

Structure

Pritchard et al., Genetics 2000

Dakar, February 2011

Clustering in genetics

- We have genetic data from a sample of individuals, each of whom is assumed to have originated from a single unknown population (no admixture).
- We wish to cluster together individuals who are genetically similar.

Structure

- Structure relies on a model-based clustering method for using multilocus genotype data to infer population structure and assign individuals to populations.
- We assume a model in which there are K populations

K=1!

	Locus 1 (f_1)	Locus 2 (f_2)	Locus 3 (f_3)
Indiv 1	0	0	1
Indiv 2	0	1	0
Indiv 3	0	0	0
Indiv 4	0	1	1
Indiv 5	1	1	1
	$n_1=1$	$n_2=3$	$n_3=3$

$n=5, L=3$

Beta-Binomial model

f_j frequency of 1 for marker j, $j=1..L$

n_j number of 1 for marker j

Prior

$$f_j \sim \beta(1,1)$$

Modèle

$$n_j \sim \text{Binomial}(n, f_j) \text{ i.i.d.}$$

Posterior

$$f_j \sim \beta(1+n_j, 1+n-n_j)$$

K=1

	Locus 1 (f_1)	Locus 2 (f_2)	Locus 3 (f_3)
Indiv 1	0	0	1
Indiv 2	0	1	0
Indiv 3	0	0	0
Indiv 4	0	1	1
Indiv 5	1	1	1
	$n_1=1$	$n_2=3$	$n_3=3$
	$f_1 \sim \beta(1+1, 1+5-1)$	$f_2 \sim \beta(1+3, 1+5-3)$	$f_3 \sim \beta(1+3, 1+5-3)$

n=5, L=3

K=2

		Locus 1 (f_{10}, f_{11})	Locus 2 (f_{20}, f_{21})	Locus 3 (f_{30}, f_{31})
$Z_1=0$	Indiv 1	0	0	1
$Z_2=1$	Indiv 2	0	1	0
$Z_3=0$	Indiv 3	0	0	0
$Z_4=1$	Indiv 4	0	1	1
$Z_5=1$	Indiv 5	1	1	1
		$n_1=1$	$n_2=3$	$n_3=3$

Param={(f_{10}, f_{20}, f_{30}), (f_{11}, f_{21}, f_{31}), (Z_1, \dots, Z_5)}

Gibbs sampling step for updating the frequencies (step I)

Given $(Z_1 \dots Z_5)$

Beta-binomial model within each population

		Locus 1 (f_{10})	Locus 2 (f_{20})	Locus 3 (f_{30})
$Z_1=0$	Indiv 1	0	0	1
$Z_3=0$	Indiv 3	0	0	0
		$n_{10}=0$	$n_{20}=0$	$n_{30}=1$
		$\beta(1+0, 1+2)$	$\beta(1+0, 1+2)$	$\beta(1+1, 1+1)$

Beta-Binomial model

Step I of the algorithm

f_{jk} frequency for marker j in pop k

n_{jk} number of 1 for marker j in pop k

Prior

$$f_{jk} \sim \beta(1,1), j=1..L, k=0..(K-1)$$

Modèle

$$n_{jk} \sim \text{Binomial}(n, f_{jk}) \text{ i.i.d.}$$

Given (Z_1, \dots, Z_n)

$$f_{jk} \sim \beta(1+n_{jk}, 1+n-n_{jk})$$

Gibbs sampling step for updating ($Z_1 \dots Z_n$)

Step II

Given the allele frequencies

		Locus 1 (f_{10}, f_{11})	Locus 2 (f_{20}, f_{21})	Locus 3 (f_{30}, f_{31})
$Z_1=1$	Indiv 1	0	0	1

$$P(Z_1=1) = c (1-f_{11})(1-f_{21})f_{31}$$

		Locus 1 (f_{10}, f_{11})	Locus 2 (f_{20}, f_{21})	Locus 3 (f_{30}, f_{31})
$Z_1=0$	Indiv 1	0	0	1

$$P(Z_1=0) = c (1-f_{10})(1-f_{20}) f_{30}$$

Gibbs sampling algorithm for structure

- Start with a given $(Z_1 \dots Z_n)$
For (it in 1..num_{it})
 - Step I
Update the allele frequencies (β distributions)
 - Step II
Update the Z_i 's (use $P(Z_i=1)$ and $P(Z_i=0)$)
- We report the proportion of $(Z_i=1)$ and $(Z_i=0)$ in the chain

R algorithm (Step I)

```
pop1<-sample(0:1,size=n,replace=T)
for (i in 1:nb_it)
{
  ###Count the number of 1 in pop 1 and pop 0
  counts1<-apply(data[pop1==1,],FUN=sum,MARGIN=2)
  counts0<-apply(data[pop1==0,],FUN=sum,MARGIN=2)
  ###Update frequencies (Step I)
  freq1<-rbeta(p,1+counts1,1+sum(pop1==1)-counts1)
  freq0<-rbeta(p,1+counts0,1+sum(pop1==0)-counts0)
  #####.....
}
```

