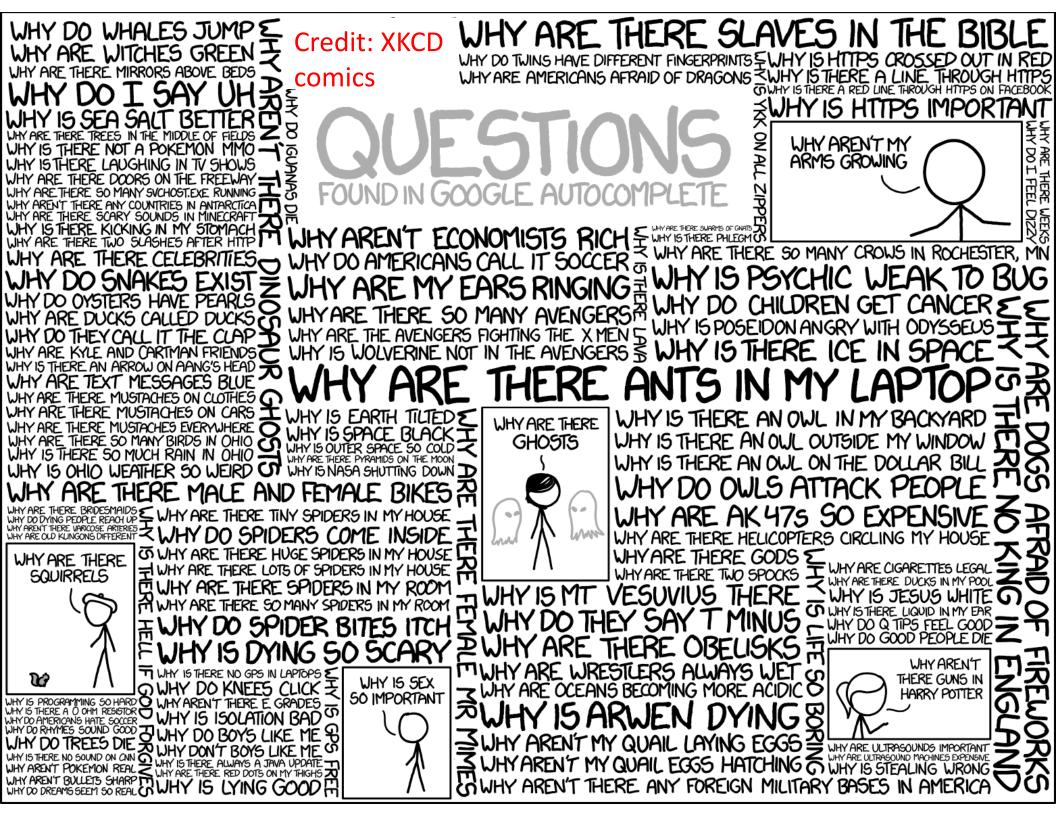
$$R_{\text{adj}}^2 = 1 - \frac{SS_E/(n-p)}{SS_T/(n-1)}$$

