

Barcelona crime longitudinal data quality

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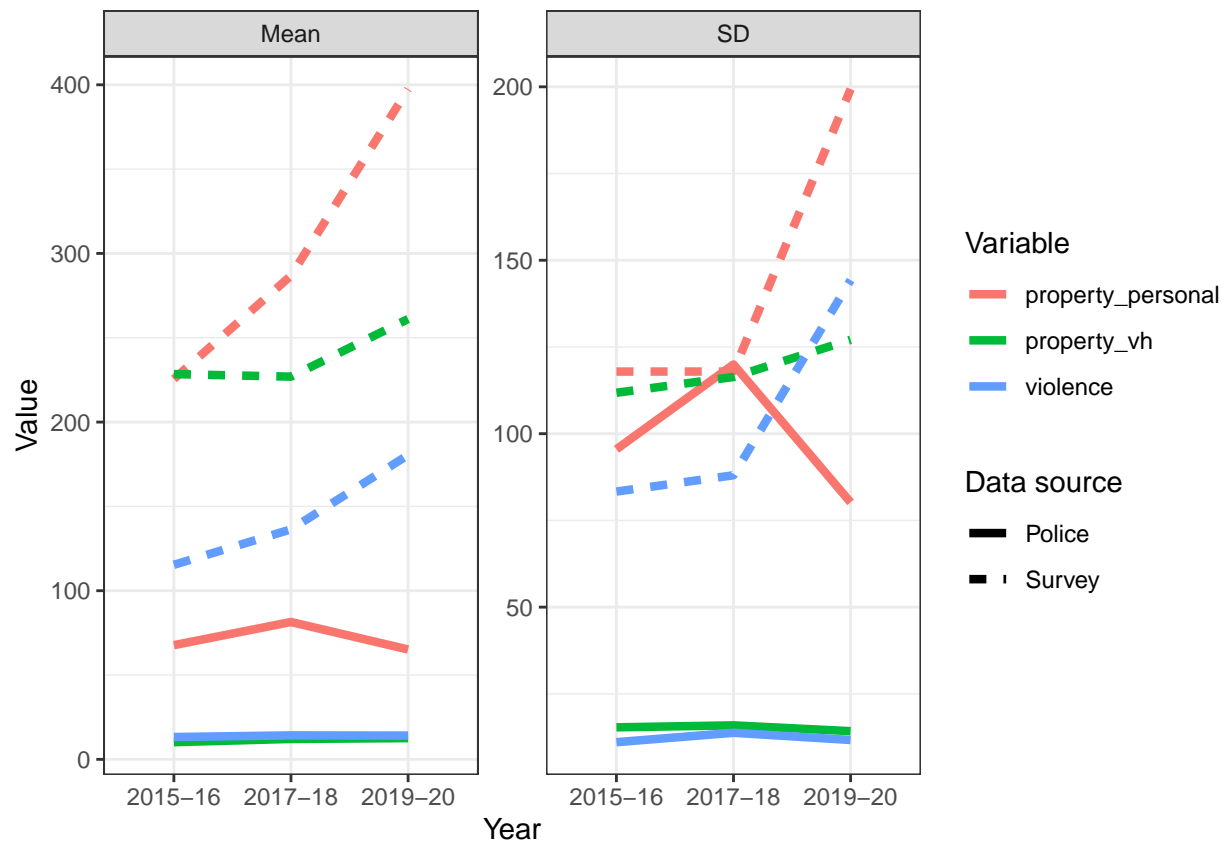
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Here I explore the Barcelona data that has three types of crimes: vehicle, property personal and violence collected in surveys and official data over regions over 6 years. We group years by two in order to avoid having regions with 0s.