R algorithm (Step II)

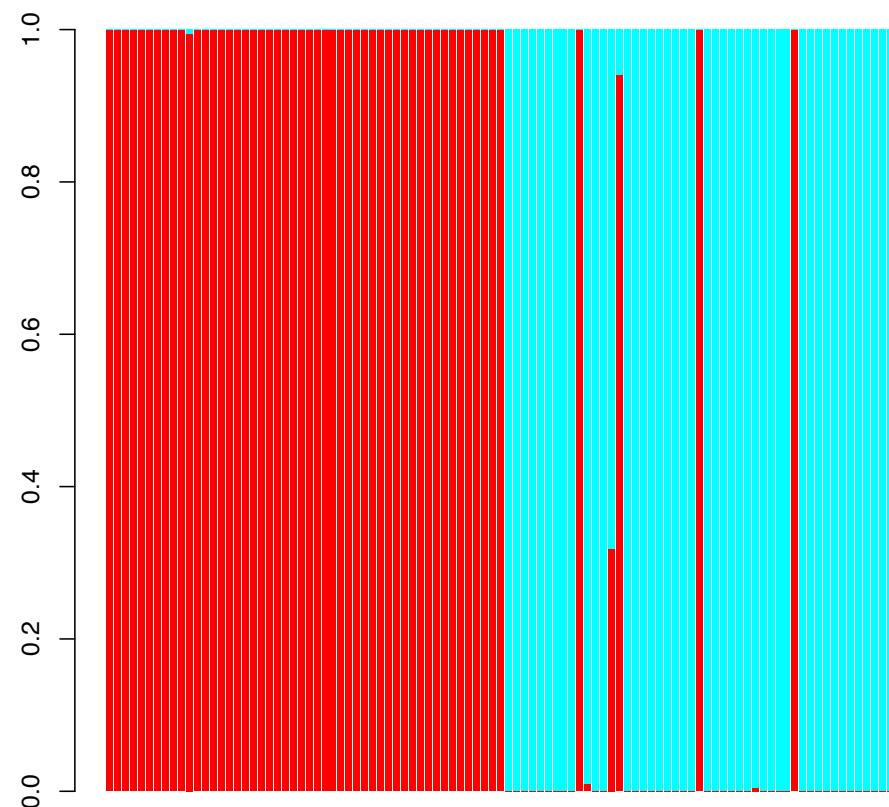
```
for (i in 1:nb_it)
{  ####...
  ###Compute proba to be in pop 1 and pop0
  prob1<-apply(data,FUN=function(x){prod(freq1^(x)*(1-freq1)^(1-x))},MARGIN=1)
  prob0<-apply(data,FUN=function(x){prod(freq0^(x)*(1-freq0)^(1-x))},MARGIN=1)
  pop1<-rbinom(n=n,size=1,prob=prob1/(prob1+prob0))
}
```

R algorithm (no admixture)

```
n<-dim(data)[1];p<-dim(data)[2];nb_it<-1000;burnin<-500;mem<-NULL;lik<-NULL
pop1<-sample(0:1,size=n,replace=T)
for (i in 1:nb_it)
{
  if(i%%100==0)
    cat("Iteration number ",i,"\n")
  ###Count the number of 1 in pop 1 and pop 0
  counts1<-apply(data[pop1==1,],FUN=sum,MARGIN=2)
  counts0<-apply(data[pop1==0,],FUN=sum,MARGIN=2)
  ###Update frequencies (Step I)
  freq1<-rbeta(p,1+counts1,1+sum(pop1==1)-counts1)
  freq0<-rbeta(p,1+counts0,1+sum(pop1==0)-counts0)
  ###Compute proba to be in pop 1 and pop0 (Step II)
  prob1<-apply(data,FUN=function(x){prod(freq1^x*(1-freq1)^(1-x))},MARGIN=1)
  prob0<-apply(data,FUN=function(x){prod(freq0^x*(1-freq0)^(1-x))},MARGIN=1)
  pop1<-rbinom(n=n,size=1,prob=prob1/(prob1+prob0))
  mem<-rbind(mem,pop1);lik<-c(lik,sum(pop1*log(prob1)+(!pop1)*log(prob0)))
}
prob_est<-apply(mem[-(1:burnin),],FUN=function(x){sum(x==1)/length(x)},MARGIN=2)
barplot(rbind(prob_est,1-prob_est),col=rainbow(2),border=NA)
```

