

HW4

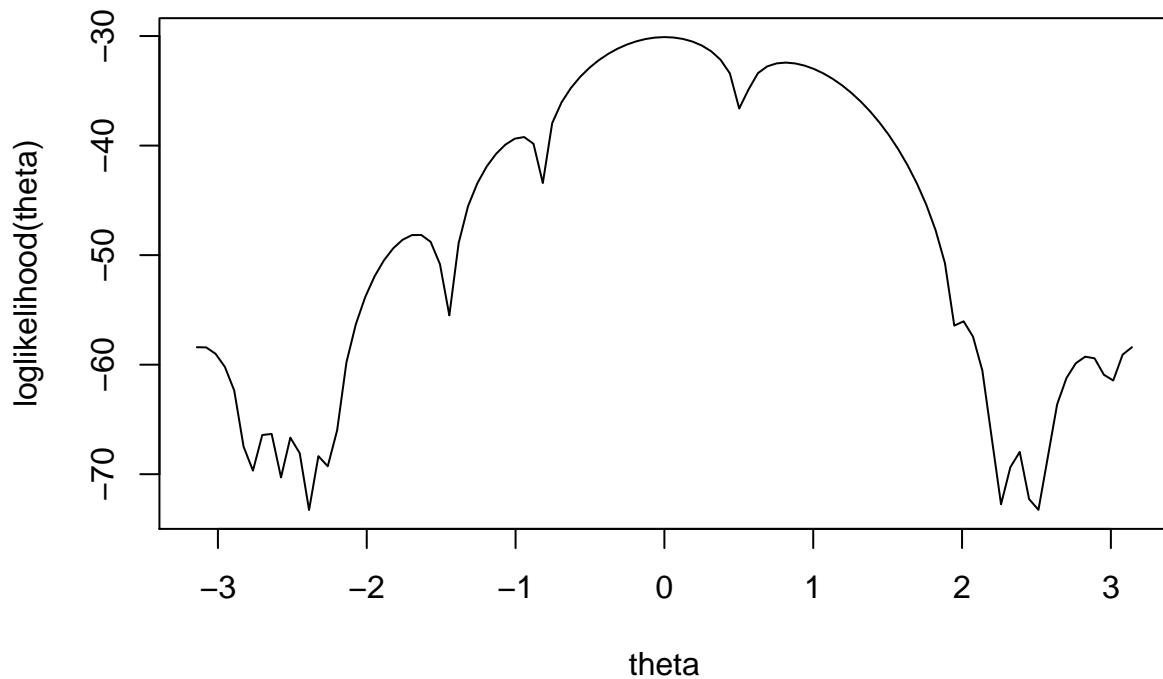
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Many local maxima

loglikelihood

```
loglkh <- function(theta){  
  sum(log((1 - cos(x - theta))/(2 * pi), base = exp(1)))  
}  
loglikelihood <- Vectorize(loglkh)  
curve(loglikelihood, -pi, pi, xname = 'theta')
```



MOM

$$E(X|\theta) = \int_{x=0}^{2\pi} \frac{1 - \cos(x - \theta)}{2\pi} x dx = \int_{x=0}^{2\pi} \frac{x}{2\pi} dx - \frac{1}{2\pi} \int_0^{2\pi} x \cos(x - \theta) dx = \pi - \frac{1}{2\pi} (x \sin(x - \theta) + \cos(x - \theta)) \Big|_0^{2\pi}$$

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$$= \pi - \frac{1}{2\pi}(-2\pi \sin(\theta)) = \pi + \sin(\theta) = \bar{X} = 3.236842$$

```
asin(mean(x)-pi)
```

```
## [1] 0.09539407
```

So $\theta = 0.09539407$.