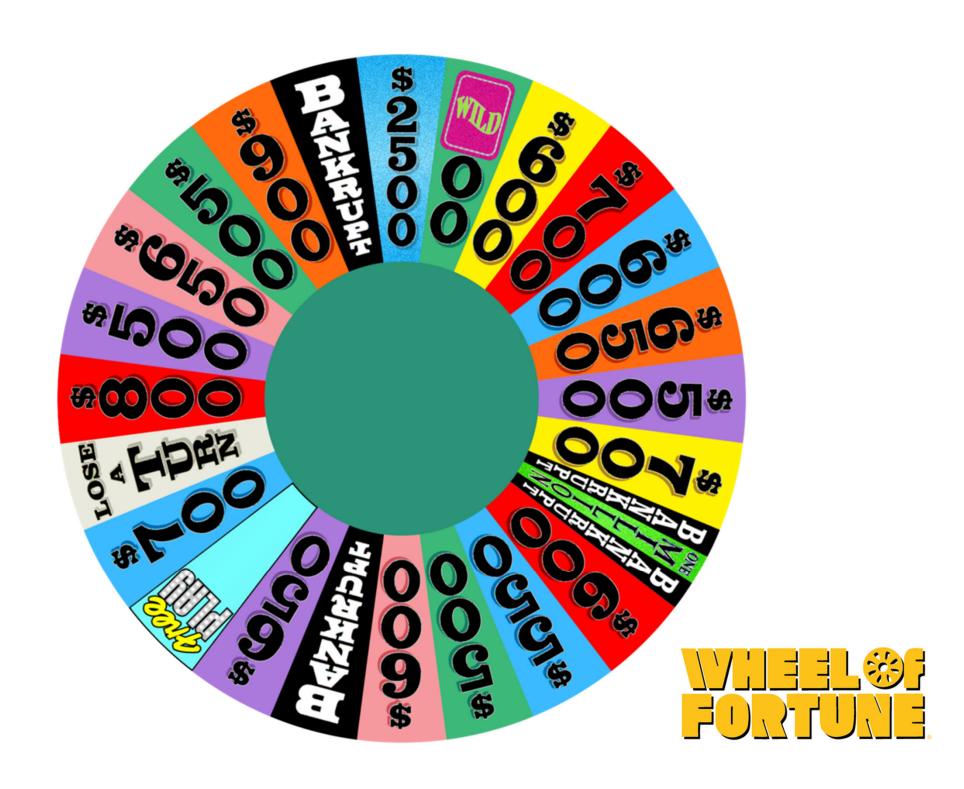
- The adjusted R<sup>2</sup> statistic penalizes adding terms to the MLR model.
- It can help guard against overfitting (including regressors that are not really useful)

### How to know where to stop adding variables?

 Adding new variables x<sub>i</sub> to MLR watch the adjusted R<sup>2</sup>

Once the adjusted R<sup>2</sup>
 no longer increases = stop.
 Now you did the best you can.





#### Matlab exercise #1: "Wheel of Fortune"

- Each group gets a pair of genes that are known to be correlated.
- Each group also gets a random pair of genes selected by the "Wheel of Fortune". They may or may not be correlated
- Download (log-transformed) expression\_table.mat
- Run command fitlm(x,y) on assigned and random pairs
- Record  $\beta_0$ ,  $\beta_1$ ,  $R^2$ , P-value of the slope  $\beta_1$  and write them on the blackboard
- Validate Matlab result for R<sup>2</sup> using your own calculations
- Look up gene names (see gene\_description in your workspace) and write down a brief description of biological functions of genes. Does their correlation make biological sense?

## Correlated pairs plausible biological connection based on short description

```
1, 6 g1=1994; g2=188;
    2, g1=2872; g2=1269;
    3, g1=1321; g2=10;
    4, g1= 886; g2=819;
    5, g1=2138; g2=1364;
no obvious biological common function
g1=1+floor(rand.*3000); g2=1+floor(rand.*3000);
disp([g1, g2])
```

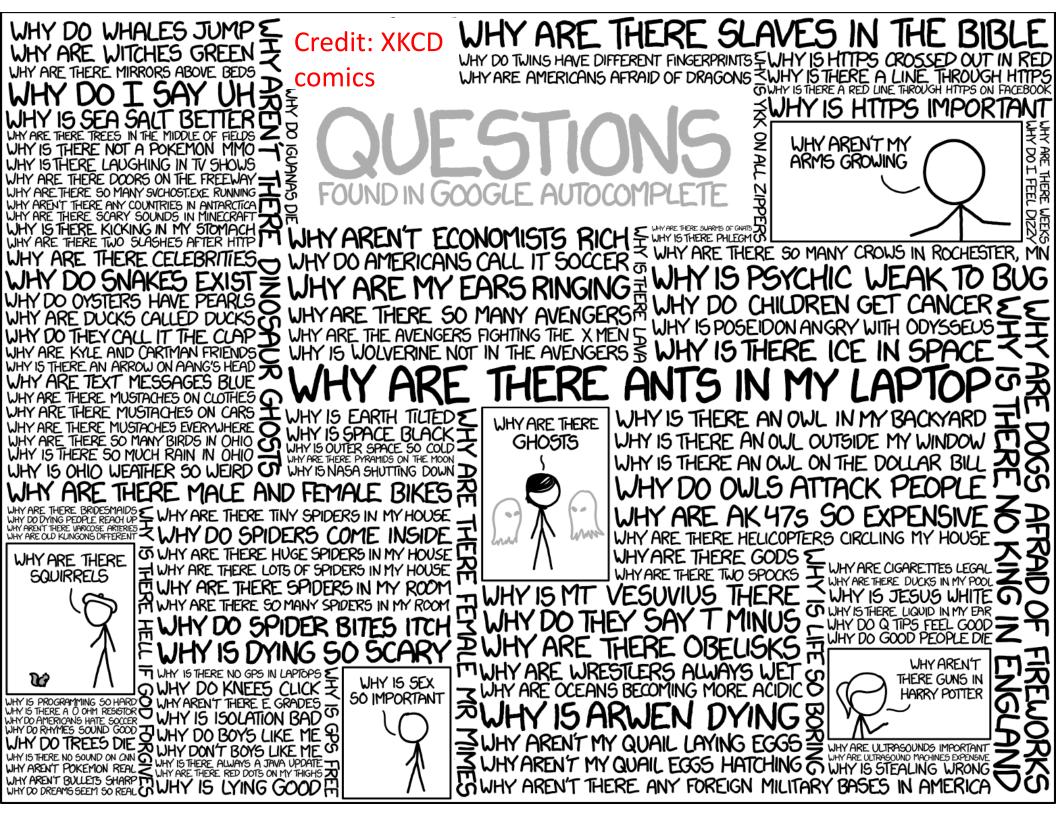
#### Random pairs

```
>> g1=floor(3000.*rand)+1; g2=floor(3000.*rand)+1;
disp([g1,g2]);
>> g1=floor(3000.*rand)+1; g2=floor(3000.*rand)+1;
disp([g1,g2]);
>> g1=floor(3000.*rand)+1; g2=floor(3000.*rand)+1;
disp([g1,g2]);
>> g1=floor(3000.*rand)+1; g2=floor(3000.*rand)+1;
disp([g1,g2]);
```

