

SOCY7706: Longitudinal Data Analysis
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Describing Longitudinal Data

Next, we examine some tools that allow us to describe change using longitudinal data. We will use an example from HRS data that focuses on employment and caregiving.

```
. use http://sarkisian.net/socy7706/hrs\_hours.dta

. reshape long r@workhours80 r@poorhealth r@married r@totalpar r@siblog h@childlg
r@allparhelptw, i(hhid pn) j(wave)
(note: j = 1 2 3 4 5 6 7 8 9)
```

Data	wide	->	long
Number of obs.	6591	->	59319
Number of variables	75	->	20
j variable (9 values)		->	wave
xij variables:			
rlworkhours80 r2workhours80 ... r9workhours80		->	rworkhours80
rlpoorhealth r2poorhealth ... r9poorhealth		->	rpoorhealth
r1married r2married ... r9married		->	rmarried
r1totalpar r2totalpar ... r9totalpar		->	rtotalpar
r1siblog r2siblog ... r9siblog		->	rsiblog
h1childlg h2childlg ... h9childlg		->	hchildlg
r1allparhelptw r2allparhelptw ... r9allparhelptw		->	rallparhelptw

```
. tab wave
```

wave	Freq.	Percent	Cum.
1	6,591	11.11	11.11
2	6,591	11.11	22.22
3	6,591	11.11	33.33
4	6,591	11.11	44.44
5	6,591	11.11	55.56
6	6,591	11.11	66.67
7	6,591	11.11	77.78
8	6,591	11.11	88.89
9	6,591	11.11	100.00
Total	59,319	100.00	

To keep things simpler for now, we will keep only two time points, but use preserve to return to the full data.

```
. preserve

. keep if wave<3
(46137 observations deleted)
```

Stata provides a number of tools for analyzing panel data. The commands all begin with the prefix `xt`. To use these commands, we first need to tell Stata that our dataset is a panel dataset. We need to have a variable that identifies the units (for example, a country or person id) and a time variable. To set a dataset as a panel, we need to use `xtset` command:

```
. xtset hhidpn wave
```

```

panel variable:  hhidpn (strongly balanced)
time variable:  wave, 1 to 2
delta: 1 unit

```

Stata thinks the dataset is strongly balanced, meaning all units are observed at all time points (at the same time and equal number of times). But it is not true – we just have empty rows that were created when we went from wide to long format.

```

. xtides

hhidpn: 10003020, 10004010, ..., 99564010      n =      6591
wave: 1, 2, ..., 2                             T =      2
Delta(wave) = 1 unit
Span(wave) = 2 periods
(hhidpn*wave uniquely identifies each observation)

Distribution of T_i:  min      5%      25%      50%      75%      95%      max
                   2         2         2         2         2         2         2

      Freq.  Percent  Cum. | Pattern
-----+-----
      6591   100.00  100.00 | 11
-----+-----
      6591   100.00      | XX

```

Xtides also thinks all cases are complete. We will now delete those empty records to have a more accurate picture. Note that those rows are not completely empty – time-invariant variables still have values there, but the time-variant ones are empty. So we will only specify time-varying variables in the egen command:

```

. egen miss=rowmiss( rworkhours80 rpoorhealth rmarried rtotalpar rsiblog hchildlg
rallparhelptw)

. tab miss
      miss |      Freq.      Percent      Cum.
-----+-----
          0 |    11,327      85.93      85.93
          1 |     1,017       7.72      93.64
          2 |         115       0.87      94.52
          3 |          90       0.68      95.20
          4 |           7       0.05      95.25
          5 |           3       0.02      95.27
          6 |          364       2.76      98.04
          7 |          259       1.96     100.00
-----+-----
      Total |    13,182     100.00

. drop if miss==7
(259 observations deleted)

. xtset  hhidpn wave
panel variable:  hhidpn (unbalanced)
time variable:  wave, 1 to 2
delta: 1 unit

```

This is more accurate now, and xtides also shows that there are missing observations at time 2.

```

. xtides

hhidpn: 10003020, 10004010, ..., 99564010      n =      6591
wave: 1, 2, ..., 2                             T =      2
Delta(wave) = 1 unit
Span(wave) = 2 periods
(hhidpn*wave uniquely identifies each observation)

Distribution of T_i:  min      5%      25%      50%      75%      95%      max
                   1        2        2        2        2        2        2
Freq.  Percent  Cum. | Pattern
-----+-----
6332   96.07   96.07 | 11
259    3.93   100.00 | 1.
-----+-----
6591  100.00           | XX

```

Next, let's examine change in a continuous variable.

```

. xtsum rworkhours80
Variable |      Mean  Std. Dev.      Min      Max | Observations
-----+-----
rwork~80 overall | 29.53971  22.79859      0      80 | N = 12477
          between |           21.33473      0      80 | n = 6580
          within  |           8.392351 -10.46029  69.53971 | T-bar = 1.8962