MLE using Newton-Raphson

```
loglikh.1 <- function(theta){
  sum(sin(theta-x)/(1-cos(theta-x)))
}
newtonRaphson(loglikh.1, asin(mean(x)-pi), dfun = NULL)$root
```

```
## [1] 0.003118157
```

start at -2.7 and 2.7

```
newtonRaphson(loglikh.1, -2.7, dfun = NULL)$root
```

```
## [1] -2.668857
```

```
newtonRaphson(loglikh.1, 2.7, dfun = NULL)$root
```

```
## [1] 2.848415
```

repeat 200 using start values between -pi to pi

```
start <- seq(-pi, pi, length.out = 200)
nr <- double(200)
for (i in start){
  nr[which(start == i)] <- newtonRaphson(loglikh.1, i, dfun = NULL)$root
}
nrtble <- data.table(start = start, MLE = nr)
data.table(cn = names(nrtble), transpose(nrtble))
```

| ## | cn | V1 | V2 | V3 | V4 | V5 | V6 |
|-------|--------------|--------------|--------------|--------------|--------------|------------|-----------|
| ## 1: | start | -3.141593 | -3.110019 | -3.078445 | -3.046871 | -3.015297 | -2.983724 |
| ## 2: | MLE | -3.112471 | -3.112471 | -3.112471 | -3.112471 | -3.112471 | -3.112471 |
| ## | V7 | V8 | V9 | V10 | V11 | V12 | V13 |
| ## 1: | -2.952150 | -2.920576 | -2.889002 | -2.857428 | -2.825855 | -2.794281 | -2.762707 |
| ## 2: | -3.112471 | -3.112471 | -3.112471 | -3.112471 | -3.112471 | -2.786557 | -2.786557 |
| ## | V14 | V15 | V16 | V17 | V18 | V19 | V20 |
| ## 1: | -2.731133 | -2.699560 | -2.667986 | -2.636412 | -2.604838 | -2.573264 | -2.541691 |
| ## 2: | -2.668857 | -2.668857 | -2.668857 | -2.668857 | -2.668857 | -2.509356 | -2.509356 |
| ## | V21 | V22 | V23 | V24 | V25 | V26 | V27 |
| ## 1: | -2.510117 | -2.478543 | -2.446969 | -2.415395 | -2.383822 | -2.352248 | -2.320674 |
| ## 2: | -2.509356 | -2.509356 | -2.509356 | -2.509356 | -2.388267 | -2.297926 | -2.297926 |
| ## | V28 | V29 | V30 | V31 | V32 | V33 | V34 |
| ## 1: | -2.289100 | -2.257526 | -2.225953 | -2.194379 | -2.162805 | -2.131231 | -2.099657 |
| ## 2: | -2.297926 | -2.297926 | -2.232192 | -1.662712 | -1.662712 | -1.662712 | -1.662712 |
| ## | V35 | V36 | V37 | V38 | V39 | V40 | V41 |
| ## 1: | -2.068084 | -2.036510 | -2.004936 | -1.973362 | -1.941788 | -1.910215 | -1.878641 |
| ## 2: | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.662712 |
| ## | V42 | V43 | V44 | V45 | V46 | V47 | V48 |
| ## 1: | -1.847067 | -1.815493 | -1.783919 | -1.752346 | -1.720772 | -1.689198 | -1.657624 |
| ## 2: | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.662712 |
| ## | V49 | V50 | V51 | V52 | V53 | V54 | V55 |
| ## 1: | -1.626050 | -1.594477 | -1.562903 | -1.531329 | -1.499755 | -1.468181 | -1.436608 |
| ## 2: | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.662712 | -1.447503 |
| ## | V56 | V57 | V58 | V59 | V60 | V61 | |
| ## 1: | -1.4050339 | -1.3734601 | -1.3418863 | -1.3103125 | -1.2787387 | -1.2471649 | |
| ## 2: | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | |
| ## | V62 | V63 | V64 | V65 | V66 | V67 | |
| ## 1: | -1.2155911 | -1.1840173 | -1.1524435 | -1.1208697 | -1.0892959 | -1.0577221 | |
| ## 2: | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | |
| ## | V68 | V69 | V70 | V71 | V72 | V73 | |
| ## 1: | -1.0261484 | -0.9945746 | -0.9630008 | -0.9314270 | -0.8998532 | -0.8682794 | |
| ## 2: | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | -0.9544058 | |
| ## | V74 | V75 | V76 | V77 | V78 | | |
| ## 1: | -0.8367056 | -0.805131786 | -0.773557990 | -0.741984195 | -0.710410399 | | |
| ## 2: | -0.9544058 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | | |
| ## | V79 | V80 | V81 | V82 | V83 | | |
| ## 1: | -0.678836604 | -0.647262808 | -0.615689013 | -0.584115217 | -0.552541421 | | |
| ## 2: | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | | |
| ## | V84 | V85 | V86 | V87 | V88 | | |
| ## 1: | -0.520967626 | -0.489393830 | -0.457820035 | -0.426246239 | -0.394672444 | | |
| ## 2: | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | | |
| ## | V89 | V90 | V91 | V92 | V93 | | |
| ## 1: | -0.363098648 | -0.331524853 | -0.299951057 | -0.268377262 | -0.236803466 | | |
| ## 2: | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | | |
| ## | V94 | V95 | V96 | V97 | V98 | | |
| ## 1: | -0.205229671 | -0.173655875 | -0.142082080 | -0.110508284 | -0.078934489 | | |
| ## 2: | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | | |

