

## Assignment-6

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### Function for Maximum likely hood estimation.

Method to estimate parameters for different distributions is below. The input parameters are:- “distribution\_type” : Type of distribution. To be given as a string input “data\_input” : Data generated using the specific distribution to be given as input

Mean and variance and given data is calculated and using maximum likelihood estimation, calculations will be done to estimate specific parameter values. These will be printed after calculations.

```
library(purrr)
library(MASS)
Parameter_finder <- function( distribution_type , data_input ) {
  # This function is used to calculate parameters for
  #different distributions of data using mean and variance
  mean_data = mean(data_input)
  var_data = var(data_input)

  if ( distribution_type == "Poisson" ) {
    print(paste0("Poisson distribution parameters - Lambda value is ", mean_data))
  }

  else if ( distribution_type == "Bernoulli" ) {
    print(paste0("Bernoulli distribution parameters - p value is", mean_data))
  }
  else if ( distribution_type == "Exponential" ) {
    print(paste0("Exponential distribution parameters - rate value is ", 1/mean_data))
  }
  else if (distribution_type == "Geometric" ) {
    p = 1/mean_data
    print(paste0("Geometric distribution parameters - p value is ", p))
  }
  else if ( distribution_type == "Normal" ) {
    print(paste0("Normal distribution parameters - Mean value is ", mean_data))
    print(paste0("Standard deviation is ", sqrt(var_data)))
  }
  else if ( distribution_type == "Binomial" ) {
    print(paste0("Binomial distribution parameters - p value is ",
      mean_data/length(data_input)))
  }
  else if ( distribution_type == "Uniform" ) {
    print(paste0("Uniform distribution parameters - a value is ", min(data_input),
      " b value is ", max(data_input)))
  }
}
```

```

}

else if ( distribution_type == "Gamma") {
  x<- data_input
  n = length(data_input)
  a <- 1
  # Defining derivative functions for newton-raphson
  f1 <- function(a) -n*digamma(a)- n*log(mean(x)) + n*log(a) + sum(log(x))
  f2 <- function(a) -n*trigamma(a) + n/a

  #Newton-Raphson for 60 iterations
  for( i in 1:60) a <- a - f1(a)/f2(a)
  # Beta value
  b <- mean(x)/a

  print(paste0("Estimate of alpha is ",a))
  # As input to R is inverse of beta
  print(paste0("Estimate of beta is ",1/b))
}

else if ( distribution_type == "Beta") {
  x <- data_input
  # Initial guesses
  a2 <- 1
  b2<- 1
  # Running Newton-Raphson 20 times
  for (i in 1:20){
    # Matrices to hold estimates
    parm <- matrix(c(a2,b2), nrow=2)
    # Two components by partial derivatives with respect to alpha and beta
    f<- matrix(c(digamma(a2+b2) - digamma(a2) + mean(log(x)), digamma(a2+b2)- digamma(b2)+ mean(log(1-x))),
    # Solving these using jacobian matrix
    J <- solve(matrix(c(trigamma(a2+b2)-trigamma(a2), trigamma(a2+b2), trigamma(a2+b2), trigamma(a2+b2)-
    # Matrix multiplication
    f2 <- parm - J%*%f
    # Getting estimates and putting them back inside
    a2 <- f2[1,]
    b2 <- f2[2,]
  }

  print(paste0("Estimate of alpha is ",a2))
  # As input to R is inverse of beta
  print(paste0("Estimate of beta is ",b2))
}

else if (distribution_type=="multinomial")
{
  n_row = nrow(data_input)
  prob = c(0,0,0,0)
  p=data_input/length(data_input)
  for(i in 1:n_row)

```

```

    prob[i] = sum(p[i,])
  print(prob)
  #print(paste0("Uniform multinomial parameters - n ", list (n=n)))
  #print(paste0("Uniform multinomial parameters - p ",prob))
}
else if (distribution_type=="multivariatenormal")
{
  mean_data <- colMeans(data_input)
  x_sub_mean <- data-mean_data
  cov <- matrix(c(0,0,0,0),2,2)
  for (i in 1:10000) {
    prod <- x_sub_mean[i,] %*% t(x_sub_mean[i,])
    cov <- cov + prod
  }
  covariance <- cov / 10000
  print(covariance)
  print(paste0("Uniform distribution parameters - Mu value is ", mean_data,
    " Sigma value is ", covariance))
}
}

