

Chapter 10

Miscellaneous examples and data tricks

Reading: Hadley Wickham, “Tidy Data” (Journal of Statistical Software, 2014):

<http://vita.had.co.nz/papers/tidy-data.pdf>

Warning: This chapter is pretty random and rough.

10.1 Tuberculosis

Data: <https://github.com/hadley/tidy-data/blob/master/data/tb.csv>

Let’s explore tuberculosis in the U.S. How do male and female rates compare?

```
library(plyr)
tb = read.csv("https://github.com/hadley/tidy-data/raw/master/data/tb.csv",
  stringsAsFactors = FALSE)
tb.US = subset(tb, iso2 == "US")
# tb.US = as_tibble(tb.US)
```

Gather up the counts into one column:

```
library(dplyr)
library(tidyr)
tb.long = tb.US %>% gather(demographic, count, -iso2, -year, na.rm = TRUE)
```

Separate sex and age using `separate()`:

```
tb.sep = tb.long %>% separate(demographic, c("sex", "age"), sep = 8)
# Get rid of totals
tb.sep = subset(tb.sep, sex != "new_sp")
tb.sep
```

##	iso2	year	sex	age	count
## 56	US	2006	new_sp_m	04	4
## 57	US	2007	new_sp_m	04	4
## 58	US	2008	new_sp_m	04	4
## 85	US	2006	new_sp_m	514	8
## 86	US	2007	new_sp_m	514	8
## 87	US	2008	new_sp_m	514	7

```

## 103  US 1995 new_sp_m 014    19
## 104  US 1996 new_sp_m 014    15
## 105  US 1997 new_sp_m 014    12
## 106  US 1998 new_sp_m 014    10
## 107  US 1999 new_sp_m 014    18
## 108  US 2000 new_sp_m 014     6
## 109  US 2001 new_sp_m 014    17
## 110  US 2002 new_sp_m 014    14
## 111  US 2003 new_sp_m 014    11
## 112  US 2004 new_sp_m 014    12
## 113  US 2005 new_sp_m 014    14
## 114  US 2006 new_sp_m 014    12
## 115  US 2007 new_sp_m 014    12
## 116  US 2008 new_sp_m 014    11
## 132  US 1995 new_sp_m 1524   355
## 133  US 1996 new_sp_m 1524   333
## 134  US 1997 new_sp_m 1524   330
## 135  US 1998 new_sp_m 1524   321
## 136  US 1999 new_sp_m 1524   331
## 137  US 2000 new_sp_m 1524   365
## 138  US 2001 new_sp_m 1524   320
## 139  US 2002 new_sp_m 1524   343
## 140  US 2003 new_sp_m 1524   365
## 141  US 2004 new_sp_m 1524   362
## 142  US 2005 new_sp_m 1524   383
## 143  US 2006 new_sp_m 1524   388
## 144  US 2007 new_sp_m 1524   414
## 145  US 2008 new_sp_m 1524   375
## 161  US 1995 new_sp_m 2534   876
## 162  US 1996 new_sp_m 2534   815
## 163  US 1997 new_sp_m 2534   701
## 164  US 1998 new_sp_m 2534   663
## 165  US 1999 new_sp_m 2534   616
## 166  US 2000 new_sp_m 2534   602
## 167  US 2001 new_sp_m 2534   613
## 168  US 2002 new_sp_m 2534   562
## 169  US 2003 new_sp_m 2534   526
## 170  US 2004 new_sp_m 2534   547
## 171  US 2005 new_sp_m 2534   535
## 172  US 2006 new_sp_m 2534   568
## 173  US 2007 new_sp_m 2534   490
## 174  US 2008 new_sp_m 2534   513
## 190  US 1995 new_sp_m 3544  1417
## 191  US 1996 new_sp_m 3544  1219
## 192  US 1997 new_sp_m 3544  1127
## 193  US 1998 new_sp_m 3544  1009
## 194  US 1999 new_sp_m 3544  1011
## 195  US 2000 new_sp_m 3544   906
## 196  US 2001 new_sp_m 3544   824
## 197  US 2002 new_sp_m 3544   813
## 198  US 2003 new_sp_m 3544   754
## 199  US 2004 new_sp_m 3544   728
## 200  US 2005 new_sp_m 3544   666
## 201  US 2006 new_sp_m 3544   659