| | V99 | V100 | V101 | V102 | V103 | | | |
|-------|--------------|--------------|-------------|-------------|-------------|-------------|-----------|----------|
| ## 1: | -0.047360693 | -0.015786898 | 0.015786898 | 0.047360693 | 0.078934489 | | | |
| ## 2: | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | | | |
| | V104 | V105 | V106 | V107 | V108 | V109 | | |
| ## 1: | 0.110508284 | 0.142082080 | 0.173655875 | 0.205229671 | 0.236803466 | 0.268377262 | | |
| ## 2: | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | | |
| | V110 | V111 | V112 | V113 | V114 | V115 | | |
| ## 1: | 0.299951057 | 0.331524853 | 0.363098648 | 0.394672444 | 0.426246239 | 0.457820035 | | |
| ## 2: | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | 0.003118157 | | |
| | V116 | V117 | V118 | V119 | V120 | V121 | V122 | |
| ## 1: | 0.489393830 | 0.5209676 | 0.5525414 | 0.5841152 | 0.6156890 | 0.6472628 | 0.6788366 | |
| ## 2: | 0.003118157 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | |
| | V123 | V124 | V125 | V126 | V127 | V128 | V129 | |
| ## 1: | 0.7104104 | 0.7419842 | 0.7735580 | 0.8051318 | 0.8367056 | 0.8682794 | 0.8998532 | |
| ## 2: | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | |
| | V130 | V131 | V132 | V133 | V134 | V135 | V136 | |
| ## 1: | 0.9314270 | 0.9630008 | 0.9945746 | 1.0261484 | 1.0577221 | 1.0892959 | 1.1208697 | |
| ## 2: | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | |
| | V137 | V138 | V139 | V140 | V141 | V142 | V143 | |
| ## 1: | 1.1524435 | 1.1840173 | 1.2155911 | 1.2471649 | 1.2787387 | 1.3103125 | 1.3418863 | |
| ## 2: | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | |
| | V144 | V145 | V146 | V147 | V148 | V149 | V150 | |
| ## 1: | 1.3734601 | 1.4050339 | 1.4366077 | 1.4681815 | 1.4997553 | 1.5313291 | 1.5629029 | |
| ## 2: | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | |
| | V151 | V152 | V153 | V154 | V155 | V156 | V157 | |
| ## 1: | 1.5944767 | 1.6260505 | 1.6576243 | 1.6891981 | 1.7207719 | 1.7523457 | 1.7839194 | |
| ## 2: | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | |
| | V158 | V159 | V160 | V161 | V162 | V163 | V164 | |
| ## 1: | 1.8154932 | 1.8470670 | 1.8786408 | 1.9102146 | 1.9417884 | 1.973362 | 2.004936 | |
| ## 2: | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 0.8126374 | 2.007223 | 2.007223 | |
| | V165 | V166 | V167 | V168 | V169 | V170 | V171 | V172 |
| ## 1: | 2.036510 | 2.068084 | 2.099657 | 2.131231 | 2.162805 | 2.194379 | 2.225953 | 2.257526 |
| ## 2: | 2.007223 | 2.007223 | 2.007223 | 2.007223 | 2.007223 | 2.007223 | 2.237013 | 2.237013 |
| | V173 | V174 | V175 | V176 | V177 | V178 | V179 | V180 |
| ## 1: | 2.289100 | 2.320674 | 2.352248 | 2.383822 | 2.415395 | 2.446969 | 2.478543 | 2.510117 |
| ## 2: | 2.374712 | 2.374712 | 2.374712 | 2.374712 | 2.374712 | 2.374712 | 2.488450 | 2.488450 |
| | V181 | V182 | V183 | V184 | V185 | V186 | V187 | V188 |
| ## 1: | 2.541691 | 2.573264 | 2.604838 | 2.636412 | 2.667986 | 2.699560 | 2.731133 | 2.762707 |
| ## 2: | 2.848415 | 2.848415 | 2.848415 | 2.848415 | 2.848415 | 2.848415 | 2.848415 | 2.848415 |
| | V189 | V190 | V191 | V192 | V193 | V194 | V195 | V196 |
| ## 1: | 2.794281 | 2.825855 | 2.857428 | 2.889002 | 2.920576 | 2.952150 | 2.983724 | 3.015297 |
| ## 2: | 2.848415 | 2.848415 | 2.848415 | 2.848415 | 2.848415 | 2.848415 | 2.848415 | 3.170715 |
| | V197 | V198 | V199 | V200 | | | | |
| ## 1: | 3.046871 | 3.078445 | 3.110019 | 3.141593 | | | | |
| ## 2: | 3.170715 | 3.170715 | 3.170715 | 3.170715 | | | | |