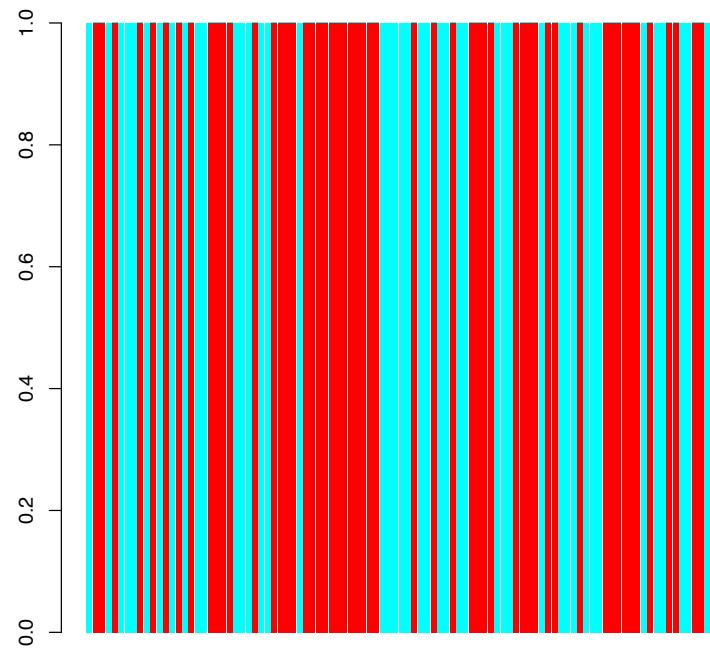
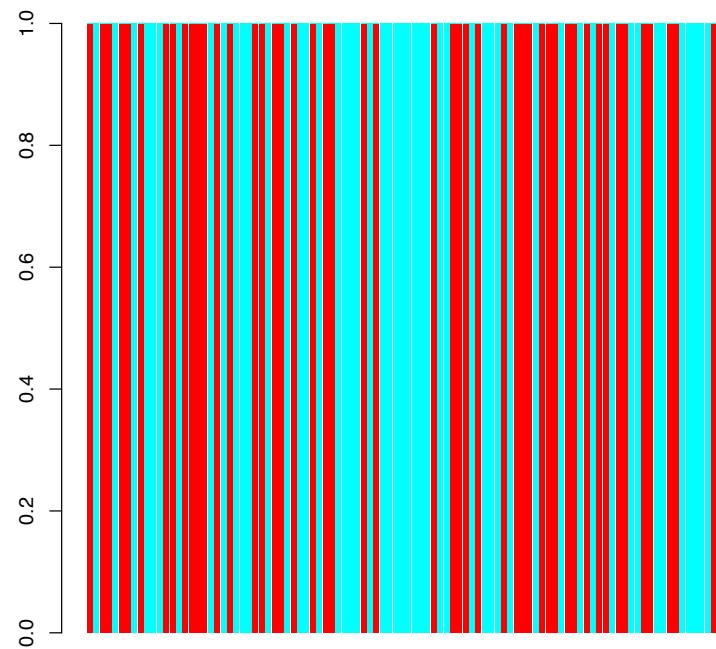
Example 1: true structure

```
### Frequencies of 1: pop 1=60%, pop 0=20%
data<-rbind(matrix(rbinom(50*200,size=1,prob=.
6),nrow=50,ncol=200),matrix(rbinom(50*200,size=1,prob=.
2),nrow=50,ncol=200))
####.....
barplot(rbind(prob_est,1-prob_est),col=rainbow(2),border=NA)
```



Example 2: no structure

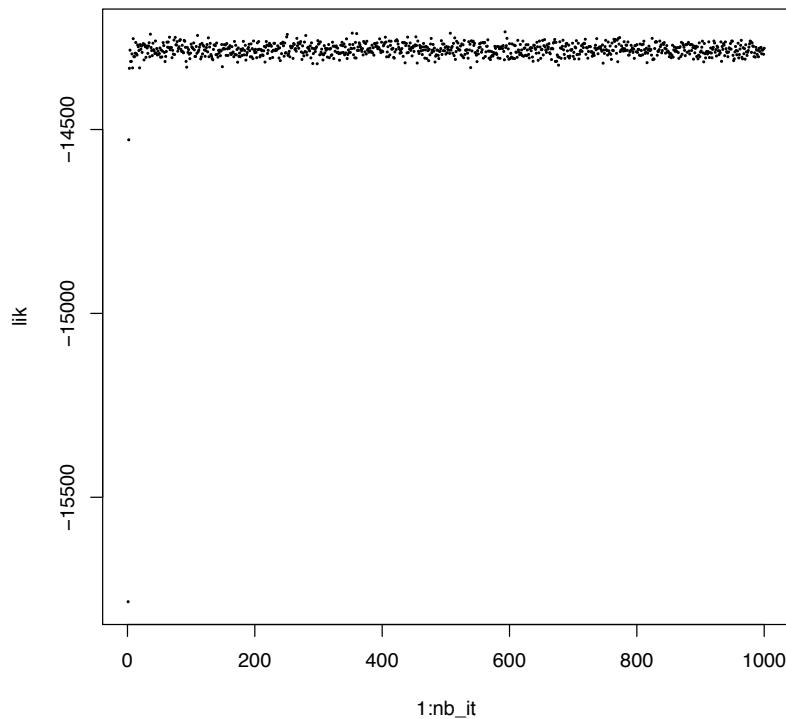
```
data<-matrix(rbinom(100*250,size=1,prob=.5),nrow=100,ncol=250)
```