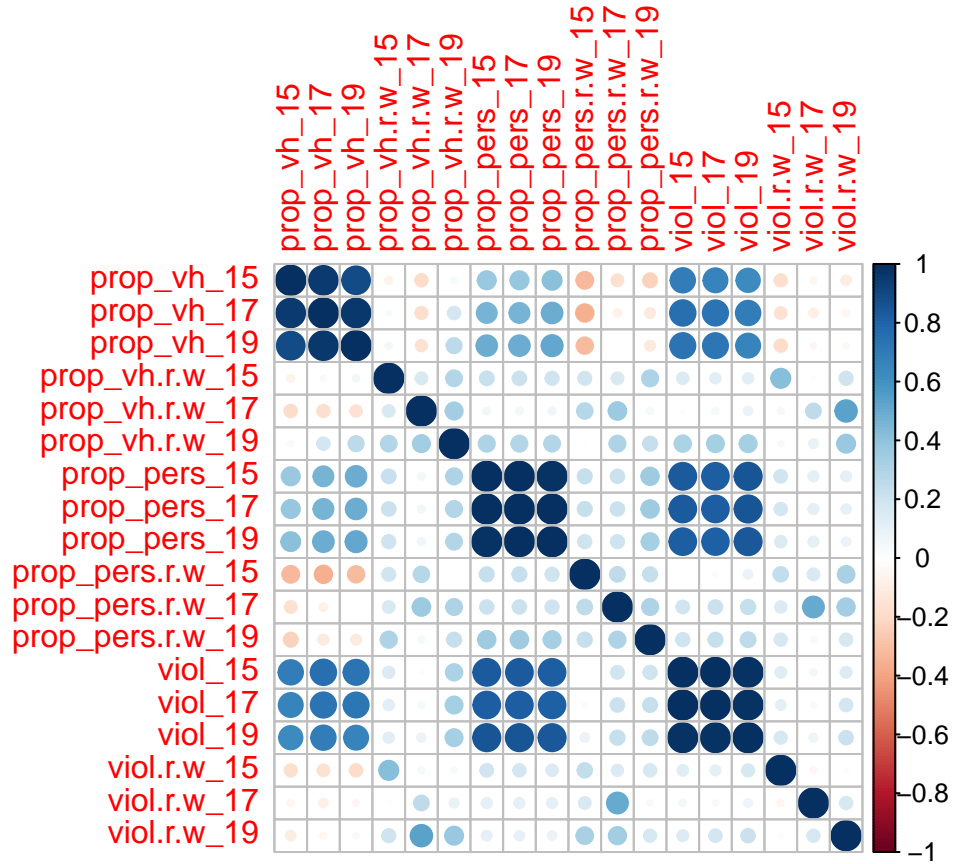
Here I concentrate on the weighted estimates from the survey (ending in “r.w”) and the official data.

Descriptives

First some descriptives. Bellow we observe quite big differences by data source.