```

Here we see overall standard deviation along with between and within standard deviations – between indicates the amount of variation across individuals (cross-sectional variation, or differences among individuals that are stable over time), and within indicates change over time within individuals (temporal variation). Between variation is essentially variation of average values for individuals over time, and within variation is variation in differences between values at each time point and averages for a given individual (i.e. individual's deviation from their own overall mean). That is why the minimum and maximum differ from those for overall and between, and can be negative. Moreover, the way they calculate minimum and maximum is such that these are not just differences from the individual's mean, but such differences plus the overall mean (in this case, 29.5). So the person who has 69.5 (maximum value) in fact only differs from his or her own mean by $69.5 - 29.5 = 40$ hours. And the minimum value, -10.5, is in fact $-10.5 - 29.5 = -40$. So it is fairly symmetric, which is what we would expect. Observation column shows that there are 12477 records, 6580 individuals, and an average of 1.8962 time points per person.

So this output allows us to decompose the variance in the variable we are describing into variance components -- into within-group and between-group variance (although they are expressed as standard deviations – to get variances, we'd have to square them). That does not explain anything, but it allows us to evaluate whether there is variation in group means (here, person-specific means), and how much of it. That's why it is always a good idea to run this basic model when starting the analyses – it's the null model of our regression analysis. If we find that there is no significant variation across individuals, then there is no need to adjust for the fact that clusters of observations come from the same individuals because all individuals are pretty much the same.

The proportion of variance due to group-level variation in means can be calculated as

$$\rho = S^2_{\text{between}} / (S^2_{\text{between}} + S^2_{\text{within}})$$

It can be interpreted as the proportion of variance due to differences across individuals. It can also be interpreted as the average correlation between two randomly chosen time points that are in the same unit; therefore, it is also known as intra-class correlation. Here, we get:

```
. di 21.33473^2 / (21.33473^2 + 8.392351^2)
.86599838
```

So 87% of the total variance in hours of work is due to person-specific effects; the rest is due to changes that individuals experience over time.

To examine change in categorical variables, we can use both xttab and xttrans.

```
. xttab rmarried
```

rmarried	Overall		Between		Within
	Freq.	Percent	Freq.	Percent	Percent
0	2662	21.20	1532	23.24	92.13
1	9895	78.80	5300	80.41	97.73
Total	12557	100.00	6832	103.66	96.47

(n = 6591)

Here we can see that overall, out of all records in the data, 78.8% indicate that the person is currently married, and 21.2% indicate that the person is currently single. Between percent indicates that 80.41% of all individuals in the data were married at some point during the study (or in this case that means that they were married at either wave 1 or wave 2), and 23.24% of individuals were single at some point during the study period. The total is larger than 100 because any person who experienced both marriage and singlehood over this time period will be counted twice. Within percent indicates that among those individuals that were married at some point, they were married 97.73% of all of their data points, and among those who were single at some point, they were single 92.13% of all of their data points. The total for within is a weighted average – the number of people with at least one 0 multiplied by the proportion of 0s among these people’s records + the number of people with at least one 1 multiplied by the proportion of 1s among these people’s records, all divided by the total of those with at least one 1 and those with at least one 0. So here:

```
. di (1532*.9213+5300*.9773)/(1532+5300)
.96474262
```

```
. xttrans rmarried, freq
```

rmarried	rmarried		Total
	0	1	
0	1,130	87	1,217
	92.85	7.15	100.00
1	154	4,595	4,749
	3.24	96.76	100.00
Total	1,284	4,682	5,966
	21.52	78.48	100.00

Xttrans shows transitions among statuses: so here we see that among those who were married at time point 1, 96.76% were still married at time point 2, while 3.24% were no longer married. Of

those who were single at time 1, 92.85% were still single at time 2 and 7.15% were no longer single.

If we have more than 2 time points, xttab and xttrans put them all together. For example, let's go back to all 9 waves.

```
. restore
```

Let's once again get rid of those "empty" observations (with no data for a given wave):

```
. egen miss=rowmiss( rworkhours80 rpoorhealth rmarried rtotalpar rsiblog hchildlg
rallparhelptw)
```

```
. tab miss
```

miss	Freq.	Percent	Cum.
0	30,546	51.49	51.49
1	15,030	25.34	76.83
2	1,435	2.42	79.25
3	143	0.24	79.49
4	7	0.01	79.50
5	3	0.01	79.51
6	7,512	12.66	92.17
7	4,643	7.83	100.00
Total	59,319	100.00	

```
. drop if miss==7
```

```
(4643 observations deleted)
```

```
. xtset hhidpn wave
```

```
panel variable: hhidpn (unbalanced)
time variable: wave, 1 to 9, but with gaps
delta: 1 unit
```

