

Problem 1

C code:

```

#include<stdio.h>

void dgesv_(int *n, int *NRHS, double *A, int *LDA,
            int *IPIV, double *B, int *LDB, int *INFO);

void dgemm_(char *TRANSA, char *TRANSB, int *M, int *n, int *K,
            double *ALPHA, double *A, int *LDA, double *B,
            int *LDB, double *BETA, double *C, int *LDC);

int main(int argc, char *argv[]) {
    FILE *fp;
    int N=0, P1=1, i, j, k;
    char tmp;
    fp = fopen(argv[1], "r");
    while((tmp=fgetc(fp))!=EOF){
        if(tmp=='\n')    N++;
    }
    rewind(fp);
    while((tmp=fgetc(fp))!= '\n'){
        if(tmp==' ') P1++;
    }
    rewind(fp);
    printf("Sample size and number of predictors are %d and %d respective\n", N, P1);

    double Y[N];
    if(atoi(argv[2])==1) P = P1;
    else P = P1 - 1;
    double X[N][P];
    for(i=0; i<N; i++){
        for(j=0; j<P1; j++){
            if(j==0){
                fscanf(fp, "%lf", &Y[i]);
                if(atoi(argv[2])==1) X[i][0] = 1;
            }
            else{
                if(atoi(argv[2])==1)
                    fscanf(fp, "%lf", &X[i][j]);
                else fscanf(fp, "%lf", &X[i][j-1]);
            }
        }
    }
}

```

```

    }
}

double x[N*P], xt[P*N], Design_Mat[P*P];
int n1=N, n2=P, n3=1, info, ipiv[P];
double coef[P], beta=0, alpha=1;
char TRANSA = 'N', TRANSB='N';

for (i=0; i<N; i++){
    for (j=0; j<P; j++){
        x[i*P+j] = X[i][j];
        xt[j*N+i] = X[i][j];
    }
}

dgemm_(&TRANSA, &TRANSB, &n2, &n2, &n1, &alpha,
      x, &n2, xt, &n1, &beta, Design_Mat, &n2);
dgemm_(&TRANSA, &TRANSB, &n2, &n3, &n1, &alpha,
      x, &n2, Y, &n1, &beta, coef, &n2);
dgesv_(&n2, &n3, Design_Mat, &n2, ipiv, coef, &n2, &info);

printf("The regression coefficients: \n");
for (i=0; i<P; i++){
    printf("%f \n", coef[i]);
}
printf("\n");
return(0);
}

```

Problem 2

(a)

$$\int_0^{\infty} (x^2 + 5)xe^{-x}dx \approx 10.99763 \pm 0.2028816$$

(b)

$$\int_0^1 \int_{-\infty}^{\infty} e^{-x^2} \cos(xy) dx dy \approx 1.634335$$

(c)

$$\int_0^{\infty} \frac{3}{4} x^4 e^{-x^{3/4}} dx \approx 2.271137 \pm 0.00957$$

R code:

```
## (a)
N <- 100000
set.seed(20170311)
rand_exp <- rexp(N, rate = 1)
mean((rand_exp^2+5)*rand_exp)
sqrt(var((rand_exp^2+5)*rand_exp)/N)*1.96

## (b)
set.seed(20170311)
rand_norm <- rnorm(N, mean = 0, sd = 1/sqrt(2))
rand_unif <- runif(N, min = 0, max = 1)
mean(cos(rand_norm*rand_unif)*sqrt(pi))

## (c)
set.seed(20170311)
rand_weibull <- rweibull(N, shape = 3, scale = 4^(1/3))
mean(rand_weibull^2)
sqrt(var(rand_weibull^2)/N)*1.96
```

Problem 3

- $\nu = 0.1$: $I \approx 0.185114 \pm 0.07967168$;
- $\nu = 1$: $I \approx 0.1363114 \pm 0.00121565$;
- $\nu = 10$: $I \approx 0.1386195 \pm 0.004512526$.

