## CAVI algorithm for normal mean and precision

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#### Simulate data

```
Simulate 50 samples from N(50, 2^2).

set.seed(30027)

X = \text{rnorm}(50, 50, 2)
```

## Implementation CAVI algorithm

```
#X: data
# muO, lambdaO : prior for mu
# a0, b0 : prior for tau
# initial values for mu*, sigma2*, a*, b*: mu.vi.init, sigma2.vi.init, a.vi.init, b.vi.init
# epsilon : If the ELBO has changed by less than epsilon, the CAVI algorithm will stop
# max.iter : maximum number of iteration
cavi.normal <- function(X, mu0, lambda0, a0, b0, mu.vi.init, sigma2.vi.init, a.vi.init,</pre>
                        b.vi.init, epsilon=1e-5, max.iter=100) {
 n = length(X)
  mu.vi = mu.vi.init
  sigma2.vi = sigma2.vi.init
  a.vi = a.vi.init
  b.vi = b.vi.init
  # store the ELBO for each iteration
  elbo = c()
  # I will store mu*, sigma2*, a*, b* for each iteration
  mu.vi.list = sigma2.vi.list = a.vi.list = b.vi.list = c()
  # compute the ELBO using initial values of mu*, sigma2*, a*, b*
  Elogq.mu = -log(sigma2.vi)/2
  Elogq.tau = log(b.vi) - lgamma(a.vi) + (a.vi -1)*digamma(a.vi) - a.vi
  A = sigma2.vi + mu.vi^2 - 2*X*mu.vi + X*X
  B = sigma2.vi + mu.vi^2 - 2*mu0*mu.vi + mu0^2
  Elogp.x.mu.tau = (n+1)/2*(-log(b.vi)+digamma(a.vi))-0.5*a.vi/b.vi*(sum(A) + lambda0*B)
      + (a0-1)*(-log(b.vi)+digamma(a.vi)) - b0*a.vi/b.vi
  elbo = c(elbo, Elogp.x.mu.tau -Elogq.mu - Elogq.tau)
  mu.vi.list = c(mu.vi.list, mu.vi)
  sigma2.vi.list = c(sigma2.vi.list, sigma2.vi)
  a.vi.list = c(a.vi.list, a.vi)
```

```
b.vi.list = c(b.vi.list, b.vi)
# set the change in the ELBO with 1
delta.elbo = 1
# number of iteration
n.iter = 1
# If the elbo has changed by less than epsilon, the CAVI will stop.
while((delta.elbo > epsilon) & (n.iter <= max.iter)){</pre>
 # Update mu.vi and sigma2.vi
 mu.vi = (lambda0*mu0 + sum(X))/(lambda0 + n)
 sigma2.vi = b.vi/a.vi/(lambda0 + n)
 # Update a.vi and b.vi
 a.vi = (n+1)/2 + a0
 A = sigma2.vi + mu.vi^2 - 2*X*mu.vi + X*X
 B = sigma2.vi + mu.vi^2 - 2*mu0*mu.vi + mu0^2
 b.vi = b0 + 0.5*sum(A) + 0.5*lambda0*B
  # compute the ELBO using the current values of mu*, sigma*, a*, b*
 Elogq.mu = -log(sigma2.vi)/2
 Elogq.tau = log(b.vi) - lgamma(a.vi) + (a.vi -1)*digamma(a.vi) - a.vi
 Elogp.x.mu.tau = (n+1)/2*(-log(b.vi)+digamma(a.vi))-0.5*a.vi/b.vi*(sum(A))
              + lambda0*B) + (a0-1)*(-log(b.vi)+digamma(a.vi)) - b0*a.vi/b.vi
 elbo = c(elbo, Elogp.x.mu.tau - Elogq.mu - Elogq.tau)
 mu.vi.list = c(mu.vi.list, mu.vi)
  sigma2.vi.list = c(sigma2.vi.list, sigma2.vi)
 a.vi.list = c(a.vi.list, a.vi)
 b.vi.list = c(b.vi.list, b.vi)
  # compute the change in the elbo
 delta.elbo = elbo[length(elbo)] - elbo[length(elbo)-1]
  # increase the number of iteration
 n.iter = n.iter + 1
}
return(list(elbo = elbo, mu.vi.list = mu.vi.list,
            sigma2.vi.list=sigma2.vi.list, a.vi.list=a.vi.list, b.vi.list=b.vi.list))
```

