

# Package ‘Homework1’

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**Type** Package

**Title** PH140.778 Advanced Statistical Computing HW1

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**Description** This package gives faster ways to fit a linear regression model and compute the multivariate normal density

**License** GPL

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## Description

This package gives faster algorithms to fit a linear regression model and compute the multivariate normal density.

## Details

Package: Homework1  
Type: Package  
Version: 1.0  
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License: GPL

This package mainly uses Cholesky decomposition when computing the inverse matrix and takes advantages of matrix properties as well.

### Author(s)

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### References

PH140.778 Advanced Statistical Computing Dr.Peng

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dmvnorm

*Computing the Multivariate Normal Density MORE Efficiently*

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### Description

This function dmvnorm() evaluates the k-dimensional multivariate Normal density with mean mu and covariance S

### Usage

```
dmvnorm(x, mu, S, log = TRUE)
```

### Arguments

x	a n*k matrix of points to be evaluated
mu	a length k vector of means for the k-dimensional Normal
S	a k*k covariance matrix
log	whether the return should be the log density. Default is TRUE

### Value

This function returns a vector of length n containing the values of the multivariate Normal density evaluated at the n points. If log = TRUE, it returns the log density at those points. If log is NOT TRUE, it will return the density values.

### Author(s)

Lu Li

### References

PH140.778 Advanced Statistical Computing Dr.Peng

## Examples

```
dmvnorm <- function(x, mu, S, log=TRUE) {
  #k dimension multivariate normal
  k=length(mu)
  if(is.matrix(x)==FALSE){
    x=as.matrix(t(x))
  }
  #n data points
  n=nrow(x)

  #check positive definite by trying to Cholesky decomposition
  Q=tryCatch({chol(S)},
    error=function(li){
      message("S cannot be a covariance matrix")
    })

  #compute Q_inverse*(x-mu)
  #note that t(x-mu)%*%inv(Q)%*%t(Q))%*%(x-mu)=crossprod(inv(Q)%*%(x-mu))
  #the easiest way to compute that is to solve t(Q)%*%(inv(Q)%*%(x-mu))=x-mu
  temp1=x-rep(1,n)%*%t(mu)
  A=forwardsolve(t(Q),t(temp1))
  temp2=diag(crossprod(A))

  #compute density
  density=(-k/2)*log(2*pi)-(1/2)*2*sum(log(diag(Q)))-(1/2)*temp2

  #check if log argument
  if(log!=TRUE){
    density=exp(density)
  }
  return(density)
}

n <- 10
n2 <- n^2
xg <- seq(0, 1, length = n)
yg <- xg
g <- data.matrix(expand.grid(xg, yg))
D <- as.matrix(dist(g))
phi <- 5

S <- exp(-phi * D)
mu <- rep(0, n2)
set.seed(1)
x <- matrix(rnorm(n2), byrow = TRUE, ncol = n2)

dmvnorm(x, mu, S, log = TRUE)
```

## Description

Comparing to the `lm.fit()` function in R, this `fastlm()` function presents a much faster way to fit a linear regression model.

## Usage

```
fastlm(X, y, na.rm = FALSE)
```

## Arguments

<code>X</code>	predictor variable, a $n \times k$ matrix
<code>y</code>	response variable, a vector of length $n$
<code>na.rm</code>	argument indicating that whether missing values in <code>X</code> or <code>y</code> should be removed. Default is FALSE

## Details

`fastlm()` takes the advantages of Cholesky decomposition to compute the inverse matrices, which makes a huge improvement in the efficiency of fitting linear regression models.

## Value

	a list of components
<code>coefficients</code>	a vector of the regression coefficients estimated using maximum likelihood
<code>vcov</code>	the $p \times p$ covariance matrix of the estimated regression coefficients

## Author(s)

Lu Li

## References

PH140.778 Advanced Statistical Computing Dr.Peng

## Examples

```
fastlm<-function(X, y, na.rm=FALSE) {
  #check argument na.rm
  if (na.rm!=FALSE) {
    Z=cbind(X,y)
    X=X[complete.cases(Z),]
    y=as.matrix(y)[complete.cases(Z)]
  }
  n<-length(y)
  p<-ncol(X)
  #calculating transpose(X)%*%X
  A<-crossprod(X)
  #calculating transpose(X)%*%y
  C<-crossprod(X,y)

  #cholesky decomposition
```

```

        Q<-chol(A)

        #solve betahat
        temp1<-forwardsolve(t(Q),C)
        betahat<-backsolve(Q,temp1)

        #calculate covirance of beta
        #note that  $t(e) * e = t(e) * y - t(y) * X * \beta$ 
        cov_beta<-chol2inv(Q)*as.numeric(crossprod(y-X%*betahat,y))/(n-p)

        return(list(coefficients=betahat,vcov=cov_beta))
    }

    set.seed(2)
    ## Generate predictor matrix
    n <- 100
    p <- 5
    X <- cbind(1, matrix(rnorm(n * (p - 1)), n, p - 1))

    ## Coefficents
    b <- rnorm(p)

    ## Response
    y <- X%*b + rnorm(n)

    fit <- fastlm(X, y)
    str(fit)

```

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