

Package demo

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Here is an example of how our package functions run. For our data set, we are using a “SGEMM GPU kernel performance Data Set,” which measures the running times of a matrix-matrix product, given different parameter combinations.

```
library(devtools)
library(tidyverse)
library(STA141CFinal)

dat = read_csv("sgemm_product.csv")

#We specifiy a specific column set
y = dat$`Run1 (ms)`
x = dat[,1:(ncol(dat)-4)]

#linear model objects
fit = linear_reg_bs(x = x, y = y, s = 10, r = 1000)
#fit$bootstrap_coefficient_estimates

fit2 = linear_reg_bs_par(x = x, y = y, s = 10, r = 1000)
#fit2$bootstrap_coefficient_estimates
```

Linear Regression with blb

95 % Confidence Interval for Variable Coefficients

```
coef_CI(fit, alpha = 0.05)
```

##	Lower_Bounds	Estimates	Upper_Bounds
## Intercept	-66.910978	218.738529060	513.421889
## MWG	-1.260043	0.006954464	1.288869
## NWG	-1.261053	-0.004824181	1.275299
## KWG	-6.341448	0.005717980	6.249883
## MDIMC	-6.667762	-0.093852118	6.856528
## NDIMC	-6.427916	-0.029233029	6.974446
## MDIMA	-5.362095	0.035059915	5.744351
## NDIMB	-5.403451	0.004079777	5.719580
## KWI	-16.113086	-0.093264542	15.951679

```
## VWM          -27.024223 -0.078427197 29.610773
## VWN          -26.135769  0.280068953 30.190972
## STRM         -94.809131  0.384049951 97.708665
## STRN         -95.747251 -0.164080409 96.485461
## SA           -95.888917 -1.007278073 95.918019
## SB           -96.725990  0.558883883 98.466003
```

```
coef_CI_par(fit,alpha = 0.05)
```

```
##           Lower_Bounds      Estimates Upper_Bounds
## Intercept -66.910978 218.738529060 513.421889
## MWG        -1.260043  0.006954464  1.288869
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```

```
(b1 = bench::mark(
  coef_CI(fit, alpha = 0.05),
  coef_CI_par(fit,alpha = 0.05))
)
```

```
## # A tibble: 2 x 6
##   expression          min    median `itr/sec` mem_alloc `gc/sec`
##   <bch:expr>      <bch:tm> <bch:tm>      <dbl>   <bch:byt>   <dbl>
## 1 coef_CI(fit, alpha = 0.05)    31.9ms  35.2ms      28.1    7.49MB     7.66
## 2 coef_CI_par(fit, alpha = 0.05) 53.6ms  54.9ms      17.5    7.8MB      8.76
```

Notice that `coef_CI_par` offers better memory allocation than `coef_CI`.

```
plan(multiprocess, workers = 4)
PI(fit, dat[1:3, 1:14], alpha = 0.05)
PI_par(fit, dat[1:3, 1:14], alpha = 0.05)

(b2 = bench::mark(
  PI(fit, x, alpha = 0.05),
  PI_par(fit, x, alpha = 0.05))
)
```

95 % Confidence Interval for Variance

```
s2_CI(fit, alpha = 0.05)
```

```
## Lower_Bound    Estimate Upper_Bound
##      621.7156    1282.3401    2133.3777
```

```
s2_CI_par(fit, alpha = 0.05)
```

```
## Lower_Bound    Estimate Upper_Bound
##      621.7156    1282.3401    2133.3777
```

```
(b3 = bench::mark(
  s2_CI(fit, alpha = 0.05),
  s2_CI_par(fit, alpha = 0.05))
)
```

```
## # A tibble: 2 x 6
##   expression          min    median `itr/sec` mem_alloc `gc/sec`
##   <bch:expr>      <bch:tm> <bch:tm>      <dbl> <bch:byt>      <dbl>
## 1 s2_CI(fit, alpha = 0.05)    2.16ms  2.34ms    398.    118KB      6.74
## 2 s2_CI_par(fit, alpha = 0.05) 30.62ms 32.55ms    29.6    413KB      7.39
```

Notice that `s2_CI_par` offers better memory allocation than `s2_CI`.