Data Extraction Evaluation

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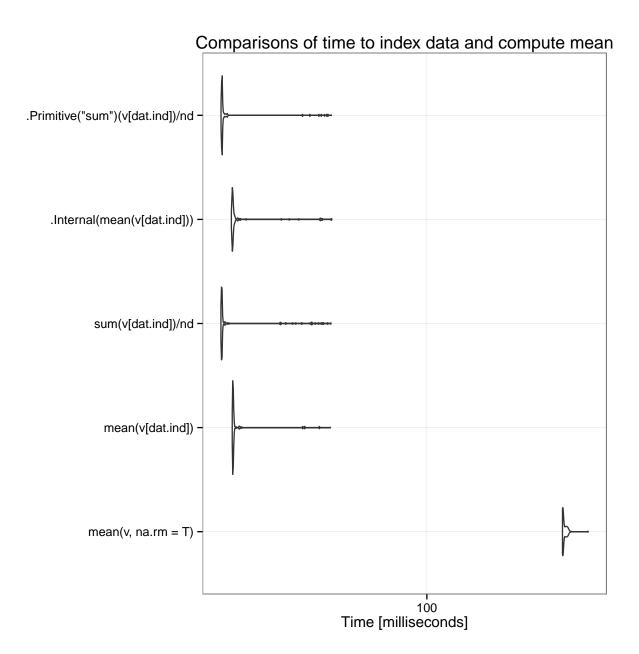
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0.1 Next steps

Combining sampling and data reduction methods while using the most efficient \mathbf{R} functions can be particularly useful when processing large numbers of high-resolution geotiff raster layers. One thing I already do when extracting from many files by shapefile is I avoid extracting by shape more than once. I do it one time to obtain the corresponding raster layer cell indices. Then on all subsequent maps I extract by cell indices which is notably faster. Ultimately, there is much more room for speed improvements in terms of efficient use of statistics than in strictly programmatic corner-cutting.

The plots below benchmark different sample mean computations. Comparisons involve the sample mean of the entire data set and do not involve the main approach outlined above which focuses on efficiency gains by taking the mean of a smaller, representative sample. This provides some insight into how it is beneficial nonetheless to considering the right programmatic approach in conjunction with statistical efficiencies.

```
Unit: milliseconds
##
                                       393.64885 394.81215 401.46656 395.44489
                     mean(v[dat.ind])
##
                                                              14.87471
##
                   sum(v[dat.ind])/nd
                                        12.44504
                                                   12.54393
                                                              15.47461
                                                                         12.59665
##
         .Internal(mean(v[dat.ind]))
                                        13.80660
                                                   13.98169
                                                              15.74200
                                                                         14.04980
    .Primitive("sum")(v[dat.ind])/nd
                                                              14.37262
                                        12.43477
                                                   12.56648
##
                     max neval
                37.87314
##
     12.68451
                38.37634
                38.26842
     12.67938
```



```
## Unit: milliseconds
## expr min lq mean median uq
## mean(v[dat.ind]) 13.893058 14.087589 16.219392 14.191308 14.419740
## sum(v[dat.ind])/nd 12.397452 12.577522 14.224463 12.666780 12.864421
## mean(d) 2.947051 2.970065 3.041987 2.997433 3.041284
## sum(d)/nd 1.483478 1.496229 1.528518 1.503382 1.529817
## max neval
## 129.023621 1000
## 41.267096 1000
## 4.709807 1000
## 3.047504 1000
```