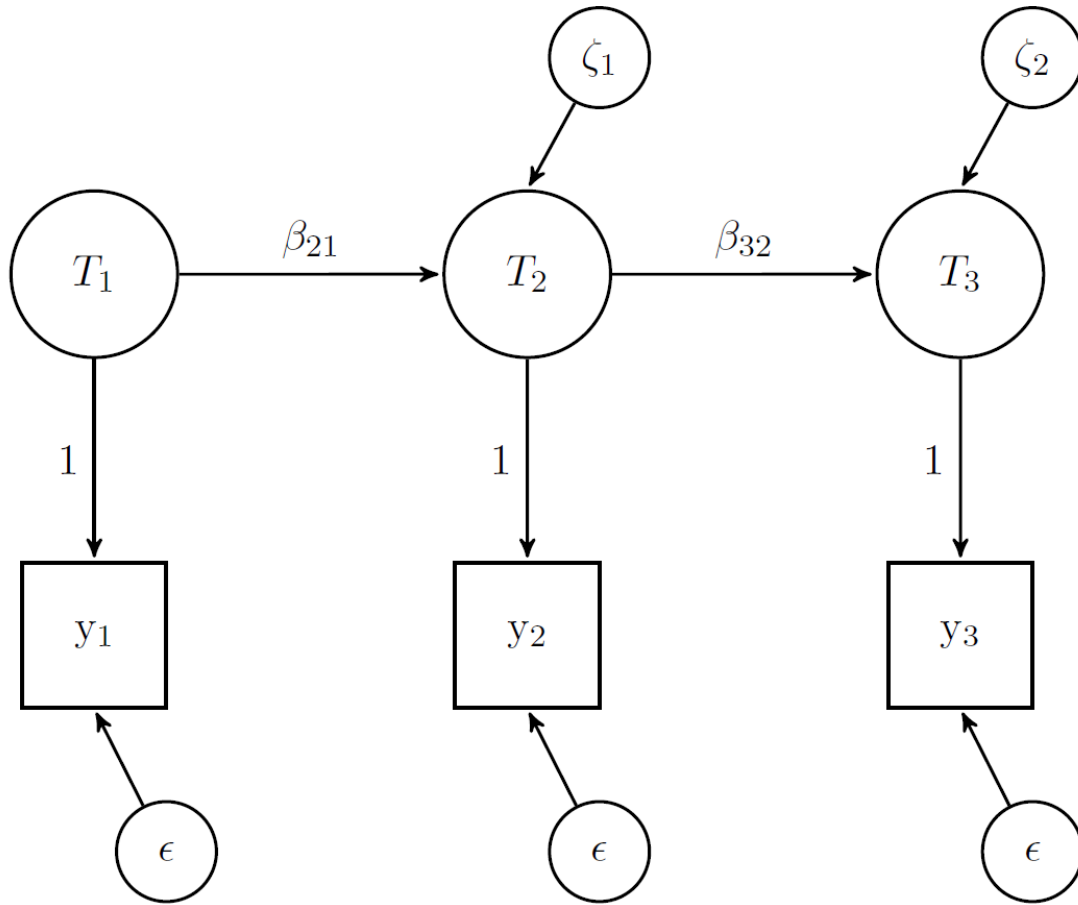


When we look at correlations also we see pretty striking patterns. First of all the consistency within measure is much higher for police data than survey data. Then, the relationship between of measures across data sources is very low. This could be problematic for any modeling.