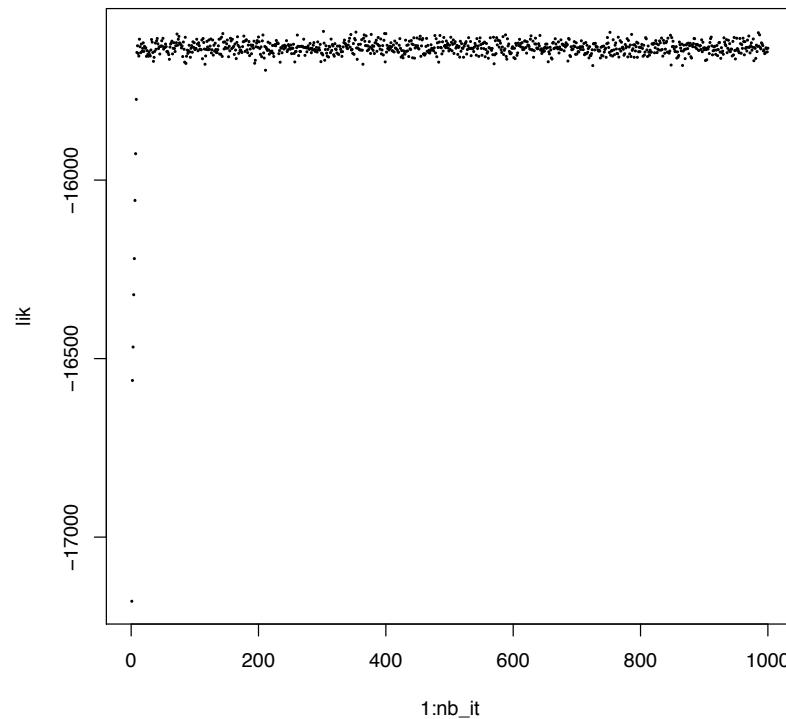
Likelihood

```
plot(1:nb_it,lik,pch=19,cex=.2)
```

Structure



No structure



Model with admixture

		Locus 1 (f_{10}, f_{11})	Locus 2 (f_{20}, f_{21})	Locus 3 (f_{30}, f_{31})
$Z_1 = (0, 0, 1)$	Indiv 1	0	0	1
	Indiv 2	0	1	0
	Indiv 3	0	0	0
	Indiv 4	0	1	1
	Indiv 5	1	1	1

Individual coefficient of admixture

- Parameter q_i proportion of genes coming from pop 1 in individual i
- Prior

$$q_i \sim \beta(\alpha, \alpha)$$

- #1 dans $Z_i \sim \text{Binom}(L, q_i)$

Model with admixture

Step III of the Gibbs sampling

- Given Z_i and the allele frequencies

$$q_i = \beta(\alpha + (\# 1 \text{ in } Z_i), \alpha + (\# 0 \text{ in } Z_i))$$

		Locus 1 (f_{11}, f_{10})	Locus 2 (f_{21}, f_{20})	Locus 3 (f_{31}, f_{30})
$Z_1 = (0, 0, 1)$	Indiv 1	0	0	1

$$q_i = \beta(\alpha+1, \alpha+2)$$

Gibbs sampling algorithm for structure with admixture

- Start with a given $(Z_1 \dots Z_n)$
For (it in $1..num_{it}$)
 - Step I (slightly modified)
Update the allele frequencies (β distributions)
 - Step II (slightly modified)
Update $Z_1 \dots Z_n$
 - Step III
Update the admixture proportion $q_1 \dots q_n$ (β distributions)
- We report the average of $q_1 \dots q_n$ along the chain

R algorithm (admixture)