```

```

## 202    US 2007 new_sp_m 3544    572
## 203    US 2008 new_sp_m 3544    495
## 219    US 1995 new_sp_m 4554   1121
## 220    US 1996 new_sp_m 4554   1073
## 221    US 1997 new_sp_m 4554    979
## 222    US 1998 new_sp_m 4554   1007
## 223    US 1999 new_sp_m 4554    930
## 224    US 2000 new_sp_m 4554    904
## 225    US 2001 new_sp_m 4554    876
## 226    US 2002 new_sp_m 4554    795
## 227    US 2003 new_sp_m 4554    828
## 228    US 2004 new_sp_m 4554    829
## 229    US 2005 new_sp_m 4554    767
## 230    US 2006 new_sp_m 4554    759
## 231    US 2007 new_sp_m 4554    744
## 232    US 2008 new_sp_m 4554    725
## 248    US 1995 new_sp_m 5564    742
## 249    US 1996 new_sp_m 5564    678
## 250    US 1997 new_sp_m 5564    679
## 251    US 1998 new_sp_m 5564    628
## 252    US 1999 new_sp_m 5564    601
## 253    US 2000 new_sp_m 5564    577
## 254    US 2001 new_sp_m 5564    524
## 255    US 2002 new_sp_m 5564    490
## 256    US 2003 new_sp_m 5564    487
## 257    US 2004 new_sp_m 5564    504
## 258    US 2005 new_sp_m 5564    499
## 259    US 2006 new_sp_m 5564    531
## 260    US 2007 new_sp_m 5564    533
## 261    US 2008 new_sp_m 5564    526
## 277    US 1995 new_sp_m    65  1099
## 278    US 1996 new_sp_m    65  1007
## 279    US 1997 new_sp_m    65   944
## 280    US 1998 new_sp_m    65   914
## 281    US 1999 new_sp_m    65   801
## 282    US 2000 new_sp_m    65   738
## 283    US 2001 new_sp_m    65   649
## 284    US 2002 new_sp_m    65   592
## 285    US 2003 new_sp_m    65   650
## 286    US 2004 new_sp_m    65   582
## 287    US 2005 new_sp_m    65   624
## 288    US 2006 new_sp_m    65   596
## 289    US 2007 new_sp_m    65   562
## 290    US 2008 new_sp_m    65   561
## 319    US 2008 new_sp_m    u     0
## 346    US 2006 new_sp_f    04     2
## 347    US 2007 new_sp_f    04     2
## 348    US 2008 new_sp_f    04     4
## 375    US 2006 new_sp_f   514     9
## 376    US 2007 new_sp_f   514    10
## 377    US 2008 new_sp_f   514    18
## 393    US 1995 new_sp_f    014    26
## 394    US 1996 new_sp_f    014    21
## 395    US 1997 new_sp_f    014    28

