

Hard

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In [[https://web.stanford.edu/~hastie/glmnet/glmnet_alpha.html][here]] the author explains the detail of the loss function. The loss function of logistic regression is

$$-\left\{\frac{1}{N} \sum [y_i(\beta_0 + x_i^\top \beta) - \log(1 + e^{(\beta_0 + x_i^\top \beta)})]\right\} + \lambda \left\{\frac{(1-\alpha)}{2} \|\beta\|_2^2 + \alpha \|\beta\|_1\right\}.$$

When $\alpha = 1$ we get what is required.

```
library(Rcpp)
library(RcppArmadillo)
library(inline)
library(ElemStatLearn)
library(testthat)
library(rbenchmark)

# Rcpp function
sourceCpp('C.cpp')
# The C code is as follow

#include <RcppArmadillo.h>
// [[Rcpp::depends(RcppArmadillo)]]
using namespace Rcpp;
using namespace std;
using namespace arma;
// [[Rcpp::export]]
SEXP cpp_l(Mat<double> X,
           vec y,
           double beta_0,
           vec beta,
           double alpha,
           double lambda) {

  double n = X.n_rows;
  vec Xb = beta_0 + X*beta;
  double l = -accu(y%Xb - log(1+exp(Xb)))/n;
  double Omega = lambda *accu((1-alpha)/2*beta%beta + alpha*abs(beta));
  //cout<<l<<endl;
  //cout<<Omega<<endl;
  return wrap(l+Omega);
}

# R implementation of the logistic loss function
r_l <- function(X,y,beta,beta_0,alpha,lambda){
  n <- dim(X)[1]
  Xb <- X%*%beta + beta_0
  l <- -sum(y*Xb - log(1+exp(Xb)))/n
  Omega <- lambda*sum((1-alpha)/2*beta*beta + alpha*abs(beta))
```

```

print(1)
print(Omega)
return(1+Omega)
}

# Start calculation
n <- length(spam$spam)
alpha <- rnorm(1)
lambda <- rnorm(1)
y <- c(1,2)
X <- matrix(c(4,5,6,7),2,2)
beta_0 <- 1
beta <- c(1,2)
all.equal(cpp_l(X,y,beta=beta,beta_0=beta_0,alpha,lambda),
          r_l(X,y,beta=beta,beta_0=beta_0,alpha,lambda))

## [1] -10
## [1] -3.513331
## [1] TRUE

```

I also want to see how efficient C++ implemtation is.

```

print(result)

##                                     test elapsed
## 1 cpp_l(X, y, beta = beta, beta_0 = beta_0, alpha, lambda)    0.02
## 3 cpp_l(X, y, beta = beta, beta_0 = beta_0, alpha, lambda)    0.00
## 5 cpp_l(X, y, beta = beta, beta_0 = beta_0, alpha, lambda)    0.00
## 2   r_l(X, y, beta = beta, beta_0 = beta_0, alpha, lambda)    0.14
## 4   r_l(X, y, beta = beta, beta_0 = beta_0, alpha, lambda)    0.25
## 6   r_l(X, y, beta = beta, beta_0 = beta_0, alpha, lambda)    0.41
## replications
## 1           2000
## 3           2000
## 5           2000
## 2           2000
## 4           2000
## 6           2000

```