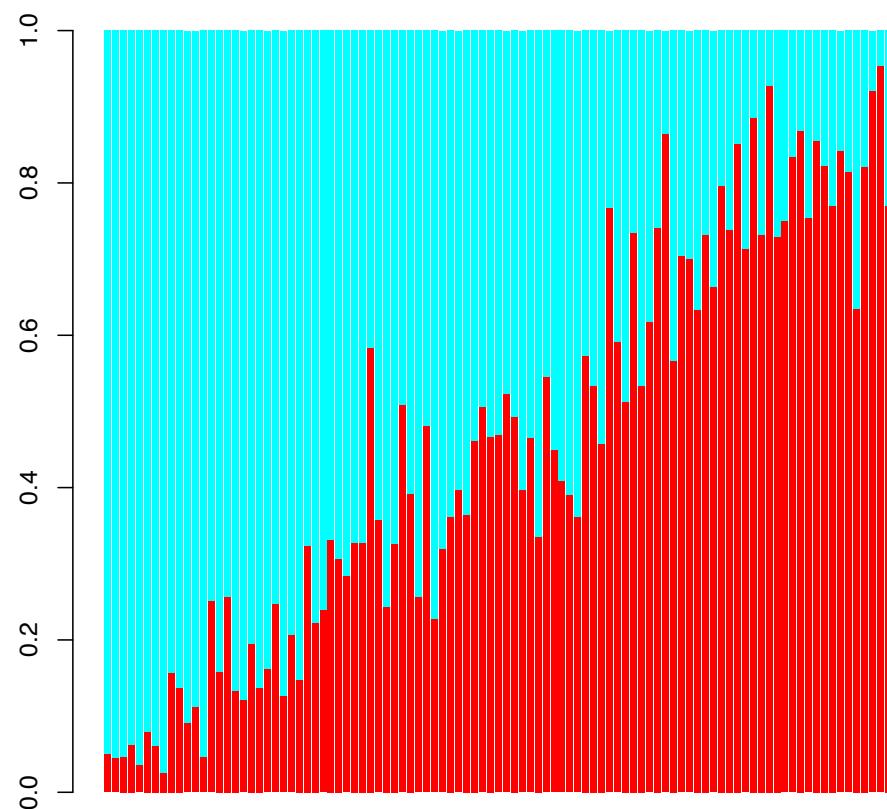
```
n<-dim(data)[1];p<-dim(data)[2];nb_it<-2000;burnin<-1000;mem<-NULL;alpha=1
pop1<-matrix(sample(0:1,size=n*p,replace=T),nrow=n,ncol=p);qadmix<-rep(.5,n);lik<-NULL
for (i in 1:nb_it)
{
  if(i%%100==0)
    cat("Iteration number ",i,"\n")
  ####Count the number of 1 in pop 1 and pop 0
  counts1<-sapply(1:p,FUN=function(x){sum(data[pop1[,x]==1,x])})
  counts0<-sapply(1:p,FUN=function(x){sum(data[pop1[,x]==0,x])})
  n1<-apply(pop1,FUN=sum,MARGIN=2)
  ####Update frequencies (Step I)
  freq1<-rbeta(p,1+counts1,1+n1-counts1)
  freq0<-rbeta(p,1+counts0,1+(n-n1)-counts0)
  ####Compute proba to be in pop 1 and pop0 (Step II)
  qaux<-matrix(qadmix,nrow=n,ncol=p)
  prob1<-as.numeric(qaux*freq1^data*(1-freq1)^(1-data))
  prob0<-as.numeric((1-qaux)*freq0^data*(1-freq0)^(1-data))
  pop1<-matrix(rbinom(n=n*p,size=1,prob=prob1/(prob1+prob0)),nrow=n,ncol=p)
  ####Update q (Step III)
  n1<-apply(pop1,FUN=sum,MARGIN=1)
  qadmix<-rbeta(n,alpha+n1,alpha+p-n1)
  mem<-rbind(mem,qadmix);lik<-c(lik,sum(pop1*log(prob1)+(!pop1)*log(prob0)))}
prob_est<-apply(mem[-(1:burnin),],FUN=mean,MARGIN=2)
barplot(rbind(prob_est,1-prob_est),col=rainbow(2),border=NA)
```

Example 1: cline

```
cline<-function(i,n)
{
v1<-rbinom(n,size=1,prob=.6);v2<-rbinom(n,size=1,prob=.2)
choice<-rbinom(n,size=1,prob=i)
return(ifelse(choice, v1, v2))
}
data<-NULL
for(i in 1:100)
data<-rbind(data,cline(i/100,250))
```