Modeling beetle data

```
beetles <- data.frame(  
  days    = c(0, 8, 28, 41, 63, 69, 97, 117, 135, 154),  
  beetles = c(2, 47, 192, 256, 768, 896, 1120, 896, 1184, 1024))
```

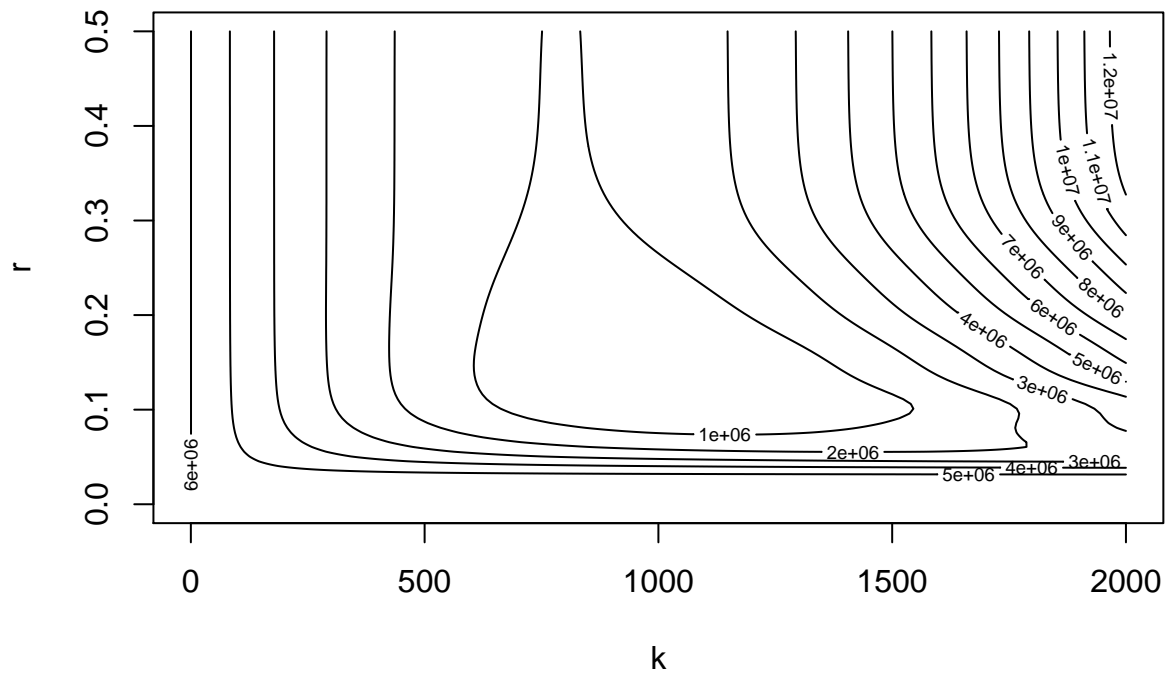
Gauss-Newton and Contour Plot

```
library(pracma) ## gauss newton method  
library(plotly) ## plot contour  
t <- beetles$days  
b <- beetles$beetles  
N0 <- b[1]  
  
Nt <- function(x){  
  (x[1] * (N0) / (N0 + (x[1] - N0) * exp(-x[2] * t)) - b)  
}  
gaussNewton(c(1000, 1), Nt)
```

```
## $xs  
## [1] 1049.4072441    0.1182684  
##  
## $fs  
## [1] 73419.7  
##  
## $niter  
## [1] 8  
##  
## $relerr  
## [1] 7.275958e-11
```

```
ct <- function(k, r){  
  sum((k * (N0) / (N0 + (k - N0) * exp(-r * t)) - b)^2)  
}  
k <- seq(0, 2000, length.out = 1e4)  
r <- seq(0, 0.5, length.out = 1e2)  
z <- outer(k,  
           r,  
           Vectorize(ct))  
contour(k, r, z, xlab = "k", ylab = "r", main = "contour plot")
```

contour plot



```
## MLE
```

```
logb <- log(b, base = exp(1))
logb
```

```
## [1] 0.6931472 3.8501476 5.2574954 5.5451774 6.6437897 6.7979404 7.0210840
## [8] 6.7979404 7.0766538 6.9314718
```

```