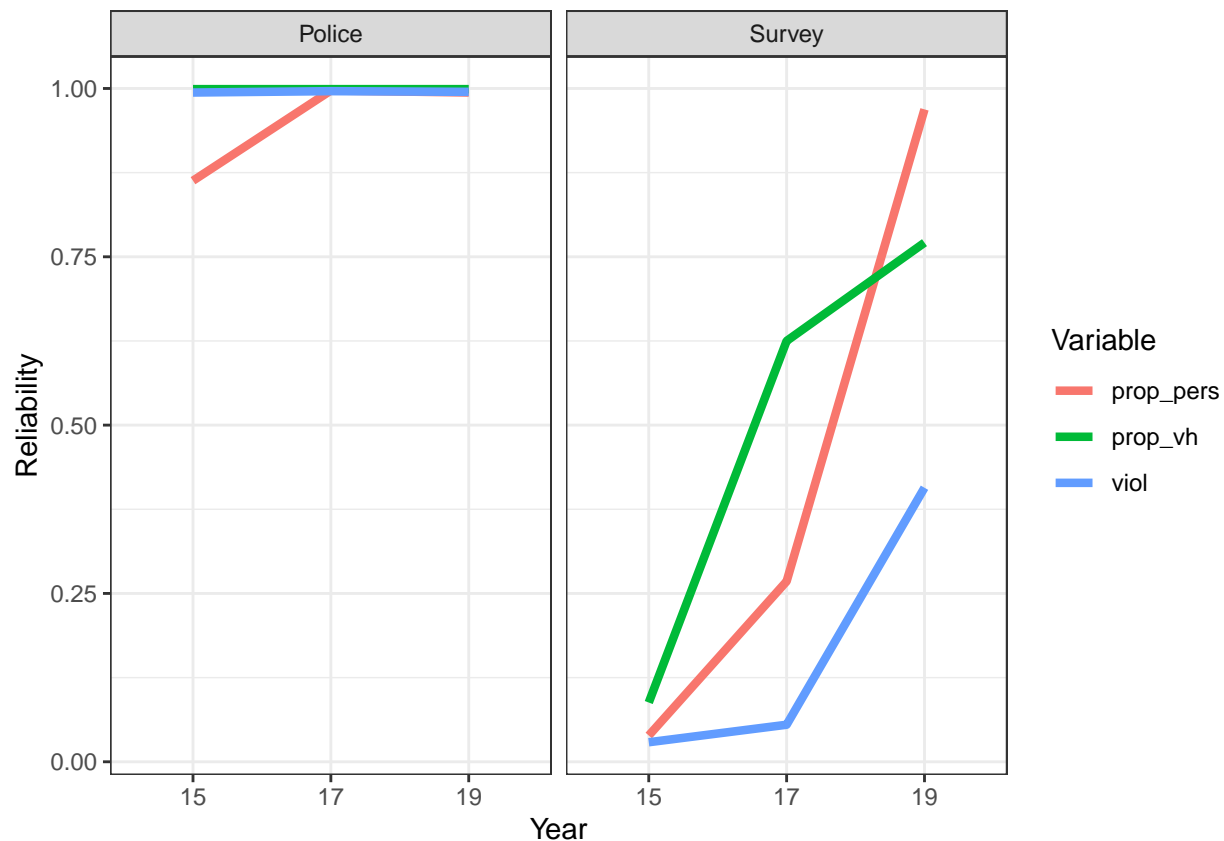
Quasi-simplex

A first way to look at the quality of the data is to use the quasi-simplex model. There assumes an auto-regressive model of true scores and estimates reliability by assuming equal variance of error over time (see book chapter on assumptions of Quasi-simplex for more info).



I estimate the models using **blavaan** which does in the background SEM using Stan for estimation. I tried to use ML but it leads to negative variances (relative common occurrence for these models).

Here we plot the reliability by variable, data source and wave as estimated by quasi-simplex. Reliabilities are extremely for the survey data (as expected given the correlation matrices above).

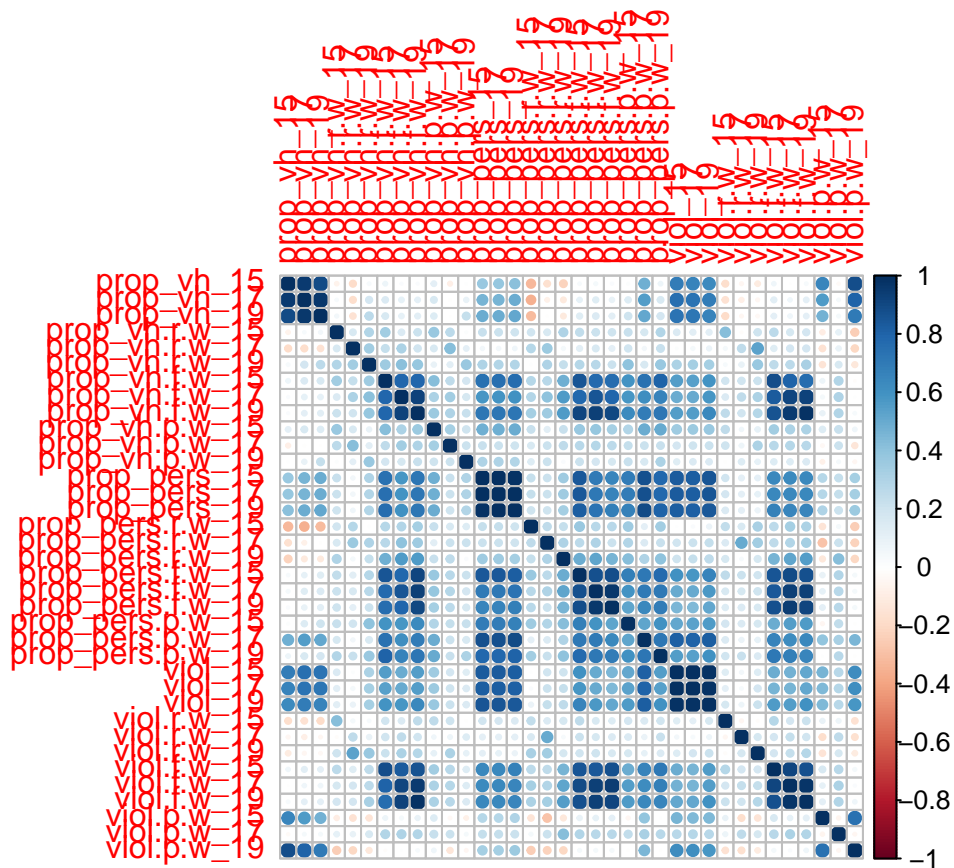


This is also obvious if we average reliability estimates.

