Computational Statistics - Suggested Solution for Exam

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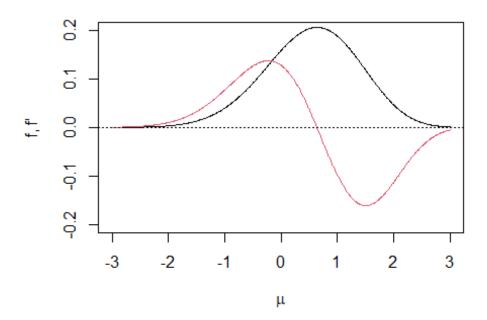
2024-01-09

Assignment 1

lines(mu, vdf(mu), col=2)

abline(h=0, lty=3)

```
Question 1.1
xobs <- 1.65
f <- function(mu){</pre>
  x <- xobs
  exp(-x+mu-exp(-x+mu))*exp(-mu^2/2)
vf <- Vectorize(f)</pre>
mu <- -300:300/100
optimize(f, c(-3, 3), maximum=TRUE)
## $maximum
## [1] 0.6369141
##
## $objective
## [1] 0.2061786
df <- function(mu){</pre>
  x <- xobs
  e1 \leftarrow exp(-x+mu-exp(-x+mu))
  e2 \leftarrow exp(-mu^2/2)
  e1*e2*(-mu) + e1*e2*(1-exp(-x+mu))
vdf <- Vectorize(df)</pre>
plot(c(min(mu), max(mu)), c(-0.2, 0.2), type="n", xlab=expression(mu),
ylab="f, f'")
lines(mu, vf(mu), col=1)
```



Question 1.2

```
secant <- function(mu1, mu2, eps=1e-6, printiter=0){</pre>
  count <- 0
  while (abs(mu1-mu2)>eps){
    mustar <- mu1 - df(mu1)*(mu1-mu2)/(df(mu1)-df(mu2))</pre>
    mu2
            <- mu1
    mu1
            <- mustar
    if (count<printiter) print(mustar) # print approximations for the first</pre>
iterations
    count <- count+1</pre>
  }
  mustar
}
sol1 <- secant(1, 1.2, printiter=10)</pre>
## [1] 0.5013346
## [1] 0.6378854
## [1] 0.6368766
## [1] 0.6369061
## [1] 0.6369061
print(paste("Solution 1:", round(sol1, 6)))
## [1] "Solution 1: 0.636906"
sol2 <- secant(1, 1.8, printiter=10)</pre>
```

```
## [1] -0.8214697
## [1] 0.08055599
## [1] -4.493835
## [1] -4.682196
## [1] -4.788694
## [1] -4.920555
## [1] -5.038775
## [1] -5.158472
## [1] -5.274191
print(paste("Solution 2:", round(sol2, 6)))
## [1] "Solution 2: -37.580087"
```

In the unsuccessful second case, the secant through f(1) and f(1.8) at the low value -0.82; while the next secant still has a negative slope, the following will have a positive slope while f>0. The algorithm will then look for a minimum and diverges slowly to -infinity. The stopping criterion is then fulfilled for mu = -37.6.

```
Question 1.3
```

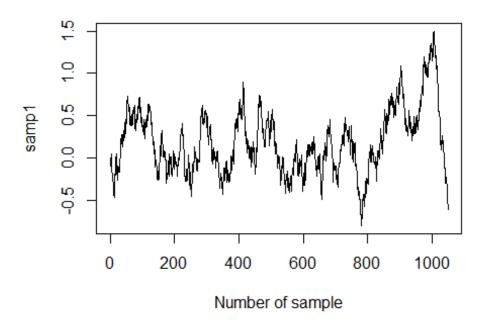
```
mcmc <- function(xstart, n, xobs, sigma=1)
{
    x <- xstart
    sample <- x
    acc <- 0
    for (i in 1:n)
    {
        xcand <- x + sigma*runif(1, -1, 1)
        if (f(xcand)/f(x) > runif(1, 0, 1)){
            x <- xcand
            acc <- acc+1
        }
        sample <- c(sample, x)
    }
    print("Acceptance rate:")
    print(acc/n)
    sample
}</pre>
```

We will later remove a run-in period of xstart and the 50 initially generated observations, therefore we generate 1050 now:

```
samp1 <- mcmc(xstart=0, n=1050, sigma=0.15)

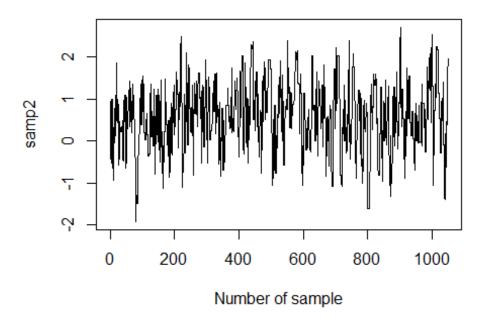
## [1] "Acceptance rate:"
## [1] 0.9819048

plot(0:1050, samp1, type="l", xlab="Number of sample")</pre>
```



```
samp2 <- mcmc(xstart=0, n=1050, sigma=1.5)
## [1] "Acceptance rate:"
## [1] 0.6495238

plot(0:1050, samp2, type="l", xlab="Number of sample")</pre>
```

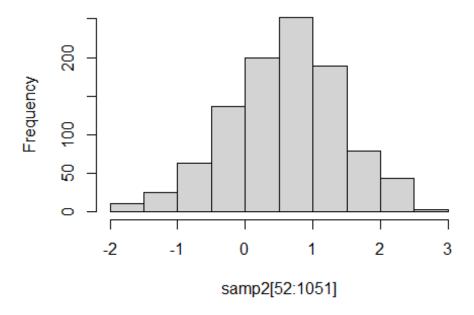


In the first case, there is too much structure suggesting that there is not yet a convergence to the target distribution; the second is much better. Further, the acceptance rate in the first case is too high (> 90%), while in the second case the acceptance rate is closer to the guidance of 44% valid for unidimensional, unimodel distributions. We choose therefore x+Unif[-1.5, 1.5] as proposal and remove the starting value and the first 50 (run-in).

Question 1.4

hist(samp2[52:1051])

Histogram of samp2[52:1051]



```
mean(samp2[52:1051]>=0)
## [1] 0.765
```

The above value is estimated probability $P(\mu \ge 0)$. Note that 1000 simulations is a small number for this purpose and the above estimate is still quite uncertain. Usually, one should draw a larger sample.

```
sortsamp <- sort(samp2[52:1051])
q <- (sortsamp[50]+sortsamp[51])/2
q
## [1] -0.8271181</pre>
```

The lowest 5% of the generated sample are number 1 to 50 in the sorted sample. An estimated 5%-quantile of the posterior is therefore between the 50th and 51st of the ordered sample. Also for this estimate, one should usually draw a larger sample.

Assignment 2

There is no solution available for this assignment at the moment.