```

## Testing distributions

Testing different distributions below:

```

# Bernoulli is Binomial with size as 1
bernoulli_data <- rbinom(1000, 1, 0.75)
distribution_type = 'Bernoulli'
Parameter_finder(distribution_type,bernoulli_data)

```

```
## [1] "Bernoulli distribution parameters - p value is 0.724"
```

```

#Binomial distribution
binom_data <- rbinom(100, 1000, 0.75)
distribution_type = 'Binomial'
Parameter_finder(distribution_type,binom_data)

```

```
## [1] "Binomial distribution parameters - p value is 7.507"
```

```

# Geometric distribution. Here parameter couldn't be estimated even with really high value of N
geom_data <- rgeom(100000, 0.25)
distribution_type = 'Geometric'
Parameter_finder(distribution_type,geom_data)

```

```
## [1] "Geometric distribution parameters - p value is 0.332927162195455"
```

```

# Poisson distribution. Lambda is accurately estimated
poisson_data <- rpois(40000, lambda = 3)
distribution_type = 'Poisson'
Parameter_finder(distribution_type,poisson_data)

```

```
## [1] "Poisson distribution parameters - Lambda value is 3.00335"
```

```
#Uniform distribution data. Here a and B are accurately estimated if gap between them is large enough.  
#If not,it seems inaccurate
```

```
uniform_data <- runif(1000000, 1, 100)  
distribution_type = 'Uniform'  
Parameter_finder(distribution_type,uniform_data)
```

```
## [1] "Uniform distribution parameters - a value is 1.00011313054711 b value is 99.9999742298387"
```

```
# Normal distribution  
norm_data <- rnorm(100000, 20, 2)  
distribution_type = 'Normal'  
Parameter_finder(distribution_type,norm_data)
```

```
## [1] "Normal distribution parameters - Mean value is 20.0066271044654"
```

```
## [1] "Standard deviation is 2.00539385828563"
```

```
# Exponential distribution  
exp_data <- rexp(100000, 5)  
distribution_type = 'Exponential'  
Parameter_finder(distribution_type,exp_data)
```

```
## [1] "Exponential distribution parameters - rate value is 4.99205457870902"
```

```
# Gamma distribution  
gamma_data <- rgamma(10000, 2, 3)  
distribution_type = 'Gamma'  
Parameter_finder(distribution_type,gamma_data)
```

```
## [1] "Estimate of alpha is 1.98007338730874"
```

```
## [1] "Estimate of beta is 2.97438096380393"
```

```
# Beta distribution  
beta_data <- rbeta(10000, 2, 8)  
distribution_type = 'Beta'  
Parameter_finder(distribution_type,beta_data)
```

```
## [1] "Estimate of alpha is 1.98320520401338"
```

```
## [1] "Estimate of beta is 7.89446294878032"
```

```
#Multinomial Distribution  
p = c(0.20,0.40,0.05,0.30)  
data = rmultinom(10000,size=4,p)  
Parameter_finder("multinomial", data)
```

```
## [1] 0.21230 0.41725 0.05250 0.31795
```

```

# Multi variate normal distribution
Sum = matrix(c(9,6,6,16),2,2)
data = mvrnorm(n = 10000, c(4, 5), Sum)
distribution_type = 'multivariatenormal'
Parameter_finder(distribution_type, data)

```

```

##           [,1]      [,2]
## [1,] 9.582251  6.004751
## [2,] 6.004751 16.076323
## [1] "Uniform distribution parameters - Mu value is 4.00804491351362 Sigma value is 9.5822506437674"
## [2] "Uniform distribution parameters - Mu value is 4.99465011934618 Sigma value is 6.00475084439262"
## [3] "Uniform distribution parameters - Mu value is 4.00804491351362 Sigma value is 6.00475084439262"
## [4] "Uniform distribution parameters - Mu value is 4.99465011934618 Sigma value is 16.0763226488718"

```