group	var	reliability	reliability_source
Police	prop_pers	0.95	0.98
Police	prop_vh	1.00	0.98
Police	viol	1.00	0.98
Survey	prop_pers	0.43	0.36
Survey	prop_vh	0.49	0.36
Survey	viol	0.16	0.36

Given this and the point David made in a recent email let's have a look at survey data about places where crimes happen ("f.w") and looking only at crimes reported to the police ("p.w").

The quasi-simplex does not work for the "f.w" (determinant of covariance is 0 for some reason). So here I compare the original survey measures with the "p.w" ones.



```
## blavaan (0.3-15) results of 1000 samples after 500 adapt/burnin iterations
##
##   Number of observations              73
##
##   Number of missing patterns          1
##
##   Statistic           MargLogLik       PPP
##   Value              -1447.154       0.134
##
## Latent Variables:
##           Estimate Post.SD pi.lower pi.upper   Std.lv   Std.all
##   t1 =~
##     prop_vh.p.w_15    1.000                2.581    0.095
##   t2 =~
##     prop_vh.p.w_17    1.000                5.640    0.203
##   t3 =~
##     prop_vh.p.w_19    1.000               31.275    0.755
##     Rhat      Prior
##
##     NA
##
##     NA
##
##     NA
##
## Regressions:
```

```

##          Estimate Post.SD pi.lower pi.upper Std.lv Std.all
## t2 ~
## t1          1.868   6.974   -14.77   14.498   0.855   0.855
## t3 ~
## t2          5.521   3.499    1.824   14.835   0.996   0.996
## Rhat Prior
##
## 1.002 normal(0,10)
##
## 1.001 normal(0,10)
##
## Intercepts:
##          Estimate Post.SD pi.lower pi.upper Std.lv Std.all
## .prop_vh.p.w_15 38.329   3.171    32.2   44.568  38.329   1.404
## .prop_vh.p.w_17 39.811   3.347    33.13  46.502  39.811   1.434
## .prop_vh.p.w_19 45.034   4.774    35.86  54.047  45.034   1.087
## t1          0.000
## .t2          0.000
## .t3          0.000
## Rhat Prior
## 1.001 normal(0,32)
## 1.001 normal(0,32)
## 1.000 normal(0,32)
## NA
## NA
## NA
##
## Variances:
##          Estimate Post.SD pi.lower pi.upper Std.lv Std.all
## .prp_v._15 (a) 739.037  85.542  594.252  923.993 739.037   0.991
## .prp_v._17 (a) 739.037  85.542  594.252  923.993 739.037   0.959
## .prp_v._19 (a) 739.037  85.542  594.252  923.993 739.037   0.430
## t1          6.660  12.837    0.009   50.882   1.000   1.000
## .t2          8.576  15.031    0.004   53.815   0.270   0.270
## .t3          8.590  20.263    0.002    58.72   0.009   0.009
## Rhat Prior
## 1.001 gamma(1,.5)[sd]
## 1.001
## 1.001
## 1.005 gamma(1,.5)[sd]
## 1.000 gamma(1,.5)[sd]
## 1.001 gamma(1,.5)[sd]
##
## blavaan (0.3-15) results of 1000 samples after 500 adapt/burnin iterations
##
## Number of observations          73
##
## Number of missing patterns      1
##
## Statistic          MargLogLik          PPP
## Value          -1438.666          0.320
##
## Latent Variables:
##          Estimate Post.SD pi.lower pi.upper Std.lv Std.all

