

# CS340 Machine learning

## BN3

# Outline

- Monte Carlo integration
- Genetic linkage analysis

# Monte Carlo integration

- Suppose we want to evaluate the integral

$$E[h(X)] = I = \int h(x)p(x)dx$$

- In low dimensions, we can use numerical integration (eg. quadrature: in matlab, quad, dblquad, triplequad).
- In higher dimensions, a better approach is to sample  $S$  values  $x^s$  from  $p(x)$  and then use the law of large numbers

$$\hat{I} = \frac{1}{S} \sum_{s=1}^S h(x^s)$$

which has standard error

$$se = \sqrt{\frac{\hat{\sigma}^2}{S}}, \quad \hat{\sigma}^2 = \frac{1}{S-1} \sum_{s=1}^S (h(x_s) - \hat{I})^2$$

# Definite integrals

- We can evaluate a definite integral by sampling uniformly within the range

$$\begin{aligned} I &= \int_a^b h(x) = (b-a) \int h(x)p(x)dx \\ p(x) &= U(x|a,b) = \frac{1}{(b-a)}I(a < x < b) \\ I &\approx \frac{1}{S} \sum_{s=1}^S h(x^s) \end{aligned}$$

- Thus the method can also be applied in non-statistical settings.

# Estimating $\pi$

- Area of circle is

$$I = \int_{-r}^r \int_{-r}^r I(x^2 + y^2 \leq r^2) dx dy$$

so  $\pi = I/r^2$ . Let  $h(x, y) = I(x^2 + y^2 \leq r^2)$

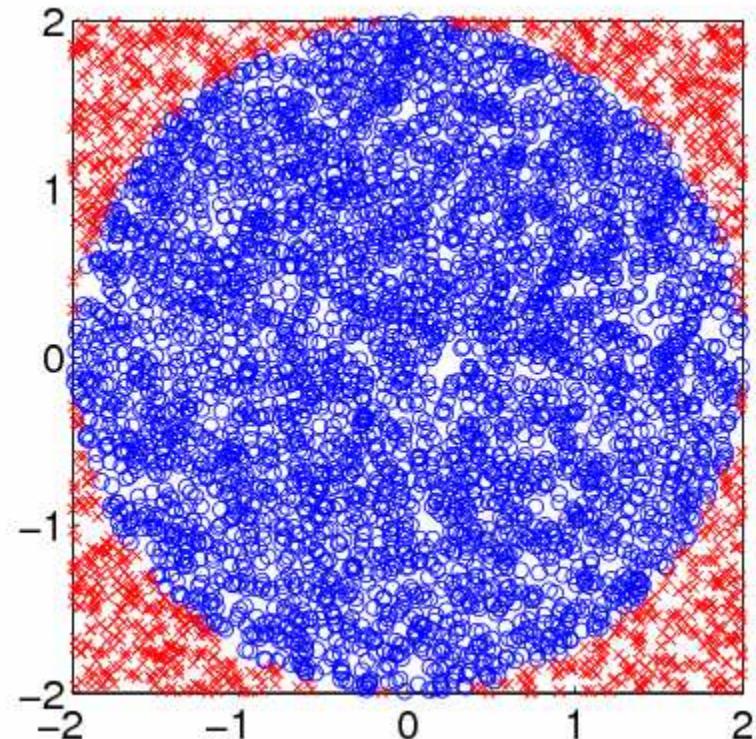
$$\begin{aligned} I &= (b_x - a_x)(b_y - a_y) \int \int h(x, y) p(x) p(y) dx dy \\ &= (2r)(2r) \int \int h(x, y) p(x) p(y) dx dy \\ &= 4r^2 \int \int h(x, y) p(x) p(y) dx dy \\ &\approx 4r^2 \frac{1}{S} \sum_s h(x^s, y^s) \end{aligned}$$

# Estimating $\pi$

- Matlab

```
r=2;  
S=5000;  
xs = unifrnd(-r,r,S,1);  
ys = unifrnd(-r,r,S,1);  
rs = xs.^2 + ys.^2;  
inside = (rs <= r^2);  
samples = 4*(r^2)*inside;  
Ihat = mean(samples)  
piHat = Ihat/(r^2)  
se = sqrt(var(samples)/S)
```

