

# Package ‘Homework1’

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**Type** Package

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

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**Author** Lu Li

**Maintainer** Lu Li<lli48@jhu.edu>

**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  
Date: 2013-11-08  
License: GPL

This package mainly uses Cholesky decomposition when computing the inverse matrix.

**Author(s)**

Lu Li

Maintainer Lu Li <lli48@jhu.edu>

**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 = FALSE, it returns the density values.

**Author(s)**

Lu Li

**References**

PH140.778 Advanced Statistical Computing Dr.Peng

## Examples

```
## The function is currently defined as
function (x, mu, S, log = TRUE)
{
  k = length(mu)
  n = nrow(x)
  Q = tryCatch({
    chol(S)
  }, error = function(li) {
    message("S cannot be a covariance matrix")
  })
  temp1 = x - rep(1, n) %*% t(mu)
  A = forwardsolve(t(Q), t(temp1))
  temp2 = diag(crossprod(A))
  density = (-k/2) * log(2 * pi) - (1/2) * 2 * sum(log(diag(Q))) -
    (1/2) * temp2
  if (log == FALSE) {
    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)
```

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fastlm

*Faster Way to Fit Linear Regression Models*


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## 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 $X$ or $y$ 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

**Examples**

```
function (X, y, na.rm = FALSE)
{
  n <- length(y)
  p <- ncol(X)

  ##Check if missing values in X and y should be removed
  if (na.rm == TRUE) {
    Z = cbind(X, y)
    X = X[complete.cases(Z), ]
    y = as.matrix(y[complete.cases(Z)])
  }
  A <- crossprod(X)
  C <- crossprod(X, y)

  ##Cholesky decomposition
  Q <- chol(A)
  temp1 <- forwardsolve(t(Q), C)
  betahat <- backsolve(Q, temp1)
  cov_beta <- chol2inv(Q) * as.numeric(crossprod(y - X %*%
    betahat)/(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))

## Coefficients
b <- rnorm(p)

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

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

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