```

```

## t1 =~
## prp_prs.p.w_15 1.000 13.600 0.510
## t2 =~
## prp_prs.p.w_17 1.000 38.648 0.860
## t3 =~
## prp_prs.p.w_19 1.000 48.069 0.903
## Rhat Prior
##
## NA
##
## NA
##
## NA
##
## Regressions:
## Estimate Post.SD pi.lower pi.upper Std.lv Std.all
## t2 ~
## t1 2.835 0.738 1.826 4.657 0.997 0.997
## t3 ~
## t2 1.242 0.131 1.009 1.511 0.998 0.998
## Rhat Prior
##
## 1.001 normal(0,10)
##
## 1.000 normal(0,10)
##
## Intercepts:
## Estimate Post.SD pi.lower pi.upper Std.lv Std.all
## .prp_prs.p.w_15 22.112 3.240 15.717 28.397 22.112 0.829
## .prp_prs.p.w_17 28.890 4.864 19.327 38.243 28.890 0.643
## .prp_prs.p.w_19 38.148 5.749 26.912 49.095 38.148 0.716
## t1 0.000 0.000 0.000
## .t2 0.000 0.000 0.000
## .t3 0.000 0.000 0.000
## Rhat Prior
## 0.999 normal(0,32)
## 1.000 normal(0,32)
## 1.000 normal(0,32)
## NA
## NA
## NA
##
## Variances:
## Estimate Post.SD pi.lower pi.upper Std.lv Std.all
## .prp_p._15 (a) 525.707 62.432 418.526 662.474 525.707 0.740
## .prp_p._17 (a) 525.707 62.432 418.526 662.474 525.707 0.260
## .prp_p._19 (a) 525.707 62.432 418.526 662.474 525.707 0.185
## t1 184.962 79.163 61.237 362.193 1.000 1.000
## .t2 7.493 15.628 0.002 53.688 0.005 0.005
## .t3 8.391 18.078 0.003 55.088 0.004 0.004
## Rhat Prior
## 1.001 gamma(1,.5)[sd]
## 1.001
## 1.001