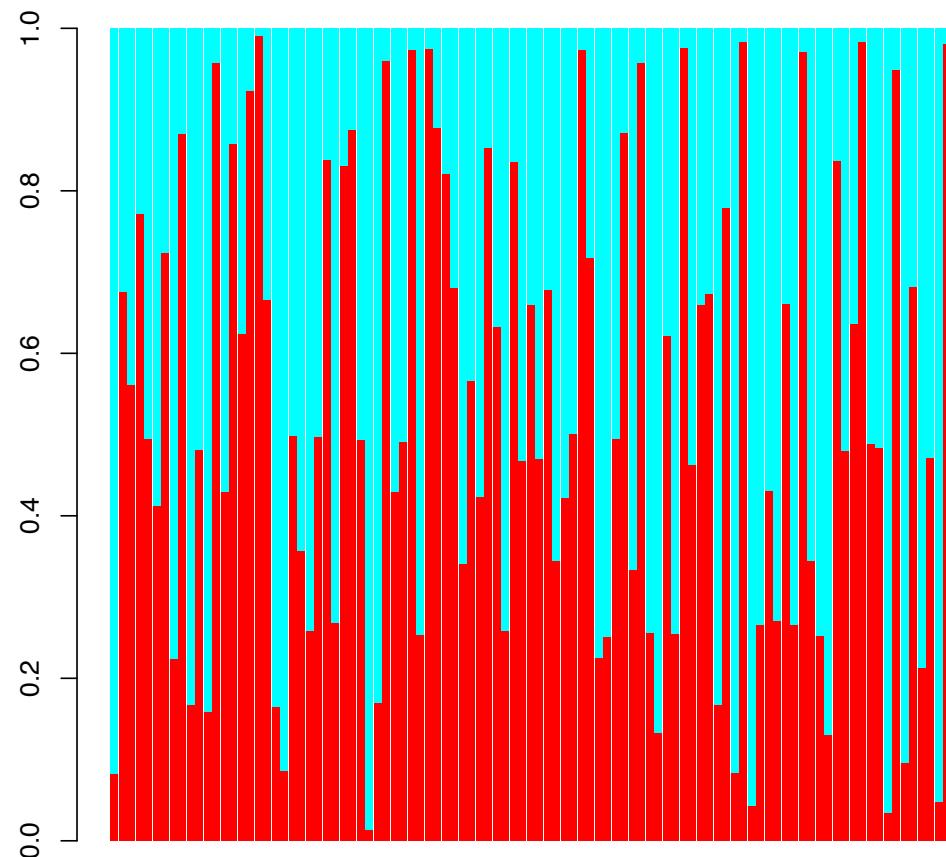
Example 1: cline

```
cline<-function(i,n)
{
v1<-rbinom(n,size=1,prob=.6);v2<-rbinom(n,size=1,prob=.2)
choice<-rbinom(n,size=1,prob=i)
return(ifelse(choice, v1, v2))
}
data<-NULL
for(i in 1:100)
data<-rbind(data,cline(i/100,250))
```



Example 2: no structure

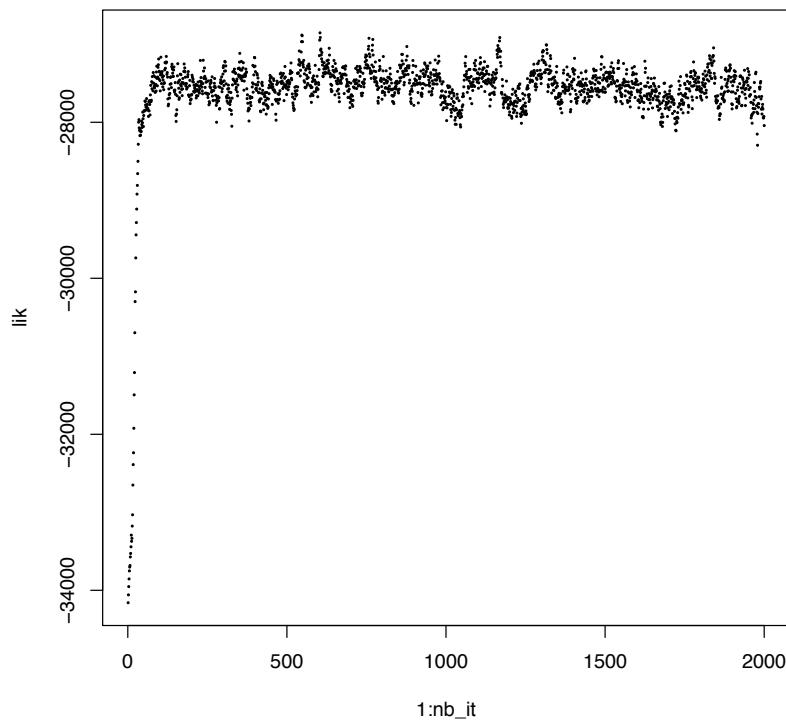
```
data<-matrix(rbinom(1000,size=1,prob=.5),nrow=100,ncol=250)
```