Let's save this file:

```
. save hrs_hours_long.dta
```

And now we can describe the data:

```
. xtides
```

```
hhidpn: 10003020, 10004010, ..., 99564010      n =      6591
wave: 1, 2, ..., 9                             T =      9
Delta(wave) = 1 unit
Span(wave) = 9 periods
(hhidpn*wave uniquely identifies each observation)
```

```
Distribution of T_i:  min      5%      25%      50%      75%      95%      max
                   1         3         9         9         9         9         9
```

Freq.	Percent	Cum.	Pattern
5540	84.05	84.05	111111111
154	2.34	86.39	11.....
137	2.08	88.47	1.....
84	1.27	89.74	1111.....
81	1.23	90.97	11111....
73	1.11	92.08	11111111.
69	1.05	93.13	111.....

```

55      0.83   93.96 | 1111111..
49      0.74   94.70 | 1111111...
349     5.30  100.00 | (other patterns)
-----+-----
6591    100.00      | XXXXXXXXXX

```

```
. xtsum rworkhours80 rpoorhealth rmarried rtotalpar rsiblog hchildlg rallparhelptw
```

Variable	Mean	Std. Dev.	Min	Max	Observations
rwork~80 overall	19.67817	22.46339	0	80	N = 46661
between		17.10868	0	80	n = 6587
within		15.5263	-48.89326	90.78928	T-bar = 7.0838
rpoorh~h overall	.2340638	.4234167	0	1	N = 47141
between		.3401185	0	1	n = 6591
within		.2765936	-.6548251	1.122953	T-bar = 7.15233
rmarried overall	.7463865	.4350836	0	1	N = 47115
between		.3947182	0	1	n = 6591
within		.1958791	-.1425024	1.635275	T-bar = 7.14838
rtotal~r overall	.8884476	.8662165	0	4	N = 46830
between		.6585949	.1111111	4	n = 6591
within		.6045806	-2.111552	4.138448	T-bar = 7.10514
rsiblog overall	1.616775	.6256698	0	3.555348	N = 54595
between		.6154972	0	3.218876	n = 6588
within		.1870047	-.0368364	3.663518	T-bar = 8.28704
hchildlg overall	1.126299	.5481416	0	2.944439	N = 44219
between		.5389137	0	2.876082	n = 6272
within		.1319155	-.4992974	2.794222	T-bar = 7.05022
rallpa~w overall	1.652933	4.103339	0	19.23077	N = 32727
between		2.651108	0	19.23077	n = 6588
within		3.228803	-14.14377	18.74695	T-bar = 4.96767

```
. for var rpoorhealth rmarried : xttab X
```

```
-> xttab rpoorhealth
```

rpoorhe~h	Overall		Between		Within
	Freq.	Percent	Freq.	Percent	Percent
0	36107	76.59	5995	90.96	82.10
1	11034	23.41	3285	49.84	50.82
Total	47141	100.00	9280	140.80	71.02

(n = 6591)

```
-> xttab rmarried
```

rmarried	Overall		Between		Within
	Freq.	Percent	Freq.	Percent	Percent
0	11949	25.36	2404	36.47	70.58
1	35166	74.64	5455	82.76	89.72
Total	47115	100.00	7859	119.24	83.87

(n = 6591)

```
. for var rpoorhealth rmarried : xttrans X

-> xttrans rpoorhealth
rpoorhealth |      rpoorhealth
      h |      0      1 |      Total
-----+-----+-----
      0 |      89.35    10.65 |      100.00
      1 |      28.72    71.28 |      100.00
-----+-----+-----
    Total |      76.09    23.91 |      100.00

-> xttrans rmarried
      |      rmarried
rmarried |      0      1 |      Total
-----+-----+-----
      0 |      95.77     4.23 |      100.00
      1 |       3.49    96.51 |      100.00
-----+-----+-----
    Total |      26.10    73.90 |      100.00
```