$$\hat{\pi} = 3.1416, \ se = 0.09$$

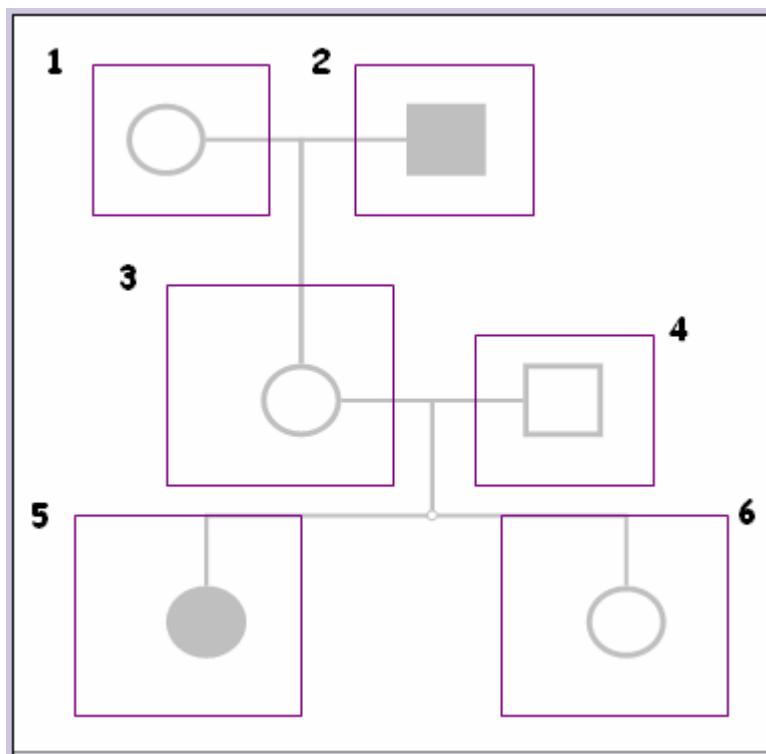


# Outline

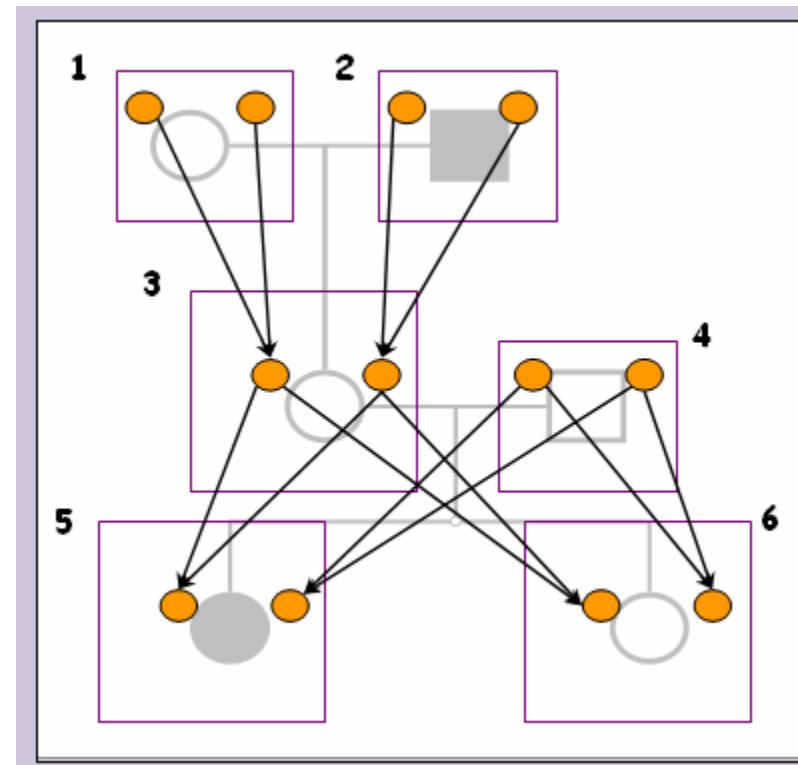
- Monte Carlo integration
- Genetic linkage analysis