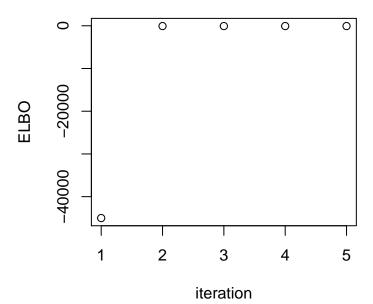
# Apply the CAVI algorithm to the simulated data set

We will consider the following priors:

$$\mu | \tau \sim N(0, \frac{1}{0.01\tau}) \quad \tau \sim Gamma(2, 2)$$

Run the CAVI algorithm with different initial values and check that the ELBO increases at each step by plotting them.

```
mu0=0
lambda0=0.01
a0=2
b0=2
cavi1 = cavi.normal(X, mu0=mu0, lambda0=lambda0, a0=a0, b0=b0, mu.vi.init=2, sigma2.vi.init=4,
                    a.vi.init = 2, b.vi.init=2, epsilon=1e-5, max.iter=100)
cavi.res = cavi1
cavi.res$elbo
## [1] -57628.02235
                       -69.47931
                                    -69.14256
                                                  -69.14251
                                                               -69.14251
plot(cavi.res$elbo, ylab='ELBO', xlab='iteration')
                    2
                             3
                                               5
           1
                                      4
                          iteration
print(paste("mu* and sigma2* = (",
            round(cavi.res$mu.vi.list[length(cavi.res$mu.vi.list)],2), ",",
            round(cavi.res$sigma2.vi.list[length(cavi.res$sigma2.vi.list)],2), ")", sep=""))
## [1] "mu* and sigma2* = (49.92, 0.08)"
print(paste("a* and b* = (",
            round(cavi.res$a.vi.list[length(cavi.res$a.vi.list)],2), ",",
            round(cavi.res$b.vi.list[length(cavi.res$b.vi.list)],2), ")", sep=""))
## [1] "a* and b* = (27.5,116.38)"
cavi2 = cavi.normal(X, mu0=mu0, lambda0=lambda0, a0=a0, b0=b0, mu.vi.init=-10, sigma2.vi.init=4,
                    a.vi.init = 10, b.vi.init=20, epsilon=1e-5, max.iter=100)
cavi.res = cavi2
cavi.res$elbo
## [1] -45008.17186
                       -69.25229
                                    -69.14253
                                                  -69.14251
                                                               -69.14251
plot(cavi.res$elbo, ylab='ELBO', xlab='iteration')
```



```
## [1] "Estimate a* and b* = (27.5,116.38)"
```

The two CAVI runs have (equally) highest ELBO. You can see that approximuated posterior distributions from the runs are the same. I will use the output from the first run:  $q_{\mu}^{*}(\mu)$  is a pdf of N(49.92, 0.08) and  $q_{\tau}^{*}(\tau)$  is a pdf of Gamma(27.5, 116.38).

## Let's compare CAVI results with the exact posterior distribution

Sample from the exact distribution

```
A = X - mean(X)
n = length(X)

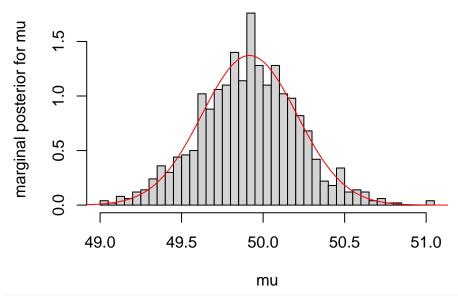
shape = a0 + n/2
rate = b0 + 0.5*sum(A*A) + n*lambda0/(lambda0 + n)*(mean(X)-mu0)*(mean(X)-mu0)/2
tau.true = rgamma(1000, shape = shape, rate = rate)

mean = rep((lambda0*mu0 + sum(X))/(lambda0 + n), 1000)
sd = sqrt(1/(lambda0+n)/tau.true)
mu.true = rnorm(1000, mean, sd)
```

Let's compare the exact and variational marginal posterior distributions.

```
xval <- seq(min(mu.true)-1, max(mu.true)+1, 0.01)
lines(xval, dnorm(xval, mu.vi, sqrt(sigma2.vi)), col="red")</pre>
```

## exact and variational marginal posterior for mu



### exact and variational marginal posterior for tau