#### Matlab code

 load expression table.mat • g1=2907; g2=288; x=exp\_t(g1,:)'; y=exp\_t(g2,:)'; figure; plot(x,y,'ko'); Im=fitIm(x,y) • y fit=lm.Fitted; hold on; plot(x,lm.Fitted,'r-'); SST=sum((y-mean(y)).^2); SSR=sum((y fit-mean(y)).^2); SSE=sum((y-y fit).^2); R2=SSR./SST disp([gene\_names(g1), gene\_names(g2)]);

disp(gene description(g1)); disp(gene description (g2));



#### Matlab exercise on #2 on MLR

- Every group works with g0=2907; g1=1527; g2=2629; g3=2881; g4=1144; g5=1066;
- Compute Multiple Linear Regression (MLR): where
   y=exp\_t (g0); x1= exp\_t (g1); x2= exp\_t (g2);
- How much better the MLR did compared to the Single Linear Regression (SLR)?
- Continue increasing the number of genes in x until R\_adj starts to decrease

#### How I did it

```
g0=2907; g1=1527; g2=2629; g3=2881;g4=1144; g5=1066;
y=exp t(g0,:)';
• %% first use one x to predict y
* x=exp t(g1,:)';
figure; plot(x,y,'ko')
lm=fitlm(x,y)
y fit=lm.Fitted;
hold on;
plot(x,lm.Fitted,'r-');
• %% now use 2 x's to predict y
 x=[exp t(g1,:)', exp t(g2,:)'];
lm2=fitlm(x,y)
y fit=lm2.Fitted;
 hold on; plot(x(:,1),y fit,'gd');

    % now use m x's to predict y

 corr matrix=corr(exp t');
• g0=2907;
[u v]=sort(corr matrix(q0,:),'descend');
* x=[exp t(v(2:m+1),:)'];
lm3=fitlm(x,y)
y fit=lm3.Fitted;
• plot(x(:,1),y_fit,'s');
```

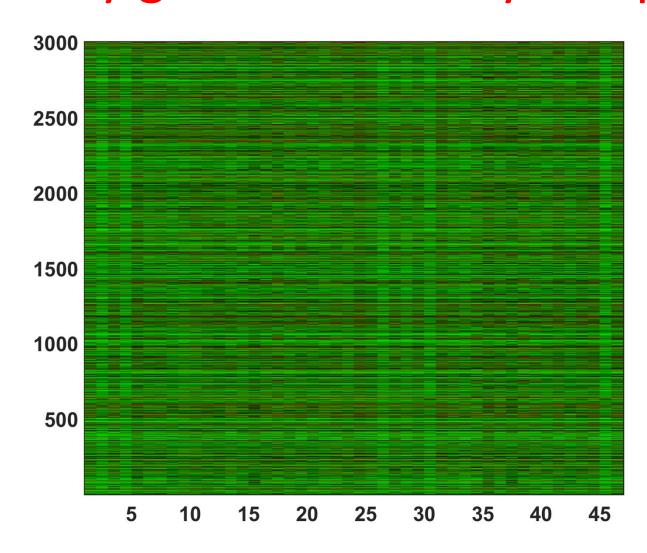
## Clustering analysis of gene expression data