It could also be helpful to continue looking at specific transitions, either taking two waves at a time or as sequences of transitions over a number of waves. The latter approach led to a technique called sequence analysis (or social sequence analysis when used in social sciences). If interested, see, for example, “Social Sequence Analysis” book by Benjamin Cornwall (2015). To see some basic sequences, let’s use a special package:

```
. net search sequence
(contacting http://www.stata.com)

33 packages found (Stata Journal and STB listed first)
-----

st0244 from http://www.stata-journal.com/software/sj12-1
SJ12-1 st0244. Scrambled Halton sequences in Mata / Scrambled Halton
sequences in Mata / by Stanislav Kolenikov, University of Missouri,
Columbia, USA / Support: skolenik@gmail.com

st0111 from http://www.stata-journal.com/software/sj6-4
SJ6-4 st0111. Sequence analysis with Stata / Sequence analysis with Stata
/ by Christian Brzinsky-Fay, Wissenschaftszentrum Berlin / Ulrich Kohler,
Wissenschaftszentrum Berlin / Magdalena Luniak, Wissenschaftszentrum
Berlin / Support: brzinsky-fay@wz-berlin.de, kohler@wz-berlin.de,

st0103 from http://www.stata-journal.com/software/sj6-2
SJ6-2 st0103. Generating Halton sequences using Mata / Generating Halton
sequences using Mata / by David Drukker, StataCorp / Richard Gates,
StataCorp / Support: rgates@stata.com / After installation, type help
sj_halton

dm55 from http://www.stata.com/stb/stb43
STB-43 dm55. Generating sequences and patterns of numeric data. / STB
insert by R. Mark Esman, Stata Corporation. / Support: mesman@stata.com /
After installation, see help fill.

dm44 from http://www.stata.com/stb/stb37
STB-37 dm44. Sequences of integers. / STB insert by Nicholas J. Cox,
University of Durham, UK. / Support: n.j.cox@durham.ac.uk / After
installation, see help seq.
```

```
sadi from http://teaching.sociology.ul.ie/sadi
Sequence Analysis Distance Measures / {bf: Brendan Halpin, Dept of
Sociology, Univer
--Break--
r(1);
```

We will install st0111 from <http://www.stata-journal.com/software/sj6-4>

```
. sqset rmarried hhidpn wave
```

```
Note: Some sequences contains gaps
Consider option -keeplongest-
```

```
Note: Some sequences have missings at the end
Consider option -rtrim-
```

```
element variable: rmarried, 0 to 1, and missings
identifier variable: hhidpn, 10003020 to 99564010
order variable: wave, 1 to 9
```

```
. sqtab
```

Sequence-Pa ttern	Freq.	Percent	Cum.
111111111	2,283	41.74	41.74
000000000	508	9.29	51.03
111	209	3.82	54.85
1	206	3.77	58.62
11	194	3.55	62.17
11111	165	3.02	65.19
1111	157	2.87	68.06
11111111	155	2.83	70.89
111111	120	2.19	73.08
1111111	117	2.14	75.22
111111110	93	1.70	76.92
00	75	1.37	78.30
0	73	1.33	79.63
111000000	70	1.28	80.91
111110000	62	1.13	82.04
111111100	59	1.08	83.12
111111000	57	1.04	84.17
110000000	54	0.99	85.15
0000	52	0.95	86.10
00000	50	0.91	87.02
111100000	49	0.90	87.91
100000000	48	0.88	88.79
00000000	45	0.82	89.61
000	41	0.75	90.36
0000000	39	0.71	91.08
011111111	36	0.66	91.74
000000	24	0.44	92.17
000011111	22	0.40	92.58
10	11	0.20	92.78
110	11	0.20	92.98
11111110	11	0.20	93.18
...[output omitted]...			
111110110	1	0.02	99.98
11111101	1	0.02	100.00
Total	5,469	100.00	

While all of these tools are helpful to better understand the nature of change in your data, such analyses rarely directly appear in articles using longitudinal data – typically, tables of descriptive statistics take one of the following approaches, or some combination:

1. Show means (and standard deviations) for each year, or some select years (e.g., the first and last year of the time sequence), or select ages – this approach aims to show how various variables in the study changed over time, but does not target change within individuals.
2. Show means and standard deviations for pooled data (entire long dataset, where units are person-years or country-years, etc.). This does not focus on change over time (either overall or individual) but rather aims at describing the dataset overall.
3. Show averages or frequencies of changes, transitions, or trajectories of individuals.

Examples:

