STAT600 HW1

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Problem 2

Link to my .tar.gz file

Problem 3

Simulate Dataset

```
set.seed(1)
Dataset100 <- foreach(i = 1:100) %do% {
    x <- rnorm(100)
    e <- rnorm(100)
    y <- 1 - x + e
    list(x = x, y = y)
}</pre>
```

Base R lm()

```
# Base lm()
lm_base <- foreach(i = 1:100) %dopar% {
    x <- Dataset100[[i]]$x
    y <- Dataset100[[i]]$y
    elapsed <- system.time({mod.base <- lm(y~x)})[3]
    coef <- c(mod.base$coefficients)
    beta0_ci <- confint(mod.base)[1,]
    beta1_ci <- confint(mod.base)[2,]
    yhat <- mod.base$fitted.values
    resids <- mod.base$fitted.values
    resids <- mod.base$residuals
    list(time=elapsed,coef=coef,beta0_ci=beta0_ci,beta1_ci=beta1_ci,yhat=yhat,resids=resids)
}
stopImplicitCluster()</pre>
```

SimpLinCpp

```
lm_cpp <- foreach(i = 1:100) %dopar% {
    x <- Dataset100[[i]] $x
    y <- Dataset100[[i]] $y
    elapsed <- system.time({mod.cpp <- SimpLinR(x,y)})[3]
    coef <- c(mod.cpp$Coefficients)
    beta0_ci <- mod.cpp$ 95% CI [1,]
    beta1_ci <- mod.cpp$ 95% CI [2,]
    yhat <- mod.cpp$Predicted Values'
    resids <- mod.cpp$Residuals
    list(time=elapsed,coef=coef,beta0_ci=beta0_ci,beta1_ci=beta1_ci,yhat=yhat,resids=resids)
}
stopImplicitCluster()</pre>
```

Problem 4

Both the base R lm() and the Cpp SimpLin produce same result (in terms of bias, coverage probability, mean squared error for the regression coefficients as well as predictive mse for \hat{y}) like they should be. Time-wise, the cpp is faster than the base R lm().

A table comparing the two methods' performance is included below, as well as histogram of the estimated coefficients.

Table 1: Performace Comparison between Base R and Rcpp

	Base R	Rcpp
Avg Time (secs)	0.0007500	0.0002200
Bias: b0	-0.0066561	-0.0066561
Bias: b1	0.0013153	0.0013153
Empirical Coverage: b0	0.9600000	0.9600000
Empirical Coverage: b1	0.9500000	0.9500000
MSE: b0	0.0076562	0.0076562
MSE: b1	0.0104577	0.0104577
Prediction MSE	0.9663379	0.9663379