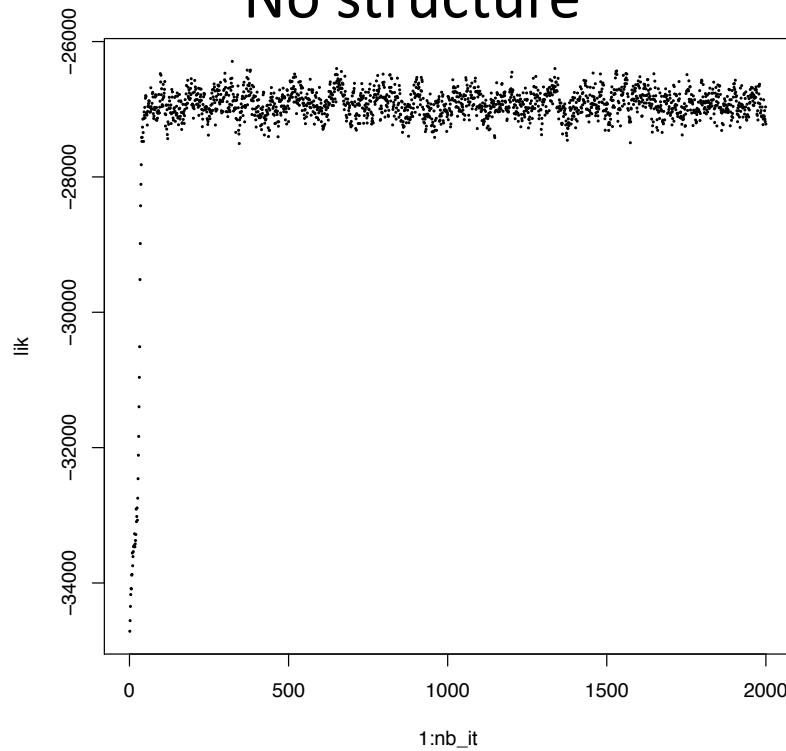
Likelihood

```
plot(1:nb_it,lik,pch=19,cex=.2)
```