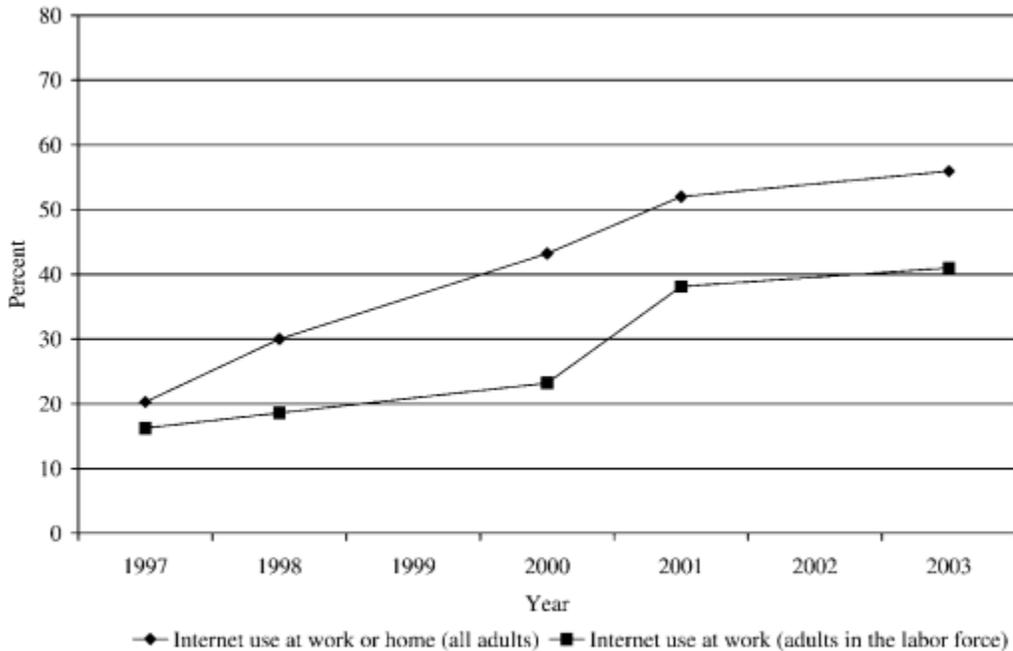


Figure 1. Internet Use, 1997 to 2003

Source: Current Population Survey Internet and Computer Use Supplements, 1997, 2000, 2001, and 2003.

From: DiMaggio, Paul, and Bart Bonikowski. 2008. "Make Money Surfing the Web? The Impact of Internet Use on the Earnings of U.S. Workers." *American Sociological Review* 73: 227–250.

Table 1. Descriptive Statistics for Dependent and Explanatory Measures, 1970 to 2010

Measures	Overall		1970		1980		1990		2000		2010	
	Mean	(SD)	Mean	(SD)	Mean	(SD)	Mean	(SD)	Mean	(SD)	Mean	(SD)
White Homicide Rate	4.3	(2.7)	4.4	(2.8)	6.1	(3.5)	4.6	(2.2)	3.3	(1.5)	2.9	(1.4)
Black Homicide Rate	31.1	(15.8)	42.2	(16.0)	35.2	(14.1)	36.0	(15.3)	21.5	(9.9)	20.4	(10.0)
B-W Index of Dissimilarity	64.5	(12.5)	75.8	(9.5)	68.0	(10.1)	63.5	(10.9)	59.6	(10.8)	55.4	(10.4)
<i>Black Measures</i>												
Poverty ^a	27.6	(7.5)	30.1	(8.8)	26.8	(5.1)	28.8	(7.9)	24.6	(6.3)	27.7	(7.5)
Unemployment ^a	12.2	(5.0)	7.6	(2.9)	11.8	(4.1)	13.0	(4.1)	11.2	(2.7)	17.3	(5.1)
Single-Parent Families ^a	53.5	(13.1)	37.6	(6.3)	49.0	(7.7)	57.1	(8.6)	60.3	(9.7)	63.7	(12.6)
High Income	9.6	(5.7)	3.8	(2.6)	7.6	(3.0)	10.1	(4.5)	13.7	(5.0)	12.7	(6.0)
Manufacturing	16.5	(9.8)	22.9	(10.7)	22.6	(10.4)	15.6	(7.4)	12.7	(5.9)	8.8	(4.4)
Residential Instability	25.2	(8.3)	14.6	(4.6)	27.7	(7.2)	28.4	(6.7)	26.5	(5.2)	28.6	(7.5)
Young Men	9.2	(2.5)	9.5	(2.9)	11.1	(3.5)	8.7	(1.7)	8.1	(1.1)	8.8	(1.6)
Foreign-Born Pop.	3.6	(5.6)	0.5	(0.8)	2.1	(2.6)	3.1	(4.4)	4.6	(6.2)	7.5	(7.9)
<i>White Measures</i>												
Poverty ^b	8.0	(2.3)	8.3	(2.2)	7.4	(1.5)	7.5	(2.0)	7.1	(1.8)	9.8	(2.8)
Unemployment ^b	5.3	(2.5)	3.7	(1.3)	5.1	(1.7)	4.6	(1.1)	4.2	(0.9)	9.0	(2.5)
Single-Parent Families ^b	17.7	(5.5)	10.9	(2.2)	15.0	(2.3)	17.6	(2.6)	20.7	(3.4)	24.1	(4.5)
High Income	27.1	(9.6)	16.2	(4.1)	23.5	(5.2)	29.0	(7.6)	34.2	(7.9)	32.8	(8.7)
Manufacturing	16.8	(8.6)	23.8	(9.7)	20.8	(8.4)	16.3	(6.3)	13.1	(5.5)	10.3	(4.2)
Residential Instability	19.0	(5.7)	12.6	(3.5)	23.7	(5.7)	20.5	(4.9)	18.6	(3.8)	19.3	(3.9)
Young Men	7.6	(1.7)	8.9	(2.0)	9.3	(1.1)	7.0	(0.9)	6.3	(0.8)	6.4	(0.8)
Foreign-Born Pop.	3.2	(2.6)	1.8	(1.3)	3.7	(2.7)	3.1	(2.5)	3.4	(2.7)	3.9	(3.1)
<i>Overall Measures</i>												
Percent Black	12.7	(9.7)	10.8	(8.7)	12.0	(9.4)	12.5	(9.6)	13.7	(10.2)	14.5	(10.3)
Incarceration Rate	283.4	(190.0)	93.7	(31.4)	138.9	(62.0)	294.1	(115.1)	446.1	(176.5)	444.4	(149.0)
Police per Capita	194.3	(74.3)	157.4	(47.8)	182.6	(63.2)	192.3	(74.1)	218.9	(80.3)	220.2	(82.3)
N	515	515	103	103	103	103	103	103	103	103	103	103