```

```

## 396  US 1998 new_sp_f 014 15
## 397  US 1999 new_sp_f 014 16
## 398  US 2000 new_sp_f 014 14
## 399  US 2001 new_sp_f 014 21
## 400  US 2002 new_sp_f 014 15
## 401  US 2003 new_sp_f 014 12
## 402  US 2004 new_sp_f 014 19
## 403  US 2005 new_sp_f 014 11
## 404  US 2006 new_sp_f 014 11
## 405  US 2007 new_sp_f 014 12
## 406  US 2008 new_sp_f 014 22
## 422  US 1995 new_sp_f 1524 280
## 423  US 1996 new_sp_f 1524 289
## 424  US 1997 new_sp_f 1524 269
## 425  US 1998 new_sp_f 1524 269
## 426  US 1999 new_sp_f 1524 232
## 427  US 2000 new_sp_f 1524 246
## 428  US 2001 new_sp_f 1524 239
## 429  US 2002 new_sp_f 1524 233
## 430  US 2003 new_sp_f 1524 277
## 431  US 2004 new_sp_f 1524 265
## 432  US 2005 new_sp_f 1524 241
## 433  US 2006 new_sp_f 1524 257
## 434  US 2007 new_sp_f 1524 257
## 435  US 2008 new_sp_f 1524 220
## 451  US 1995 new_sp_f 2534 579
## 452  US 1996 new_sp_f 2534 487
## 453  US 1997 new_sp_f 2534 449
## 454  US 1998 new_sp_f 2534 425
## 455  US 1999 new_sp_f 2534 391
## 456  US 2000 new_sp_f 2534 376
## 457  US 2001 new_sp_f 2534 410
## 458  US 2002 new_sp_f 2534 423
## 459  US 2003 new_sp_f 2534 353
## 460  US 2004 new_sp_f 2534 339
## 461  US 2005 new_sp_f 2534 348
## 462  US 2006 new_sp_f 2534 384
## 463  US 2007 new_sp_f 2534 338
## 464  US 2008 new_sp_f 2534 329
## 480  US 1995 new_sp_f 3544 499
## 481  US 1996 new_sp_f 3544 478
## 482  US 1997 new_sp_f 3544 447
## 483  US 1998 new_sp_f 3544 424
## 484  US 1999 new_sp_f 3544 394
## 485  US 2000 new_sp_f 3544 349
## 486  US 2001 new_sp_f 3544 346
## 487  US 2002 new_sp_f 3544 362
## 488  US 2003 new_sp_f 3544 310
## 489  US 2004 new_sp_f 3544 302
## 490  US 2005 new_sp_f 3544 276
## 491  US 2006 new_sp_f 3544 263
## 492  US 2007 new_sp_f 3544 260
## 493  US 2008 new_sp_f 3544 269
## 509  US 1995 new_sp_f 4554 285

```

```
## 510    US 1996 new_sp_f 4554    279
## 511    US 1997 new_sp_f 4554    254
## 512    US 1998 new_sp_f 4554    267
## 513    US 1999 new_sp_f 4554    245
## 514    US 2000 new_sp_f 4554    253
## 515    US 2001 new_sp_f 4554    247
## 516    US 2002 new_sp_f 4554    255
## 517    US 2003 new_sp_f 4554    269
## 518    US 2004 new_sp_f 4554    252
## 519    US 2005 new_sp_f 4554    242
## 520    US 2006 new_sp_f 4554    212
## 521    US 2007 new_sp_f 4554    225
## 522    US 2008 new_sp_f 4554    224
## 538    US 1995 new_sp_f 5564    202
## 539    US 1996 new_sp_f 5564    217
## 540    US 1997 new_sp_f 5564    201
## 541    US 1998 new_sp_f 5564    179
## 542    US 1999 new_sp_f 5564    244
## 543    US 2000 new_sp_f 5564    152
## 544    US 2001 new_sp_f 5564    176
## 545    US 2002 new_sp_f 5564    167
## 546    US 2003 new_sp_f 5564    169
## 547    US 2004 new_sp_f 5564    166
## 548    US 2005 new_sp_f 5564    161
## 549    US 2006 new_sp_f 5564    146
## 550    US 2007 new_sp_f 5564    135
## 551    US 2008 new_sp_f 5564    172
## 567    US 1995 new_sp_f    65    591
## 568    US 1996 new_sp_f    65    541
## 569    US 1997 new_sp_f    65    514
## 570    US 1998 new_sp_f    65    492
## 571    US 1999 new_sp_f    65    444
## 572    US 2000 new_sp_f    65    396
## 573    US 2001 new_sp_f    65    389
## 574    US 2002 new_sp_f    65    370
## 575    US 2003 new_sp_f    65    354
## 576    US 2004 new_sp_f    65    344
## 577    US 2005 new_sp_f    65    322
## 578    US 2006 new_sp_f    65    303
## 579    US 2007 new_sp_f    65    308
## 580    US 2008 new_sp_f    65    300
## 609    US 2008 new_sp_f    u      0
```

```
# Rename categories
tb.sep$sex[tb.sep$sex == "new_sp_m"] = "male"
tb.sep$sex[tb.sep$sex == "new_sp_f"] = "female"
tb.sep$sex = factor(tb.sep$sex, levels = c("male", "female"))
tb.sep
```

```
##      iso2 year    sex  age count
## 56    US 2006   male   04     4
## 57    US 2007   male   04     4
## 58    US 2008   male   04     4
## 85    US 2006   male  514     8
## 86    US 2007   male  514     8
```

