

# Compatible model selection

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## 1 Gaussian example

We consider an  $n$ -sample from  $F = \mathcal{N}(0, 1)$ .

### 1.1 Known variance - True model belongs to the parametric family

We consider the parametric family  $Q_\theta = \mathcal{N}(\theta, 1)$  where  $\theta \in \mathbb{R}$ .

In such a case, we have  $\beta = \max_{\theta \in \mathbb{R}} \int \log(q(y|\theta))f(y)dy = -\log(2\pi)/2 - 1/2$ .

For the  $n$ -sample we have  $\beta_n = -n/2 \log(2\pi) - n/2$ .

```
beta <- -1/2*log(2*pi)-1/2
N <- 1000 ; n <- 500 ; B <- 500 ; alpha <- 0.1
statboot <- rep(0,B) ; moy <- ampli <- cover <- 0
for (i in 1:N)
{
  x <- rnorm(n)
  sn2 <- var(x)*(n-1)/n
  for (j in 1:B)
  {
    xstar <- sample(x,n,rep=TRUE)
    sn2star <- var(xstar)*(n-1)/n
    statboot[j] <- sn2/2-sn2star/2
  }
  interval <- -1/2*log(2*pi)-sn2/2-
    quantile(statboot,c(1-alpha/2,alpha/2),names=FALSE)
  moy <- moy+sum(interval)/2
  ampli <- ampli+diff(interval)
  cover <- cover + (beta >= interval[1] & beta <= interval[2])
}
c(beta,moy/N,ampli/N,cover/N)

## [1] -1.418939 -1.417958 0.102508 0.905000
```

## 1.2 Unknown variance - True model belongs to the parametric family

We consider the parametric family  $Q_{(\mu, \sigma^2)} = \mathcal{N}(\mu, \sigma^2)$  where  $(\mu, \sigma^2) \in \mathbb{R} \times \mathbb{R}_+^*$ .

In such a case, we have  $\beta = \max_{(\mu, \sigma^2) \in \mathbb{R} \times \mathbb{R}_+^*} \int \log(q(y | (\mu, \sigma^2))) f(y) dy = -\log(2\pi)/2 - 1/2$ .

For the  $n$ -sample, we have  $\beta_n = -n/2 \log(2\pi) - n/2$ .

```
beta <- -1/2*log(2*pi)-1/2
N <- 1000 ; n <- 500 ; B <- 500 ; alpha <- 0.1
statboot <- rep(0,B) ; moy <- ampli <- cover <- 0
for (i in 1:N)
{
  x <- rnorm(n)
  sn2 <- var(x)*(n-1)/n
  for (j in 1:B)
  {
    xstar <- sample(x,n,rep=TRUE)
    sn2star <- var(xstar)*(n-1)/n
    statboot[j] <- log(sn2)/2-log(sn2star)/2
  }
  interval <- -1/2*(log(2*pi)+1)-log(sn2)/2-
    quantile(statboot,c(1-alpha/2,alpha/2),names=FALSE)
  moy <- moy+sum(interval)/2
  ampli <- ampli+diff(interval)
  cover <- cover + (beta >= interval[1] & beta <= interval[2])
}
c(beta,moy/N,ampli/N,cover/N)
```

```
## [1] -1.4189385 -1.4200766 0.1029049 0.8870000
```

## 2 Regression example

We consider an  $n$ -sample from the Gaussian regression model

$$F = [y|x_1, x_2] = \mathcal{N}(1 + x_1 + x_2, 1).$$

```
n <- 500
X <- cbind(rep(1,n),matrix(rnorm(n*2),n,2))
Xm <- solve(t(X)%*%X)
p <- dim(X)[2]
betan <- -n/2*log(2*pi)-n/2
```

### 2.1 Known variance - True model belongs to the parametric family

We consider the parametric family  $Q_{(\theta_0, \theta_1, \theta_2)} = \mathcal{N}(\theta_0 + \theta_1 x_1 + \theta_2 x_2, 1)$ . In such a case, for the  $n$ -sample, we have  $\beta_n = -n/2 \log(2\pi) - n/2$ .

```

N <- 1000 ; B <- 500 ; alpha <- 0.1
statboot <- rep(0,B) ; moy <- ampli <- cover <- 0
# tp <- txtProgressBar(min = 1, max = N, style = 3, char = "*")
for (i in 1:N)
{
  y <- 1*X[,1]+1*X[,2]+1*X[,3]+rnorm(n,0,1)
  yhat <- X%*%Xm%*%t(X)%*%y
  r <- y-yhat
  radj <- r/sqrt(1-p/n)
  for (j in 1:B)
  {
    radjstar <- sample(radj,n,rep=TRUE)
    ystar <- yhat+radjstar
    yhatstar <- X%*%Xm%*%t(X)%*%ystar
    statboot[j] <- sum(r^2)/2-sum((ystar-yhatstar)^2)/2
  }
  interval <- -n/2*log(2*pi)-sum(r^2)/2-
    quantile(statboot,c(1-alpha/2,alpha/2),names=FALSE)
  moy <- moy+sum(interval)/2
  ampli <- ampli+diff(interval)
  cover <- cover + (betan >= interval[1] & betan <= interval[2])
#   setTxtProgressBar(tp,i)
}
c(betan,moy/N,ampli/N,cover/N)

```

```
## [1] -709.46927 -706.63660 51.09528 0.89300
```

## 2.2 Unknown variance - True model belongs to the parametric family

We consider the parametric family  $Q_{(\theta_0, \theta_1, \theta_2, \theta_3)} = \mathcal{N}(\theta_0 + \theta_1 x_1 + \theta_2 x_2, \theta_3)$  where  $\theta_3 > 0$ . In such a case, for the  $n$ -sample, we have  $\beta_n = -n/2 \log(2\pi) - n/2$ .

```

N <- 1000 ; B <- 500 ; alpha <- 0.1
statboot <- rep(0,B) ; moy <- ampli <- cover <- 0
# tp <- txtProgressBar(min = 1, max = N, style = 3, char = "*")
for (i in 1:N)
{
  y <- 1*X[,1]+1*X[,2]+1*X[,3]+rnorm(n,0,1)
  yhat <- X%*%Xm%*%t(X)%*%y
  r <- y-yhat
  radj <- r/sqrt(1-p/n)
  for (j in 1:B)
  {
    radjstar <- sample(radj,n,rep=TRUE)
    ystar <- yhat+radjstar
    yhatstar <- X%*%Xm%*%t(X)%*%ystar
    statboot[j] <- n*log(mean(r^2))/2-n*log(mean((ystar-yhatstar)^2))/2
  }
}

```

```

    }
    interva <- -n/2*(log(2*pi)+1)-n*log(mean(r^2))/2-
      quantile(statboot,c(1-alpha/2,alpha/2),names=FALSE)
    moy <- moy+sum(interva)/2
    ampli <- ampli+diff(interva)
    cover <- cover + (betan >= interva[1] & betan <= interva[2])
    #   setTxtProgressBar(tp,i)
  }
  c(betan,moy/N,ampli/N,cover/N)

## [1] -709.46927 -708.44143  51.55772  0.89300

```