Note: Results are unweighted.

^aMeasures comprise the Black Disadvantage Index.

^bMeasures comprise the White Disadvantage Index.

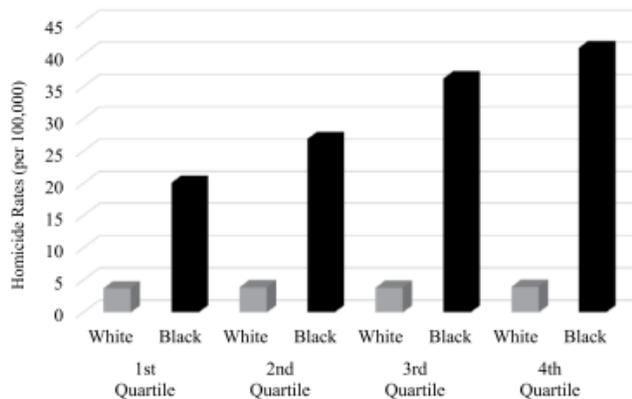


Figure 1. Black and White Homicide Rates across Different Levels of Segregation, 1970 to 2010

Note: The analysis includes 103 MSAs. Homicide rates are weighted by the size of the respective populations. The dissimilarity range for each quartile is as follows: 1st (29.1 to 56.3); 2nd (56.3 to 65.1); 3rd (65.2 to 73.6); and 4th (73.8 to 91.5).

From: Light, Michael T. and Julia T. Thomas. 2019. "Segregation and Violence Reconsidered: Do Whites Benefit from Residential Segregation?" *American Sociological Review*, 84(4): 690–725.

Table 1. Health Problems by Gender at Age 40 and 50 among All Individuals and among Married Individuals

	Age 40				Age 50				Range
	Total		Married		Total		Married		
	Mean	SD	Mean	SD	Mean	SD	Mean	SD	
Women									
Chronic Conditions	.56	.88	.51	.81	1.17	1.28	1.02	1.16	[0, 7]
Self-Rated Health	2.32	1.01	2.20	.97	2.58	1.06	2.43	.99	[1, 5]
Physical Functioning	48.32	8.76	47.55	7.88	51.29	10.88	50.04	9.92	[33.4, 88.8]
Men									
Chronic Conditions	.43	.73	.39	.68	.98	1.13	.89	1.07	[0, 7]
Self-Rated Health	2.24	.96	2.18	.93	2.49	1.03	2.35	.96	[1, 5]
Physical Functioning	47.00	6.73	46.83	6.57	49.57	9.00	48.67	8.01	[32.8, 88.4]

Note: Range represents observed range across both waves. Self-rated health and physical functioning are reverse coded so that higher values indicate worse health for all variables in the analysis. The theoretical range for physical function is 1-excellent to 100-poor.

Table 3. Individual-Level Descriptive Statistics (For Person-Years in Full Sample and Married Subsample)

	Total		Married		Range
	Mean	SD	Mean	SD	
Age (years)	45.8	4.8	45.8	4.8	[39, 55]
Household income (\$k)	75.1	74.9	95.2	80.5	[0, 498]
Education (years)	13.5	2.5	13.7	2.6	[0, 20]
Duration in state of residence (years)	9.1	2.4	9.1	2.3	[0, 10]
Men	46.3%		46.9%		
Non-white	19.3%		12.8%		
Married	64.1%		100%		
Have children	80.8%		88.7%		
Have health insurance	84.9%		91.0%		
Divorced during period	10.1%		7.9%		
N (person-years)	6,754		4,336		

From: Homan, Patricia. 2019. "Structural Sexism and Health in the United States: A New Perspective on Health Inequality and the Gender System." *American Sociological Review*, 84(3): 486–516.