# Pedigree analysis

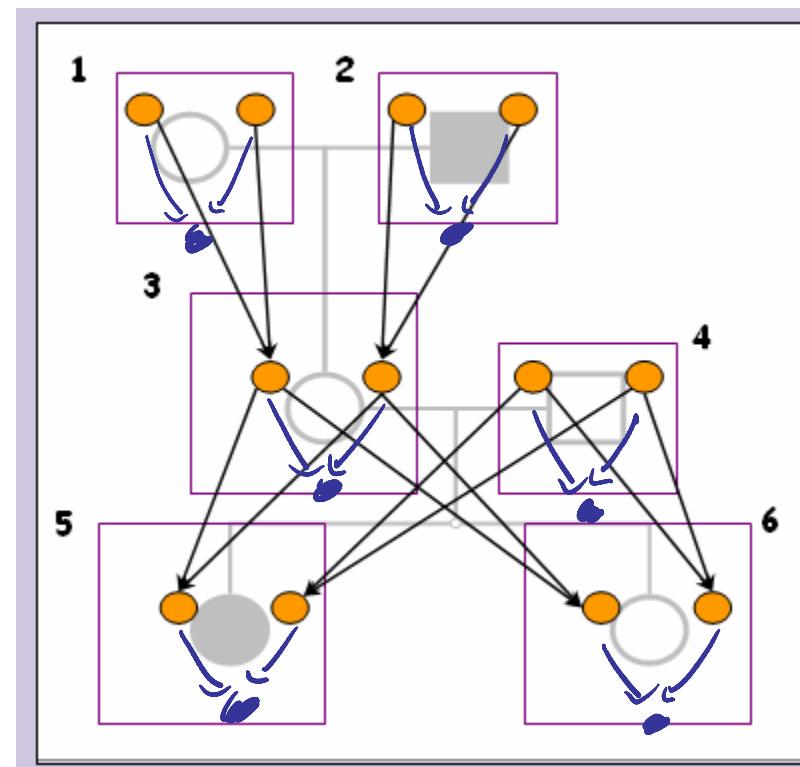
Family tree



Maternal and paternal copies of each gene



# Genotype, phenotype



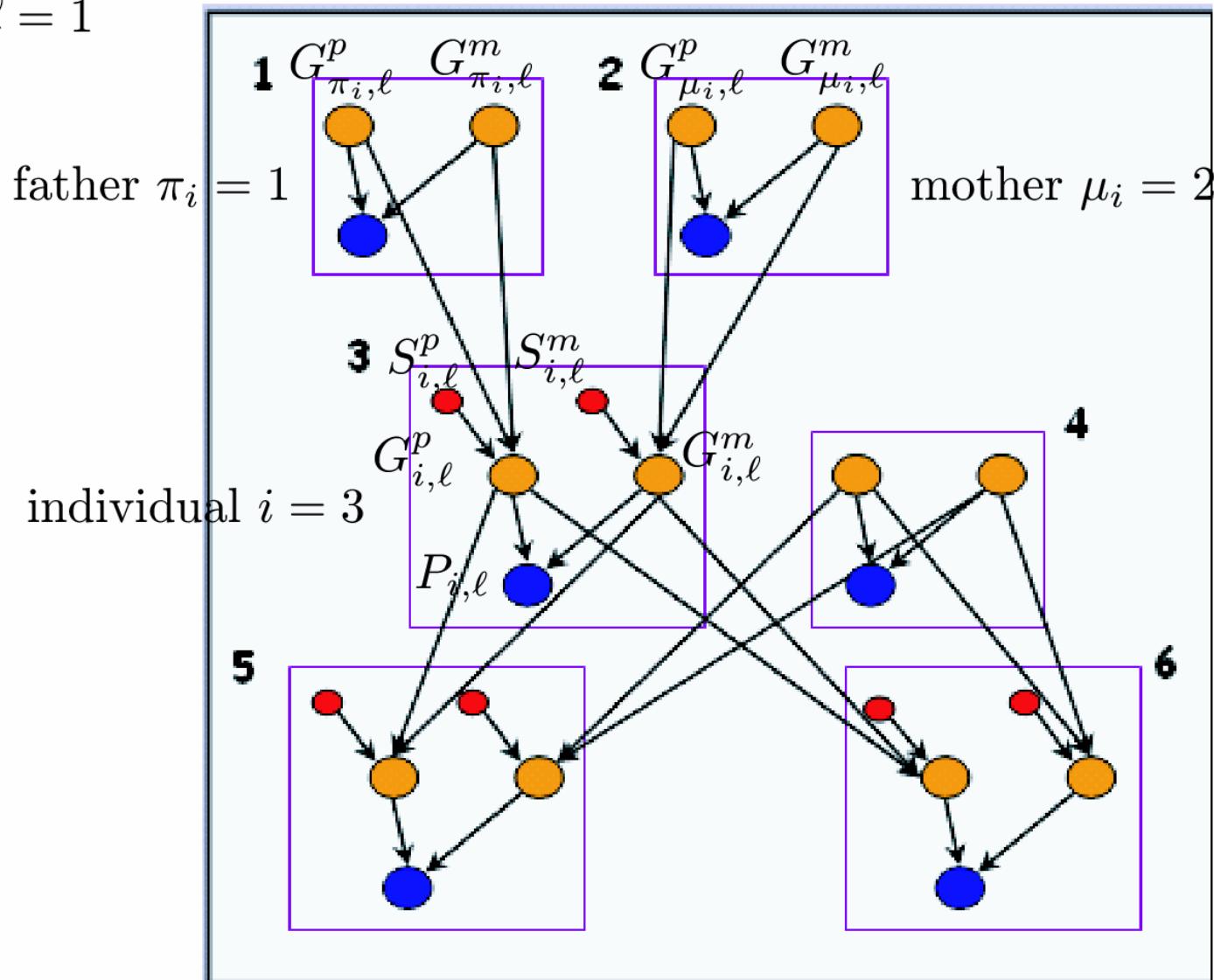
# Penetrance model

- Phenotype = bloodtype, genotype = alleles

$G_p$	$G_m$	$p(P = a)$	$p(P = b)$	$p(P = o)$	$p(P = ab)$
a	a	1	0	0	0
a	b	0	0	0	1
a	o	1	0	0	0
b	a	0	0	0	1
b	b	0	1	0	0
b	o	0	1	0	0
o	a	1	0	0	0
o	b	0	1	0	0
o	o	0	0	1	0