```

```

##      1.000 gamma(1,.5)[sd]
##      1.000 gamma(1,.5)[sd]
##      1.000 gamma(1,.5)[sd]
##
## ** WARNING ** blavaan (0.3-15) did NOT converge after 500 adapt+burnin iterations
## ** WARNING ** Proceed with caution
##
##      Number of observations              73
##
##      Number of missing patterns          1
##
##      Statistic              MargLogLik      PPP
##      Value                  -2108.091      0.114
##
## Latent Variables:
##      Estimate Post.SD pi.lower pi.upper Std.lv Std.all
##      t1 =~
##      viol.p.w_15      1.000              31.439      0.913
##      t2 =~
##      viol.p.w_17      1.000              2.936      0.205
##      t3 =~
##      viol.p.w_19      1.000              29.720      0.904
##      Rhat      Prior
##
##      NA
##
##      NA
##
##      NA
##
## Regressions:
##      Estimate Post.SD pi.lower pi.upper Std.lv Std.all
##      t2 ~
##      t1              0.034      0.104      -0.149      0.155      0.362      0.362
##      t3 ~
##      t2              10.075      27.352      -35.542      38.951      0.995      0.995
##      Rhat      Prior
##
##      4.670      normal(0,10)
##
##      5.857      normal(0,10)
##
## Intercepts:
##      Estimate Post.SD pi.lower pi.upper Std.lv Std.all
##      .viol.p.w_15      14.103      3.848      6.64      21.66      14.103      0.410
##      .viol.p.w_17      9.799      1.711      6.564      13.143      9.799      0.684
##      .viol.p.w_19      27.637      12.711      2.704      52.702      27.637      0.841
##      t1              0.000              0.000      0.000
##      .t2              0.000              0.000      0.000
##      .t3              0.000              0.000      0.000
##      Rhat      Prior
##      1.000      normal(0,32)
##      1.010      normal(0,32)
##      1.000      normal(0,32)

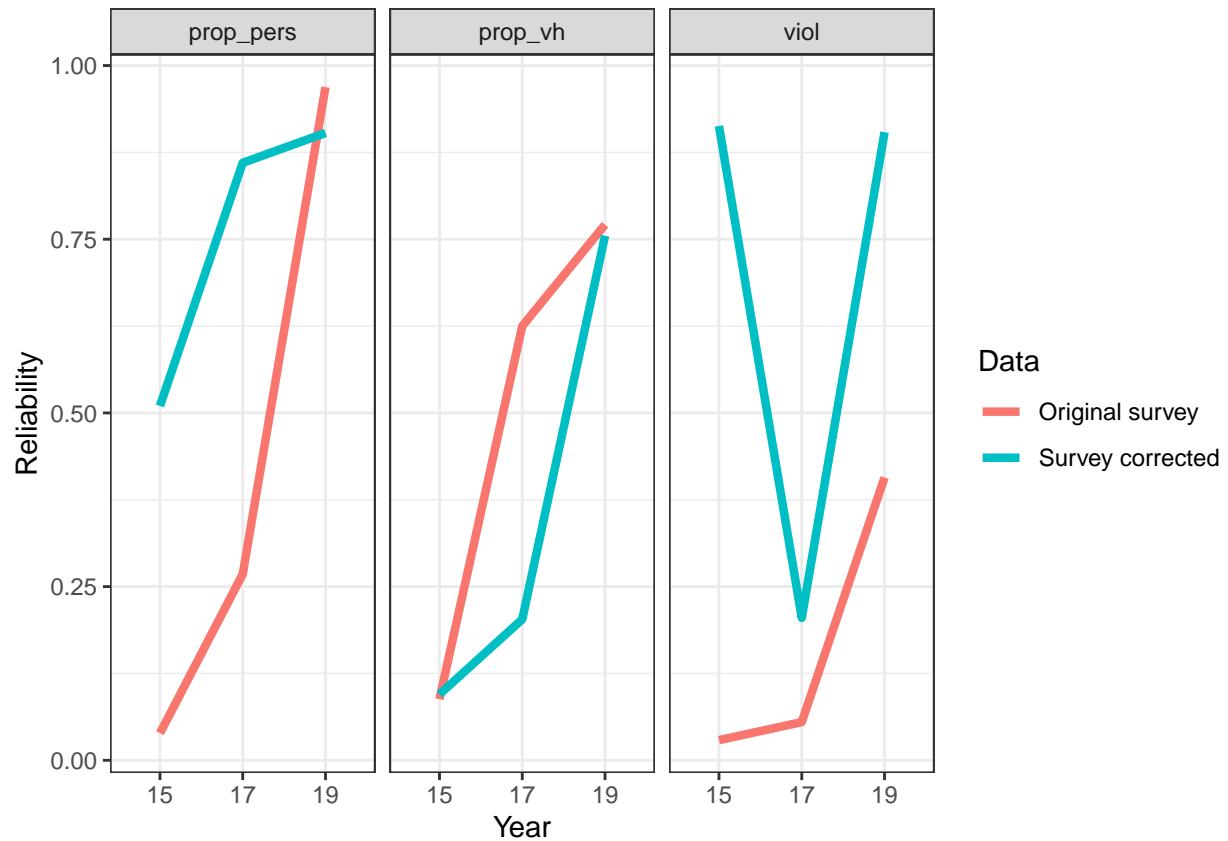
```



```

##      NA
##      NA
##      NA
##
## Variances:
##           Estimate Post.SD pi.lower pi.upper Std.lv Std.all
## .vil.p.w_15 (a)  196.510   36.077  140.673   281.73  196.510   0.166
## .vil.p.w_17 (a)  196.510   36.077  140.673   281.73  196.510   0.958
## .vil.p.w_19 (a)  196.510   36.077  140.673   281.73  196.510   0.182
##      t1          988.384  174.605  683.453 1360.856    1.000    1.000
##      t2           7.491    3.281    2.943   15.485    0.869    0.869
##      t3           8.303   19.913    0.006   51.333    0.009    0.009
##      Rhat      Prior
## 1.009 gamma(1,.5)[sd]
## 1.009
## 1.009
## 1.001 gamma(1,.5)[sd]
## 1.011 gamma(1,.5)[sd]
## 1.000 gamma(1,.5)[sd]

```



var	group	reliability	reliability_source
prop_pers	Original survey	0.43	0.36
prop_pers	Survey corrected	0.76	0.59
prop_vh	Original survey	0.49	0.36
prop_vh	Survey corrected	0.35	0.59
viol	Original survey	0.16	0.36

var	group	reliability	reliability_source
viol	Survey corrected	0.67	0.59

Longitudinal variance decomposition