Chapter 11 in
Jonathan Pevsner,
Bioinformatics and Functional Genomics,

3<sup>rd</sup> edition

(Chapter 9 in 2<sup>nd</sup> edition)

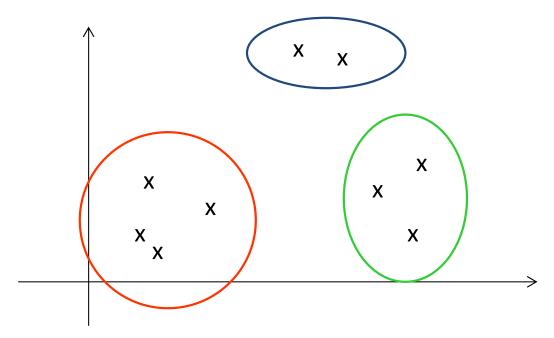
# How to find the entire groups of mutually correlated genes if you have many genes and many samples?



#### Clustering to the rescue!

#### What is clustering?

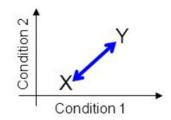
- The goal of clustering is to
  - group data points that are close (or similar) to each other
  - Usually, one needs to identify such groups (or clusters) in an unsupervised manner
  - Sometimes one takes into account prior information (Bayesian methods)
- Need to define some distance d<sub>ij</sub> between objects i and j
- Clustering is easy in 2 dimensions but hard in 3000 dimensions -> need to somehow reduce dimensionality



#### How to define the distance?

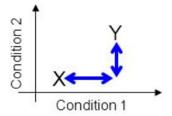
- Euclidean distance:
  - Most commonly used distance
  - Sphere shaped cluster
  - Corresponds to the geometric distance into the multidimensional space

$$d(X,Y) = \sqrt{\sum_{i} (x_i - y_i)^2}$$



- City Block (Manhattan) distance:
  - Sum of differences across dimensions
  - Less sensitive to outliers
  - Diamond shaped clusters

$$d(X,Y) = \sum_{i} |x_i - y_i|$$



The Canberra distance metric is calculated in R by

$$\sum \left(\frac{|x_i - y_i|}{|x_i + y_i|}\right).$$

Correlation coefficient distance

$$d(X,Y) = 1 - \rho(X,Y) = 1 - \frac{Cov(X,Y)}{\sqrt{(Var(X) \cdot Var(Y))}}$$

#### Common types of clustering algorithms

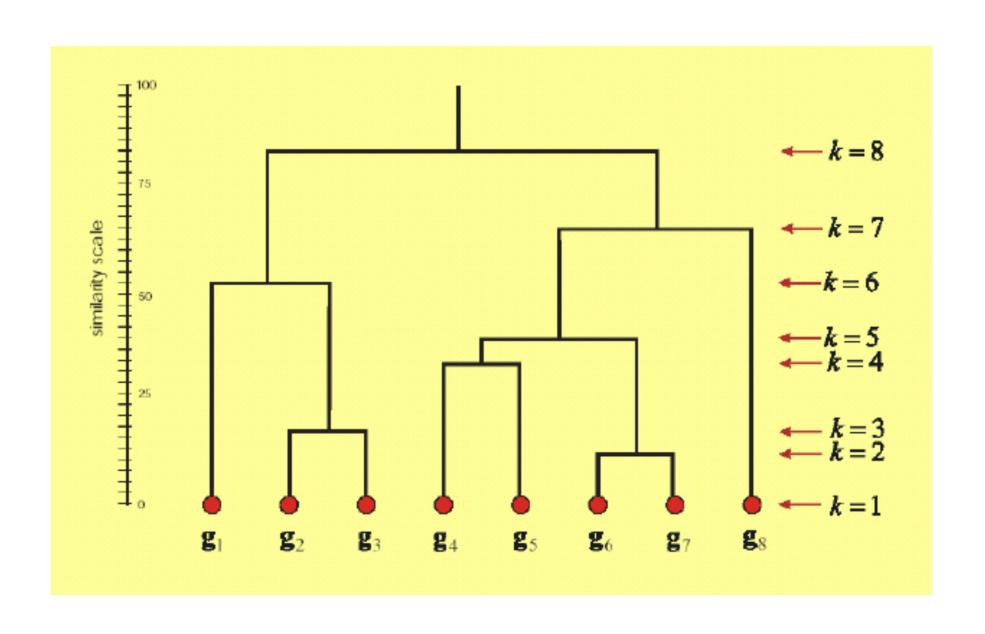
- Hierarchical if one doesn't know in advance the # of clusters
  - Agglomerative: start with N clusters and gradually merge them into 1 cluster
  - Divisive: start with 1 cluster and gradually break it up into N clusters
- Non-hierarchical algorithms
  - K-means clustering:
    - <u>Iteratively</u> apply the following two steps:
    - Calculate the centroid (center of mass) of each cluster
    - Assign each to the cluster to the nearest centroid
  - Principal Component Analysis (PCA)
    - plot pairs of top eigenvectors of the covariance matrix Cov(X<sub>i</sub>, X<sub>i</sub>) and uses visual information to group

