final-project-vignette

The blblm package is helpful for making linear models using the Bag of Little Bootstraps (BLB) algorithm. BLB is a good way to deal with bootstrapping for big data, so we provide some examples of how to use it over traditional bootstrap methods here for the user's understanding.

First, please make sure the package is installed. Please run install.packages(blblm) if needed. Note that the blblm package has dependencies with the following packages: purrr, stats, magrittr %>%, parallel, and furrr

```
library(blblm)
library(future)
```

We start by creating a a BLB Linear Model with the blblm function. We will be using the Theoph datset for demonstration purposes.

We can use blb.1m to create a BLB Linear Model. To speed up computation, we can set parallelize = TRUE to allow the user to use more cores for faster calculations. Please note that the number of cores available to use depends on the user's computer.

We can view our model with the print.blblm function.

```
print(blb.lm)
#> blblm model: Wt ~ Dose * conc
```

Using this model we created, we can get summary statistics and model information. For example, we can get the coefficients using <code>coef(blb.lm)</code>

We will benchmark to see how the parallelized version compares to a non-parallelized version for a dataset

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we generate here.

```
new.x <- runif(10000)
new.x2 <- rnorm(10000)
new.y <- rpois(10000, 5)

new.df <- data.frame(new.x, new.x2, new.y)

bench::mark(
    coef(blblm(new.y ~ new.x * new.x2, m = 10, data = new.df, parallelize = TRUE)),
    coef(blblm::blblm(new.y ~ new.x * new.x2, m = 10, data = new.df, parallelize = FALSE)),
    check = FALSE
)

#> Warning: Some expressions had a GC in every iteration; so filtering is disabled.

#> # A tibble: 2 x 6

#> # ... with 6 more variables: expression <br/>
#> # median <br/>
## median <br/
```

We see that the parallelized version is faster, so it should be used when datasets are large. For smaller datasets, parallelization may be slower than a non-parallelized version.

Another common use of a BLB'd Linear Model is to create confidence intervals for certain parameters.

```
confint(blb.lm, c("Dose", "conc"))
#> 2.5% 97.5%
#> Dose -14.905255 -13.7141643
#> conc -1.534285 -0.4674047
```

We can calculate σ for our model with the sigma function.

```
sigma(blb.lm)
#> [1] 0.852151
```

We can also create a confidence interval for σ with the confidence parameter.

We can also predict values using new parameters with the predict function.

```
predict(blb.lm, data.frame(Dose = c(3.9, 4.0), conc = c(1, 1.5), Wt = c(80, 85)))
#> 1     2
#> 79.69536 78.26787
```

Additionally, we can create confidence intervals for these parameters.

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We also test our speed for model creation and compare the c++ version fastlm with base R lm.

```
x <- runif(100)
y <- runif(100)
df \leftarrow data.frame(x = x, y = y)
bench::mark(
  blblm(y ~x, data = df, m = 10, use_cpp = FALSE, B = 1000),
  blblm(y \sim x, data = df, m = 10, use\_cpp = TRUE, B = 1000),
  check = FALSE
)
#> # A tibble: 2 x 6
    expression
                                                                        min
                                                                              median
    <bch:expr>
#>
                                                                   <bch:tm> <bch:tm>
\#>1 blblm(y \sim x, data = df, m = 10, use_cpp = FALSE, B = 1000)
                                                                      669ms
                                                                               669ms
\# 2 blblm(y ~ x, data = df, m = 10, use cpp = TRUE, B = 1000)
                                                                      201ms
                                                                               219ms
#> # ... with 3 more variables: itr/sec <dbl>, mem_alloc <bch:byt>, gc/sec <dbl>
```

For more information on functions in this package, please run <code>?[function_name]</code> to see the documentation page we have updated via ROxygen. These documentations were created for each function in this package to be used. Our functions tend to need <code>data.frame</code> objects and often call on the model we generate with the <code>blblm</code> function. There are some under the hood functions in the package that are not exported, but still documented for clarity's sake.

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