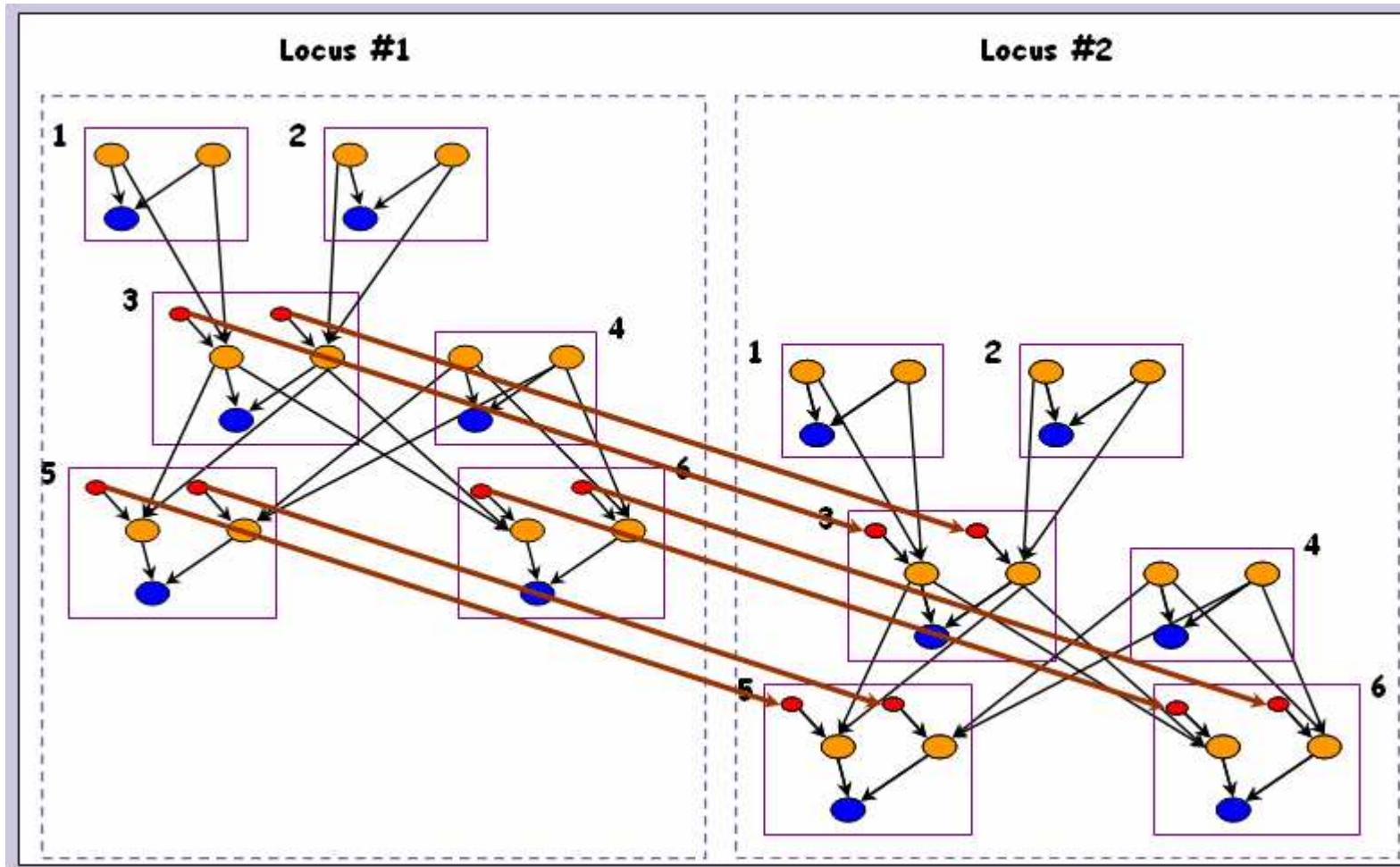
# Haplotypes

locus  $\ell = 1$



$$p(G_{i, \ell}^p | G_{\pi_i, \ell}^p, G_{\pi_i, \ell}^m, S_{i, \ell}^p) = \begin{cases} \delta(G_{i, \ell}^p - G_{\pi_i, \ell}^p) & \text{if } S_{i, \ell}^p = 0 \\ \delta(G_{i, \ell}^p - G_{\pi_i, \ell}^m) & \text{if } S_{i, \ell}^p = 1 \end{cases}$$

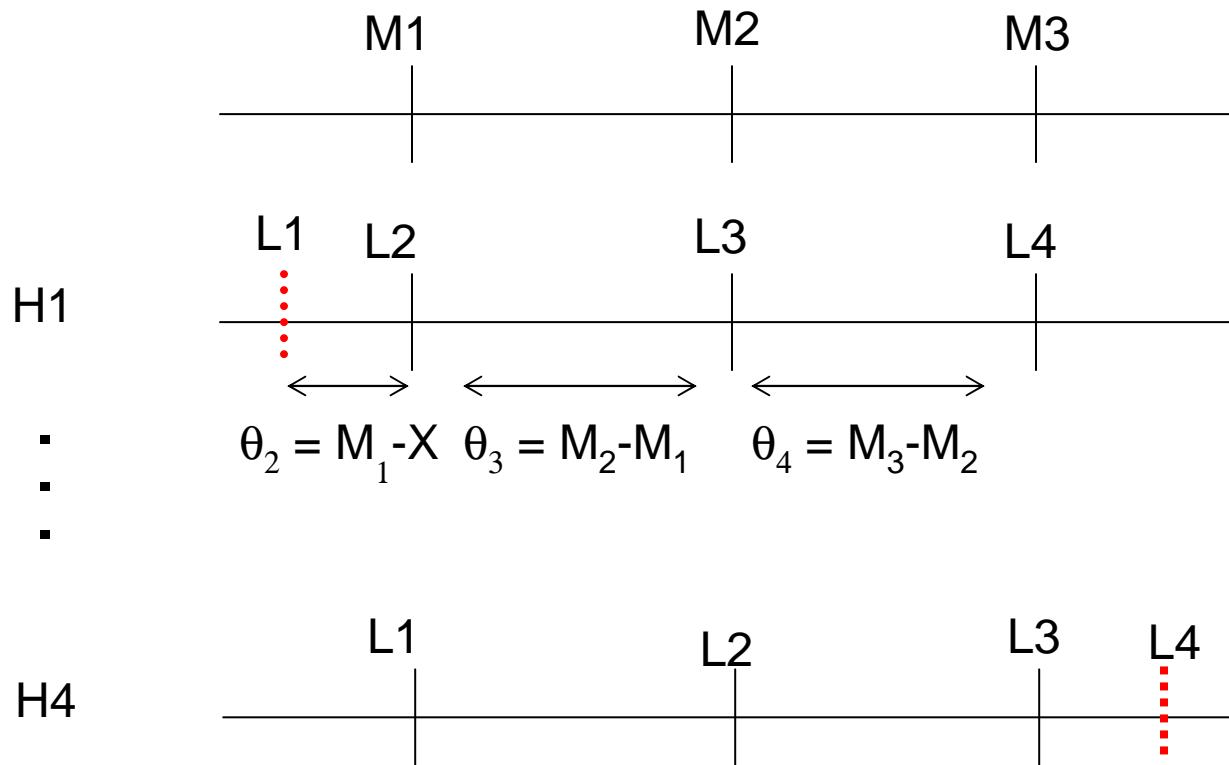
# Cross-over



$S_{i,\ell-1}$	$p(S_{i,\ell} = 0)$	$p(S_{i,\ell} = 1)$
0	$1 - \theta_\ell$	$\theta_\ell$
1	$\theta_\ell$	$1 - \theta_\ell$

# Linkage analysis

- Compute likelihood of data as a function of unknown gene location  $X$



$$p(D|\boldsymbol{\theta}(X)) = \sum_{G_{1:n,1:L}^p} \sum_{G_{1:n,1:L}^m} \sum_{S_{1:n,1:L}^p} \sum_{S_{1:n,1:L}^m} p(G_{1:n,1:L}^p, G_{1:n,1:L}^m, S_{1:n,1:L}^p, S_{1:n,1:L}^m, D|\boldsymbol{\theta}(X))$$