## 87	US	2008	male	514	7
## 103	US	1995	male	014	19
## 104	US	1996	male	014	15
## 105	US	1997	male	014	12
## 106	US	1998	male	014	10
## 107	US	1999	male	014	18
## 108	US	2000	male	014	6
## 109	US	2001	male	014	17
## 110	US	2002	male	014	14
## 111	US	2003	male	014	11
## 112	US	2004	male	014	12
## 113	US	2005	male	014	14
## 114	US	2006	male	014	12
## 115	US	2007	male	014	12
## 116	US	2008	male	014	11
## 132	US	1995	male	1524	355
## 133	US	1996	male	1524	333
## 134	US	1997	male	1524	330
## 135	US	1998	male	1524	321
## 136	US	1999	male	1524	331
## 137	US	2000	male	1524	365
## 138	US	2001	male	1524	320
## 139	US	2002	male	1524	343
## 140	US	2003	male	1524	365
## 141	US	2004	male	1524	362
## 142	US	2005	male	1524	383
## 143	US	2006	male	1524	388
## 144	US	2007	male	1524	414
## 145	US	2008	male	1524	375
## 161	US	1995	male	2534	876
## 162	US	1996	male	2534	815
## 163	US	1997	male	2534	701
## 164	US	1998	male	2534	663
## 165	US	1999	male	2534	616
## 166	US	2000	male	2534	602
## 167	US	2001	male	2534	613
## 168	US	2002	male	2534	562
## 169	US	2003	male	2534	526
## 170	US	2004	male	2534	547
## 171	US	2005	male	2534	535
## 172	US	2006	male	2534	568
## 173	US	2007	male	2534	490
## 174	US	2008	male	2534	513
## 190	US	1995	male	3544	1417
## 191	US	1996	male	3544	1219
## 192	US	1997	male	3544	1127
## 193	US	1998	male	3544	1009
## 194	US	1999	male	3544	1011
## 195	US	2000	male	3544	906
## 196	US	2001	male	3544	824
## 197	US	2002	male	3544	813
## 198	US	2003	male	3544	754
## 199	US	2004	male	3544	728
## 200	US	2005	male	3544	666