#### Hierarchical clustering

#### **UPGMA** algorithm

- Hierarchical agglomerative clustering algorithm
- UPGMA = Unweighted Pair Group Method with Arithmetic mean
- Iterative algorithm:
- Start with a pair with the smallest d(X,Y)
- Cluster these two together and replace it with their arithmetic mean (X+Y)/2
- Recalculate all distances to this new "cluster node"
- Repeat until all nodes are merged

#### Output of UPGMA algorithm



### Clustering in Matlab

# Choices of distance metrics in clustergram(... 'RowPDistValue' ..., 'ColumnPDistValue' ...,)

Metric	Description		
'euclidean'	Euclidean distance (default).		
'seuclidean'	Standardized Euclidean distance. Each coordinate difference between rows in X is scaled by dividing by the corresponding element of the standard deviation S=nanstd(X). To specify another value for S, use D=pdist(X, 'seuclidean', S).		
'cityblock'	City block metric.		
'minkowski'	Minkowski distance. The default exponent is 2. To specify a different exponent, use $D = pdist(X, 'minkowski', P)$ , where P is a scalar positive value of the exponent.		
'chebychev'	Chebychev distance (maximum coordinate difference).		
'mahalanobis'	Mahalanobis distance, using the sample covariance of X as computed by nancov. To compute the distance with a different covariance, use D = pdist (X, 'mahalanobis', C), where the matrix C is symmetric and positive definite.		
'cosine'	One minus the cosine of the included angle between points (treated as vectors).		
'correlation'	One minus the sample correlation between points (treated as sequences of values).		
'spearman'	One minus the sample Spearman's rank correlation between observations (treated as sequences of values).		
'hamming'	Hamming distance, which is the percentage of coordinates that differ.		
'jaccard'	One minus the Jaccard coefficient, which is the percentage of nonzero coordinates that differ.		
custom distance function	A distance function specified using @: D = pdist(X,@distfun) A distance function must be of form		
	d2 = distfun(XI,XJ)		
	taking as arguments a 1-by- $n$ vector XI, corresponding to a single row of X, and an $m2$ -by- $n$ matrix XJ, corresponding to multiple rows of X. distfun must accept a matrix XJ with an arbitrary number of rows. distfun must return an $m2$ -by-1 vector of distances d2, whose $k$ th element is the distance between XI and XJ ( $k$ ,:).		

### Choices of hierarchical clustering algorithm in clustergram( ...'linkage',...)

Х		Matrix with two or more rows. The rows represent observations, the columns represent categories or dimensions.		
method	Algorithm for computing distance between clusters.			
	Method	Description		
	'average'	Unweighted average distance (UPGMA)		
	'centroid'	Centroid distance (UPGMC), appropriate for Euclidean distances only		
	'complete'	Furthest distance		
	'median'	Weighted center of mass distance (WPGMC), appropriate for Euclidean distances only		
	'single'	Shortest distance		
	'ward'	Inner squared distance (minimum variance algorithm), appropriate for Euclidean distances only		
	'weighted'	Weighted average distance (WPGMA)		
	Default: 'single'			