

# Biology 644

**Old Title:** Bioinformatics for Molecular Biologists

**Potential New Title:** Integrated Bioinformatics  
Using R for Both Wet and Dry Scientists

# Bayesian Inference

- Bayesian inference derives the posterior probability  $P(H|E)$  as a consequence of three antecedents, a prior probability  $P(H)$ , a likelihood  $P(E|H)$ , and a normalizing Constant  $P(E)$ . Bayesian inference computes the posterior probability according to Bayes' rule (diachronic interpretation):

$$P(H|E) = \frac{P(E|H) \cdot P(H)}{P(E)}$$

- Posterior probability  $P(H|E)$  = Our update in the probability of the hypothesis  $H$  given some evidence  $E$
- Prior probability  $P(H)$  = Probability of our hypothesis  $H$  before we saw the evidence
- Likelihood  $P(E|H)$  = Probability of seeing the evidence  $E$  if our hypothesis  $H$  is true
- Normalizing Constant  $P(E)$  = Probability of the  $E$  evidence under any circumstances

# The Cookie Problem

- Suppose there are two bowls of cookies.
  - Bowl 1 contains 30 vanilla cookies and 10 chocolate cookies.
  - Bowl 2 contains 20 of each.
- Now suppose you choose one of the bowls at random and, without looking, select a cookie at random. The cookie is vanilla.
- What is the probability that it came from Bowl 1?
- This is a conditional probability; we want  $p(\text{Bowl 1} \mid \text{vanilla})$ , but it is not obvious how to compute it.
- If I ask a different question—the probability of a vanilla cookie given Bowl 1—it would be easy:  $p(\text{vanilla} \mid \text{Bowl 1}) = 3/4$
- We use Bayes Theorem!

$$P(H|E) = \frac{P(E|H) \cdot P(H)}{P(E)} \qquad P(B_1|V) = \frac{\left(\frac{3}{4}\right)\left(\frac{1}{2}\right)}{\frac{5}{8}}$$

- which reduces to 3/5.

# Bayesian Chaining

- **Bayes' Rule** can be written as follows:

$$P(H|E) = \frac{P(H) \cdot P(E|H)}{P(E)} = P(H) \cdot \left( \frac{P(E|H)}{P(E)} \right)$$

Impact of E on P(H)



- Now if we have 2 types of **independent evidence**  $E_1, E_2$  that affects **P(H)** then we can **chain their impacts on H together** by having the **Posterior Probability of  $E_1$**  be the **Prior Probability before  $E_2$** :

$$P(H|E_1, E_2) = P(H) \cdot \left( \frac{P(E_1|H)}{P(E_1)} \right) \cdot \left( \frac{P(E_2|H)}{P(E_2)} \right)$$

- In general, if we have **N independent evidences**  $E_1, \dots, E_N$  then we have:

$$P(H|E_1, \dots, E_N) = P(H) \cdot \left( \frac{\prod_{i=1}^N P(E_i|H)}{\prod_{i=1}^N P(E_i)} \right)$$

# eBayes R Package

- Uses Bayesian methods to shrink the estimated sample variances towards a pooled estimate, resulting in far more stable inference when the number of arrays is small.
  - Compromise between unpooled and pooled t-Tests (i.e. - a method of partial-pooling)
  - Uses the evidence about the information from the total ensemble of genes to assist in the inference about each gene individually
- A number of summary statistics are computed by the eBayes() function for each gene and each contrast:
  - M-value (M) is the log2-fold expression or fold change for a gene.
  - A-value (A) is the average expression level for a gene across all the arrays and channels.
  - Moderated t-statistic (t) is the ratio of the M-value over its posterior residual standard deviation (instead of standard deviation). Has the same interpretation as an ordinary t-statistic except that the standard deviations have been moderated across genes, borrowing information from the ensemble of genes to aid with inference about each individual gene (i.e. intelligent partial-pooling).
  - The moderated t-statistic follows a t-distribution with augmented degrees of freedom.
  - p-value for the moderated t-statistic, usually after some multiple hypothesis correction.
  - Moderated F-statistic (F) also borrows information from the ensemble
  - B-statistic (lods or B) is the posterior log-odds that a gene is differentially expressed.

# Contrasts

- Sometimes there may be comparisons **between the levels of a treatment factor** that you are particularly keen to assess. In this case you can **test the significance** of these individual comparisons using contrasts. Within the **ANOVA** table, the sums of squares and significance of these comparisons will be printed for each factor.
- **Contrast Factor** = the factor for which the contrasts are to be applied in the ANOVA.
- **Contrast Matrix** = a matrix containing the contrasts to be applied in the ANOVA when the selected contrast type is either Regression or Comparison.
  - Each row in the matrix represents **a separate contrast**, and the columns in the matrix correspond to the factor levels.
  - The row labels in the matrix will be used to label the contrasts in the ANOVA table.
- Contrasts have the property that the **sum of the values making up the contrast should be zero**.
- The **simplest contrast** is an individual difference between two levels of a factor and these would be given values -1 and 1.
- Typically a contrast table is used when you have **more than 2 levels**

# Contrast Table

The following table gives common sets of contrasts in a 4 level factor. Any treatment level that is not involved in the contrast is give a value of 0.

Factor level				Contrast Type
A	B	C	D	
-1	1	0	0	Difference between A and B
0	0	-1	1	Difference between C and D
-1	-1	1	1	Difference between average of A and B and that of C and D
-2	1	1	0	Difference between A and the average of B & C
-3	1	1	1	Difference between A and the average of B,C and D
-3	-1	1	3	Linear trend across A, B, C and D

# Gene Ontology (GO)

- “Ontologies” consist of a **representation** of things that are **detectable** or directly observable, and the **relationships** between those things.
- The **Gene Ontology** project provides an ontology of defined terms representing **gene product** properties. The ontology covers three domains:
  - **cellular component** = the parts of a cell or its extracellular environment
  - **molecular function** = the elemental activities of a gene product at the molecular level, such as binding or catalysis
  - **biological process** = operations or sets of molecular events with a defined beginning and end, pertinent to the functioning of integrated living units: cells, tissues, organs, and organisms.
- The **GO ontology** is structured as a **directed acyclic graph (DAG)**, and each term has defined relationships to one or more other terms in the same domain, and sometimes to other domains.
- The GO vocabulary is designed to be **species-neutral**, and includes terms applicable to **prokaryotes** and **eukaryotes**, **single** and **multicellular organisms**.



# Example GO term

- **id:** GO:0000016
- **name:** lactase activity
- **namespace:** molecular\_function
- **def:** "Catalysis of the reaction: lactose + H<sub>2</sub>O = D-glucose + D-galactose."  
[EC:3.2.1.108]
- **synonym:** "lactase-phlorizin hydrolase activity" BROAD [EC:3.2.1.108]
- **synonym:** "lactose galactohydrolase activity" EXACT [EC:3.2.1.108]
- **xref:** EC:3.2.1.108
- **xref:** MetaCyc:LACTASE-RXN
- **xref:** Reactome:20536
- **is\_a:** GO:0004553 ! hydrolase activity, hydrolyzing O-glycosyl compounds