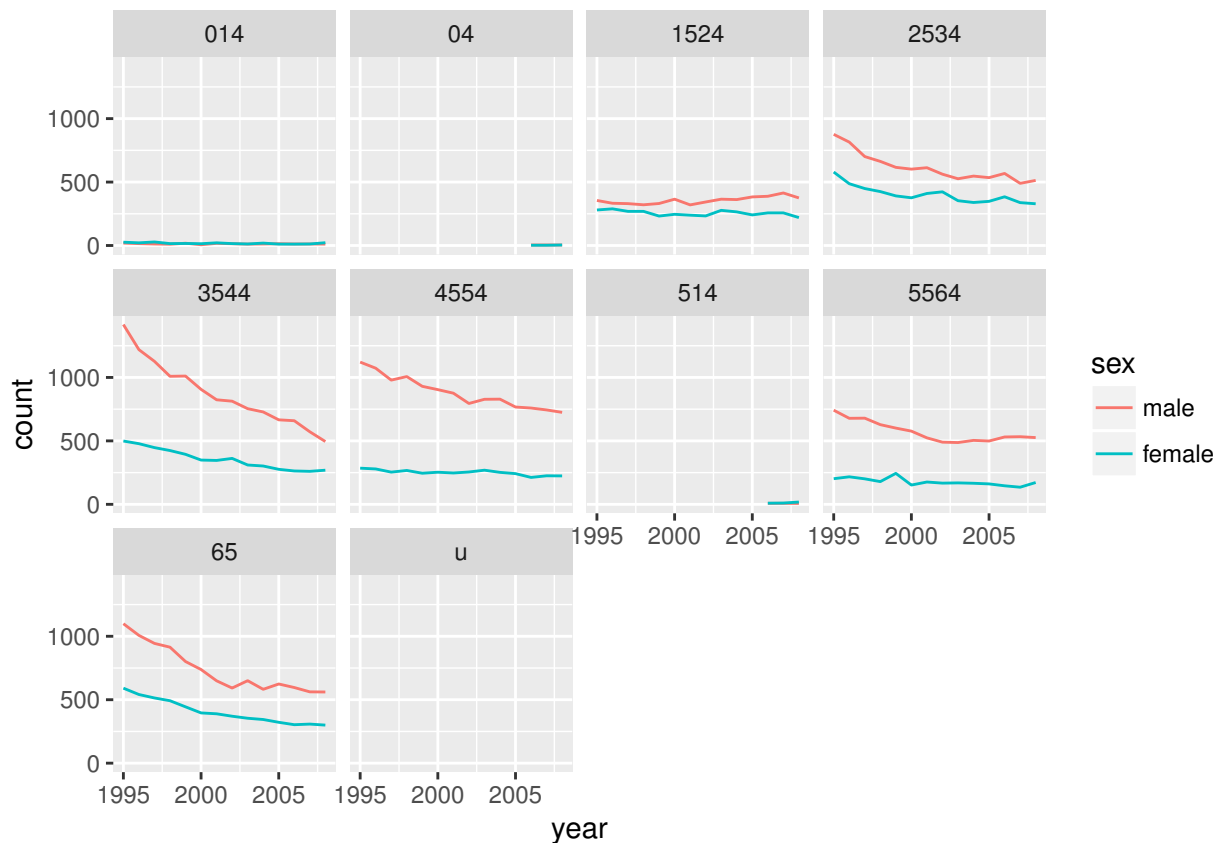
##	201	US	2006	male	3544	659
##	202	US	2007	male	3544	572
##	203	US	2008	male	3544	495
##	219	US	1995	male	4554	1121
##	220	US	1996	male	4554	1073
##	221	US	1997	male	4554	979
##	222	US	1998	male	4554	1007
##	223	US	1999	male	4554	930
##	224	US	2000	male	4554	904
##	225	US	2001	male	4554	876
##	226	US	2002	male	4554	795
##	227	US	2003	male	4554	828
##	228	US	2004	male	4554	829
##	229	US	2005	male	4554	767
##	230	US	2006	male	4554	759
##	231	US	2007	male	4554	744
##	232	US	2008	male	4554	725
##	248	US	1995	male	5564	742
##	249	US	1996	male	5564	678
##	250	US	1997	male	5564	679
##	251	US	1998	male	5564	628
##	252	US	1999	male	5564	601
##	253	US	2000	male	5564	577
##	254	US	2001	male	5564	524
##	255	US	2002	male	5564	490
##	256	US	2003	male	5564	487
##	257	US	2004	male	5564	504
##	258	US	2005	male	5564	499
##	259	US	2006	male	5564	531
##	260	US	2007	male	5564	533
##	261	US	2008	male	5564	526
##	277	US	1995	male	65	1099
##	278	US	1996	male	65	1007
##	279	US	1997	male	65	944
##	280	US	1998	male	65	914
##	281	US	1999	male	65	801
##	282	US	2000	male	65	738
##	283	US	2001	male	65	649
##	284	US	2002	male	65	592
##	285	US	2003	male	65	650
##	286	US	2004	male	65	582
##	287	US	2005	male	65	624
##	288	US	2006	male	65	596
##	289	US	2007	male	65	562
##	290	US	2008	male	65	561
##	319	US	2008	male	u	0
##	346	US	2006	female	04	2
##	347	US	2007	female	04	2
##	348	US	2008	female	04	4
##	375	US	2006	female	514	9
##	376	US	2007	female	514	10
##	377	US	2008	female	514	18
##	393	US	1995	female	014	26
##	394	US	1996	female	014	21