t
```

```
## [1] 0 8 28 41 63 69 97 117 135 154
```

```
llk <- function(x){
  k <- x[1]
  r <- x[2]
  sigmasq <- x[3]
  -sum(-(log(2 * pi * (sigmasq)) / 2 )- (logb - log((k * NO)/(NO + (k - NO) * exp(-r * t)))) ^ 2 / (2 * sigmasq)))
}
llk(c(1e3, .2, 1))
```

```
## [1] 11.4255
```

```
rs <- optim(c(1000, 0.2, 1), llk, method = "BFGS", hessian = TRUE)
rs
```

```
## $par
## [1] 954.3360540 0.1781541 0.4253134
##
## $value
## [1] 9.915059
##
## $counts
## function gradient
##      53      25
##
## $convergence
## [1] 0
##
## $message
## NULL
##
## $hessian
##           [,1]      [,2]      [,3]
## [1,] 1.472245e-05 5.355685e-02 -0.0040449715
## [2,] 5.355685e-02 6.409873e+02 -0.0008940906
## [3,] -4.044971e-03 -8.940906e-04 27.6453070223
```

```
solve(rs$hessian)
```

```
##           [,1]      [,2]      [,3]
## [1,] 103565.245366 -8.653235974 15.15305023
## [2,] -8.653236 0.002283101 -0.00126604
## [3,] 15.153050 -0.001266040 0.03838961
```

Reference

<https://stackoverflow.com/questions/19079152/contour-plot-of-a-custom-function-in-r> [jun-
[yan/stat-5361](https://github.com/jun-yan/stat-5361)]<https://github.com/jun-yan/stat-5361>