rsiblog					
1		1.673055	.0081439	1.657092	1.689017
2		1.680897	.0086198	1.664002	1.697792
3		1.691382	.0091705	1.673407	1.709356
4		1.666379	.0096965	1.647374	1.685385
5		1.678038	.0106761	1.657112	1.698963
6		1.680657	.0116672	1.657788	1.703525
7		1.678899	.0131027	1.653217	1.704581
8		1.656552	.014932	1.627284	1.685819
9		1.652234	.0177086	1.617525	1.686944

hchildlg					
1		1.107862	.0070198	1.094103	1.121621
2		1.118057	.0074358	1.103482	1.132631
3		1.109619	.0081834	1.093579	1.125659
4		1.124737	.0087356	1.107614	1.141859
5		1.131209	.0096739	1.112248	1.15017
6		1.129443	.0106084	1.10865	1.150236
7		1.140981	.0119127	1.117632	1.16433
8		1.136373	.0136879	1.109544	1.163202
9		1.13422	.0158981	1.103059	1.165381

rallparhelptw					
1		.6292256	.0373714	.555976	.7024751
2		1.228155	.0464493	1.137113	1.319198
3		1.584609	.0566464	1.473579	1.695638
4		1.896702	.0668625	1.765648	2.027755
5		1.763599	.0711459	1.62415	1.903048
6		2.296546	.0912181	2.117755	2.475338
7		2.563283	.1127419	2.342304	2.784261
8		2.68663	.1313707	2.429138	2.944122
9		2.63579	.1508951	2.340029	2.93155

Means of time-invariant variables:

. mean raedyrs female age white black latino otherrace minority if wave==1

Mean estimation		Number of obs = 6,588		
	Mean	Std. Err.	[95% Conf. Interval]	
raedyrs	12.27277	.0389866	12.19634	12.3492
female	.4845173	.0061577	.4724462	.4965884
age	55.45835	.0381301	55.3836	55.53309
white	.7328476	.0054518	.7221602	.743535
black	.153309	.0044392	.1446068	.1620113
latino	.0910747	.003545	.0841253	.0980241
otherrace	.0227687	.0018379	.0191658	.0263716
minority	.2671524	.0054518	.256465	.2778398

A graph of means of work hours over time by gender:

. reg rworkhours80 i.female##i.wave

Source	SS	df	MS	Number of obs	=	46,661
Model	4111549.35	17	241855.844	F(17, 46643)	=	580.49
Residual	19433266.7	46,643	416.638439	Prob > F	=	0.0000
				R-squared	=	0.1746
				Adj R-squared	=	0.1743


```

      wave          =          1
10._at : female    =          1
      wave          =          2
11._at : female    =          1
      wave          =          3
12._at : female    =          1
      wave          =          4
13._at : female    =          1
      wave          =          5
14._at : female    =          1
      wave          =          6
15._at : female    =          1
      wave          =          7
16._at : female    =          1
      wave          =          8

```

```

-----
      |              Delta-method
      |      Margin   Std. Err.   t   P>|t|   [95% Conf. Interval]
-----+-----
   _at |
   1  |   37.42107   .3516128  106.43  0.000   36.7319   38.11023
   2  |   34.52337   .371614   92.90  0.000   33.795   35.25174
   3  |   31.13619   .3824137   81.42  0.000   30.38665  31.88572
   4  |   27.1968    .3936997   69.08  0.000   26.42514  27.96846
   5  |   22.76769   .4046887   56.26  0.000   21.97449  23.56088
   6  |   17.76257   .4126318   43.05  0.000   16.9538   18.57133
   7  |   14.43839   .4244161   34.02  0.000   13.60653  15.27025
   8  |   11.48436   .4377739   26.23  0.000   10.62632  12.3424
   9  |   23.64286   .3620785   65.30  0.000   22.93318  24.35254
  10  |   21.69093   .3782544   57.34  0.000   20.94955  22.43232
  11  |   18.93071   .3877586   48.82  0.000   18.1707   19.69072
  12  |   16.90015   .396962    42.57  0.000   16.1221   17.6782
  13  |   14.24841   .4072582   34.99  0.000   13.45018  15.04664
  14  |   10.79761   .4139875   26.08  0.000   9.986193  11.60904
  15  |   8.890971   .421241    21.11  0.000   8.065332  9.71661
  16  |   6.733036   .4312764   15.61  0.000   5.887728  7.578344
-----

