

Short benchmark of Fleiss' kappa and Brennan-Prediger

Benchmark for the aggregated functions

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The most feature complete R package for agreement coefficients is `irrCAC`. It implements Fleiss' kappa and the Brennan-Prediger coefficient for both aggregated and long form data. Despite the fact that their method of inference, based on U -statistics, does not look the same as ours (based on moments) they are equivalent in the case of aggregated data. Compare the confidence intervals below to be convinced of this.

```
library("quadagree")
irrCAC::bp.coeff.dist(dat.fleiss1971, weights = "quadratic")
```

	coeff.name	coeff	stderr	conf.int	p.value	pa	pe
1	Brennan-Prediger	0.3338889	0.1036175	(0.122,0.546)	0.003134299	0.8334722	0.75

```
bpci_aggr(dat.fleiss1971)
```

Call: call

95% confidence interval (n = 30).

	0.025	0.975
	0.1219674	0.5458104

Sample estimates.

	kappa	sd
	0.3338889	0.5579971

For aggregated data, `quadagree` supports user-supplied `values` vectors, transforms, and studentized bootstrapping, which `irrCAC` does not. But is it faster? It turns out it is roughly twice as fast for reasonable numbers of raters. This suggests there is no benefit in using the moment formulation (as done in `quadagree`) when calculating these coefficients, as `quadagree` is likely to be more optimized than `irrCAC`.

Benchmarks

We will run three benchmarks of various sizes using the `microbenchmark` package. We start off with `dat.fleiss1971`, which contains $n = 30$ rows.

```
x = dat.fleiss1971
irr_bp <- \ (x) irrCAC::bp.coeff.dist(x, weights = "quadratic")
irr_fleiss <- \ (x) irrCAC::fleiss.kappa.dist(x, weights = "quadratic")
microbenchmark::microbenchmark(
  irr_bp(x),
  bpci_aggr(x),
  irr_fleiss(x),
  fleissci_aggr(x),
  times = 1000
)
```

Unit: microseconds

	expr	min	lq	mean	median	uq	max	neval
	irr_bp(x)	280.1	311.20	379.1637	327.60	369.35	3863.9	1000
	bpci_aggr(x)	104.5	131.10	163.1652	145.50	163.70	1950.9	1000
	irr_fleiss(x)	284.5	323.50	386.6799	340.40	387.80	4610.3	1000
	fleissci_aggr(x)	116.5	142.05	179.5976	158.25	179.85	3079.4	1000

So `quadagree` is roughly twice as fast. Let's see what happens when $n = 300$.

```
x = dat.fleiss1971
x = rbind(x, x, x, x, x, x, x, x, x, x)
microbenchmark::microbenchmark(
  irr_bp(x),
  bpci_aggr(x),
  irr_fleiss(x),
  fleissci_aggr(x),
  times = 1000
)
```

Unit: microseconds

	expr	min	lq	mean	median	uq	max	neval
irr_bp(x)	304.6	344.15	411.8070	367.80	410.70	3977.3	1000	
bpci_aggr(x)	110.8	138.90	170.4819	154.60	176.60	1890.5	1000	
irr_fleiss(x)	320.3	356.45	433.3315	380.25	433.90	4988.6	1000	
fleissci_aggr(x)	129.2	161.05	209.1645	178.20	204.35	4142.2	1000	

The run time is almost the same for all methods as it was for $n = 30$, suggesting that there is substantial overhead to both methods. Let's check $n = 3000$.

```
# recall that x has 300 elements.
x = rbind(x, x, x, x, x, x, x, x, x, x)
microbenchmark::microbenchmark(
  irr_bp(x),
  bpci_aggr(x),
  irr_fleiss(x),
  fleissci_aggr(x),
  times = 1000
)
```

Unit: microseconds

	expr	min	lq	mean	median	uq	max	neval
irr_bp(x)	537.1	751.15	1020.6125	796.65	964.60	57192.8	1000	
bpci_aggr(x)	193.8	273.35	344.1337	296.30	344.25	6094.7	1000	
irr_fleiss(x)	559.0	799.80	1037.9914	844.05	998.35	9290.1	1000	
fleissci_aggr(x)	260.6	379.90	472.3803	406.55	469.00	6021.9	1000	

It appears that `bpci_aggr` is pulling ahead of `irrCAC::bp.coeff.dist`.

Let's finish off with a larger number of categories.

```
# recall that x has 3000 elements.
x = cbind(x, x, x, x, x, x, x, x, x, x)
microbenchmark::microbenchmark(
  irr_bp(x),
  bpci_aggr(x),
  irr_fleiss(x),
  fleissci_aggr(x),
  times = 1000
)
```

Unit: microseconds

	expr	min	lq	mean	median	uq	max	neval
	irr_bp(x)	4493.7	4859.95	6186.148	5318.00	6854.55	54246.6	1000
	bpci_aggr(x)	724.8	851.85	1136.090	911.90	1073.80	6765.2	1000
	irr_fleiss(x)	4658.3	5047.55	6339.300	5429.15	7318.15	33798.4	1000
	fleissci_aggr(x)	786.1	925.90	1193.103	981.90	1139.40	8777.2	1000

So `quadegree` could be substantially faster on data with very many categories and items rated.