R code:

```
##
v = 0.1
N <- 100000
set.seed(20170311)
rand_norm <- rnorm(N, mean = 1.5, sd = v)
mean(v*exp(-rand_norm^2/2)/exp(-(rand_norm-1.5)^2/(2*v^(2))))*
  ifelse(rand_norm<2&&rand_norm>1,1,0))
```

```

sqrt(var(v*exp(-rand_norm^2/2)/exp(-(rand_norm-1.5)^2/(2*v^(2))))*
      ifelse(rand_norm<2&rand_norm>1,1,0))/N)*1.96

## v = 1
v = 1
set.seed(20170311)
rand_norm <- rnorm(N, mean = 1.5, sd = v)
mean(v*exp(-rand_norm^2/2)/exp(-(rand_norm-1.5)^2/(2*v^(2))))*
      ifelse(rand_norm<2&rand_norm>1,1,0))
sqrt(var(v*exp(-rand_norm^2/2)/exp(-(rand_norm-1.5)^2/(2*v^(2))))*
      ifelse(rand_norm<2&rand_norm>1,1,0))/N)*1.96

## v = 10
v = 10
set.seed(20170311)
rand_norm <- rnorm(N, mean = 1.5, sd = v)
mean(v*exp(-rand_norm^2/2)/exp(-(rand_norm-1.5)^2/(2*v^(2))))*
      ifelse(rand_norm<2&rand_norm>1,1,0))
sqrt(var(v*exp(-rand_norm^2/2)/exp(-(rand_norm-1.5)^2/(2*v^(2))))*
      ifelse(rand_norm<2&rand_norm>1,1,0))/N)*1.96

```

Problem 4

- (a) $\hat{I} = 0.6977335$
 (b) $E\{c(U)\} = 1.5$. The optimal value of $b = -0.4779221$. $\hat{I}_{CV} = 0.6927941$.
 (c)

$$\widehat{var}(\hat{I}_{MC}) = 1.295893 \times 10^{-5} > \widehat{var}(\hat{I}_{CV}) = 4.006056 \times 10^{-7}$$

- (d) We define a new function $c_1(x) = \sqrt{1+x}$. The new estimator is denoted by

$$\hat{I}_{new} = \frac{1}{n} \sum_{i=1}^n h(U_i) - b \left[\frac{1}{n} \sum_{i=1}^n c_1(U_i) - E\{c_1(U)\} \right].$$

The variance of \hat{I}_{new} is smaller than the variance of \hat{I}_{CV} because

$$\begin{aligned} \left| \text{Corr} \left(\frac{1}{1+x}, \sqrt{1+x} \right) \right| &= \frac{E \frac{1}{\sqrt{1+x}} - E \frac{1}{1+x} E \sqrt{1+x}}{\sqrt{\text{var}(\frac{1}{1+x}) \text{var}(\sqrt{1+x})}} = 0.990998 \\ \left| \text{Corr} \left(\frac{1}{1+x}, 1+x \right) \right| &= \frac{E 1 - E \frac{1}{1+x} E(1+x)}{\sqrt{\text{var}(\frac{1}{1+x}) \text{var}(1+x)}} = 0.9841661 < 0.990998 \end{aligned}$$

R code:

```
## (a)
```

```
n = 1500
```

```
set.seed(20170311)
```

```
rand_unif <- runif(n, 0, 1)
```

```
mean(1/(1+rand_unif))
```

```
## (b)
```

```
b_opt <- sum((1/(1+rand_unif) - mean(1/(1+rand_unif))) *  
            ((1+rand_unif) - mean((1+rand_unif)))) /  
            (var(1+rand_unif) * (n-1))
```

```
mean(1/(1+rand_unif)) - b_opt * (mean(1+rand_unif) - 1.5)
```

```
## (c)
```

```
var(1/(1+rand_unif)) / n
```

```
var(1/(1+rand_unif) - b_opt * (1+rand_unif)) / n
```

```
## (d)
```

```
(1 - log(2) * 1.5) / ((1/2 - log(2)^2) / 12)^(1/2)
```

```
(2 * sqrt(2) - 2 - log(2) * ((4 * sqrt(2) - 2) / 3)) /
```

```
((1/2 - log(2)^2) * (1.5 - ((4 * sqrt(2) - 2) / 3)^2))^(1/2)
```