```

```

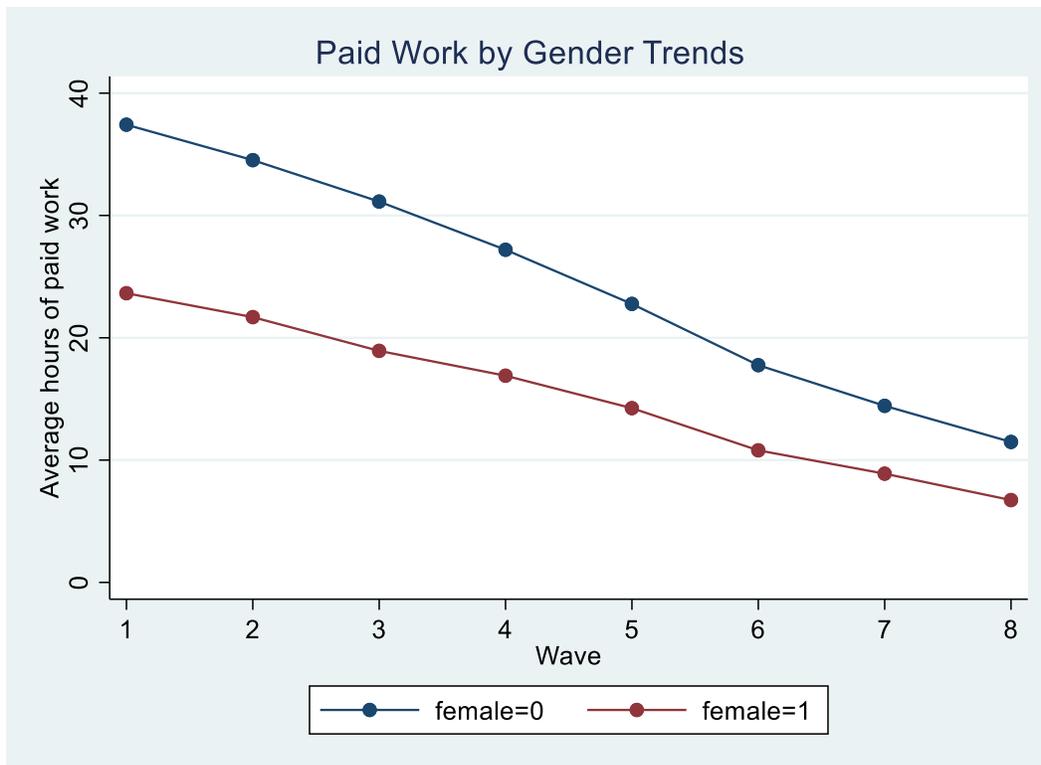
. marginsplot, noci ytitle("Average hours of paid work") xtitle("Wave") title("Paid
Work by Gender Trends")

```

```

Variables that uniquely identify margins: wave female

```



Examining the nature of changes wave to wave:

```
. gen diff_hours=d.rworkhours80
(16,141 missing values generated)
```

```
. mean diff_hours, over(wave)
Mean estimation      Number of obs   =   38,535
```

- 2: wave = 2
- 3: wave = 3
- 4: wave = 4
- 5: wave = 5
- 6: wave = 6
- 7: wave = 7
- 8: wave = 8
- 9: wave = 9

Over	Mean	Std. Err.	[95% Conf. Interval]	
diff_hours				
2	-2.818721	.2218065	-3.253468	-2.383975
3	-3.307249	.2321794	-3.762327	-2.852172
4	-3.213501	.2324172	-3.669044	-2.757957
5	-3.753209	.2310037	-4.205982	-3.300436
6	-4.402008	.2398089	-4.872039	-3.931976
7	-2.631661	.2151473	-3.053355	-2.209967
8	-2.50271	.1999011	-2.894522	-2.110899
9	-1.550472	.1853217	-1.913708	-1.187237

```
. gen diff_hours_cat=(diff_hours>0) if diff_hours<.
(16,141 missing values generated)
```

```
. replace diff_hours_cat=-1 if diff_hours<0
(9,729 real changes made)
```

```
. tab diff_hours_cat, m
```

diff_hours_cat	Freq.	Percent	Cum.
-1	9,729	17.79	17.79
0	22,835	41.76	59.56
1	5,971	10.92	70.48
.	16,141	29.52	100.00
Total	54,676	100.00	

```
. tab diff_hours_cat wave, col
```

```
+-----+
| Key   |
+-----+
|       |
| frequency |
| column percentage |
+-----+
```

diff_hours_cat	wave						Total
	2	3	4	5	6	7	
-1	1,704 28.90	1,622 30.15	1,421 27.88	1,320 27.33	1,245 27.17	946 21.09	9,729 25.25
0	2,922 49.55	2,739 50.91	2,741 53.79	2,782 57.60	2,752 60.06	2,968 66.18	22,835 59.26
1	1,271 21.55	1,019 18.94	934 18.33	728 15.07	585 12.77	571 12.73	5,971 15.50
Total	5,897 100.00	5,380 100.00	5,096 100.00	4,830 100.00	4,582 100.00	4,485 100.00	38,535 100.00

diff_hours_cat	wave		Total
	8	9	
-1	823 19.40	648 16.11	9,729 25.25
0	2,969 69.97	2,962 73.64	22,835 59.26
1	451 10.63	412 10.24	5,971 15.50
Total	4,243 100.00	4,022 100.00	38,535 100.00