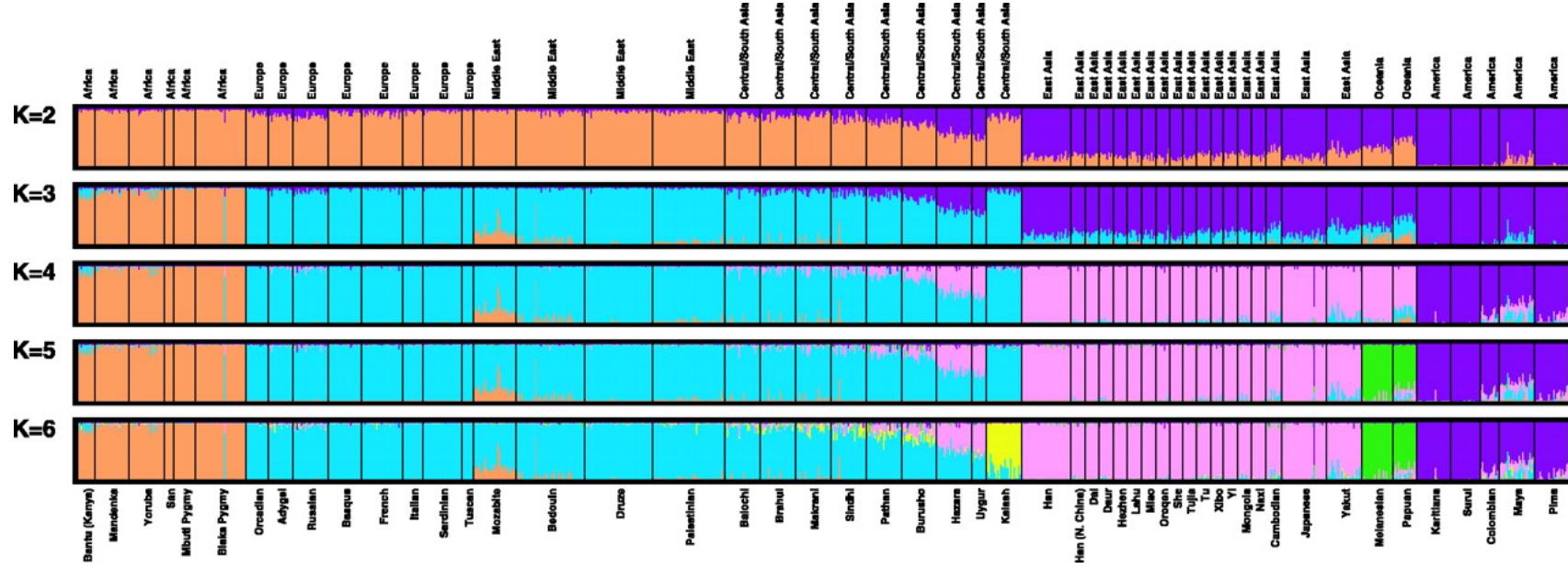
Cline



No structure



Worldwide human population structure



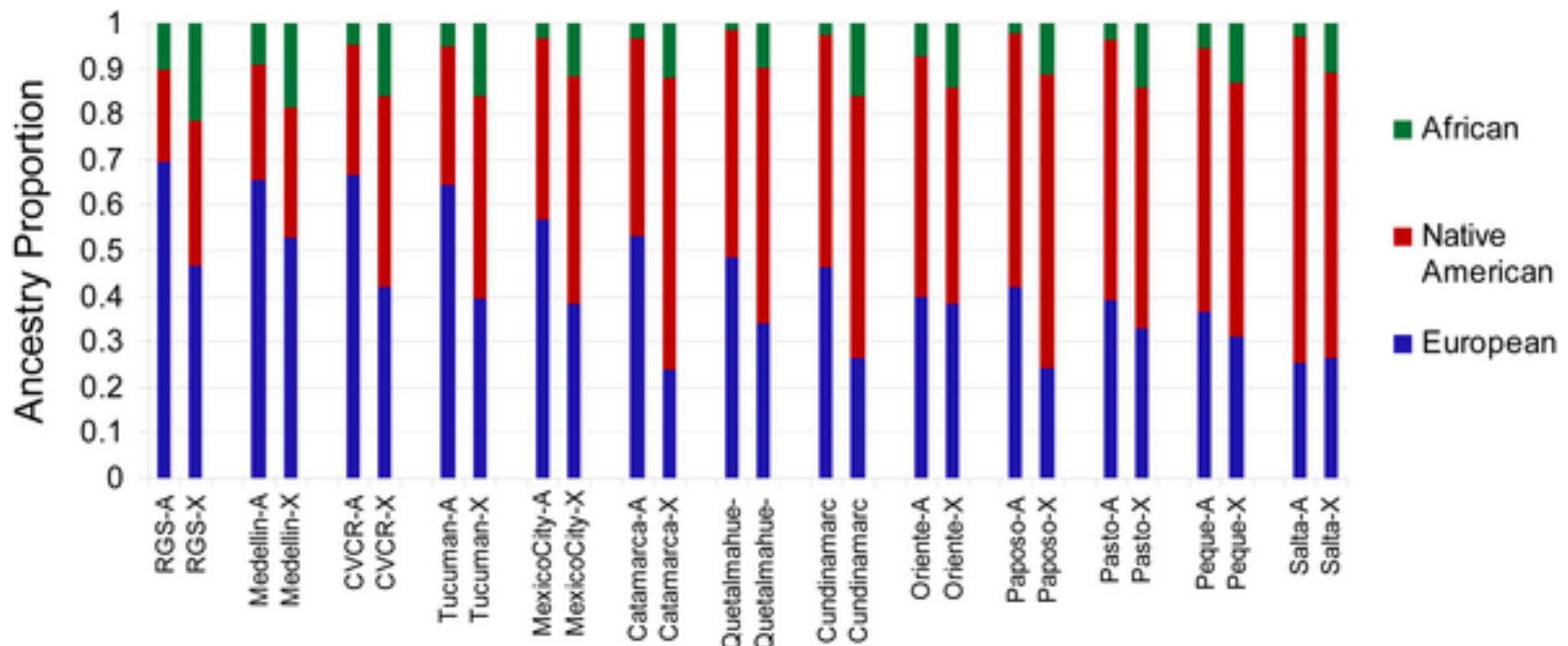
N A Rosenberg et al. Science 2002;298:2381-2385



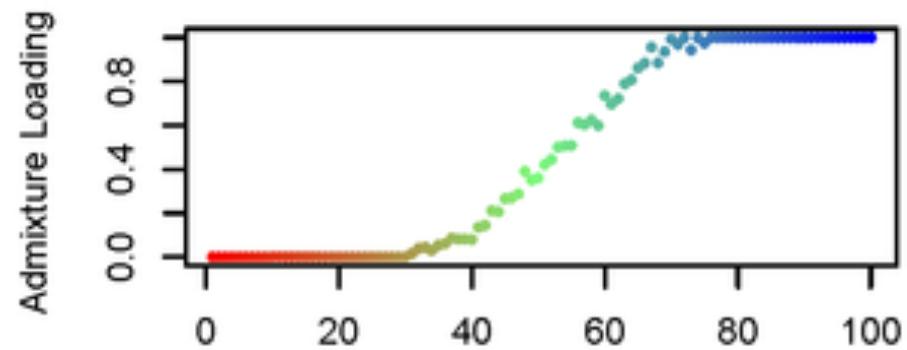
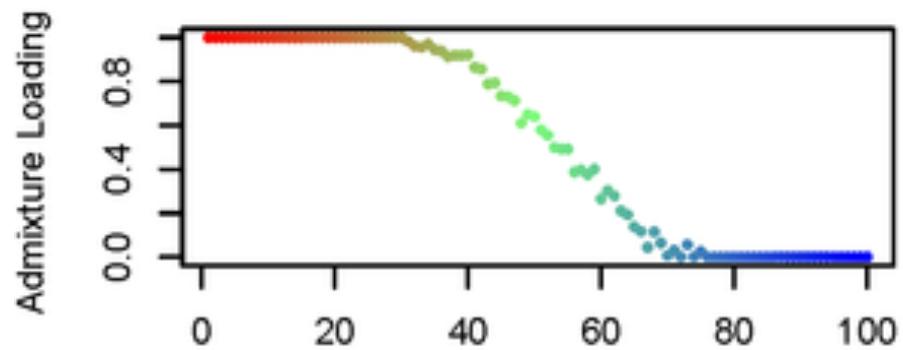
Admixture in Latin American population



Admixture in Latin American population



Word of caution



Word of caution