##	395	US	1997	female	014	28
##	396	US	1998	female	014	15
##	397	US	1999	female	014	16
##	398	US	2000	female	014	14
##	399	US	2001	female	014	21
##	400	US	2002	female	014	15
##	401	US	2003	female	014	12
##	402	US	2004	female	014	19
##	403	US	2005	female	014	11
##	404	US	2006	female	014	11
##	405	US	2007	female	014	12
##	406	US	2008	female	014	22
##	422	US	1995	female	1524	280
##	423	US	1996	female	1524	289
##	424	US	1997	female	1524	269
##	425	US	1998	female	1524	269
##	426	US	1999	female	1524	232
##	427	US	2000	female	1524	246
##	428	US	2001	female	1524	239
##	429	US	2002	female	1524	233
##	430	US	2003	female	1524	277
##	431	US	2004	female	1524	265
##	432	US	2005	female	1524	241
##	433	US	2006	female	1524	257
##	434	US	2007	female	1524	257
##	435	US	2008	female	1524	220
##	451	US	1995	female	2534	579
##	452	US	1996	female	2534	487
##	453	US	1997	female	2534	449
##	454	US	1998	female	2534	425
##	455	US	1999	female	2534	391
##	456	US	2000	female	2534	376
##	457	US	2001	female	2534	410
##	458	US	2002	female	2534	423
##	459	US	2003	female	2534	353
##	460	US	2004	female	2534	339
##	461	US	2005	female	2534	348
##	462	US	2006	female	2534	384
##	463	US	2007	female	2534	338
##	464	US	2008	female	2534	329
##	480	US	1995	female	3544	499
##	481	US	1996	female	3544	478
##	482	US	1997	female	3544	447
##	483	US	1998	female	3544	424
##	484	US	1999	female	3544	394
##	485	US	2000	female	3544	349
##	486	US	2001	female	3544	346
##	487	US	2002	female	3544	362
##	488	US	2003	female	3544	310
##	489	US	2004	female	3544	302
##	490	US	2005	female	3544	276
##	491	US	2006	female	3544	263
##	492	US	2007	female	3544	260
##	493	US	2008	female	3544	269


```
## 509 US 1995 female 4554 285
## 510 US 1996 female 4554 279
## 511 US 1997 female 4554 254
## 512 US 1998 female 4554 267
## 513 US 1999 female 4554 245
## 514 US 2000 female 4554 253
## 515 US 2001 female 4554 247
## 516 US 2002 female 4554 255
## 517 US 2003 female 4554 269
## 518 US 2004 female 4554 252
## 519 US 2005 female 4554 242
## 520 US 2006 female 4554 212
## 521 US 2007 female 4554 225
## 522 US 2008 female 4554 224
## 538 US 1995 female 5564 202
## 539 US 1996 female 5564 217
## 540 US 1997 female 5564 201
## 541 US 1998 female 5564 179
## 542 US 1999 female 5564 244
## 543 US 2000 female 5564 152
## 544 US 2001 female 5564 176
## 545 US 2002 female 5564 167
## 546 US 2003 female 5564 169
## 547 US 2004 female 5564 166
## 548 US 2005 female 5564 161
## 549 US 2006 female 5564 146
## 550 US 2007 female 5564 135
## 551 US 2008 female 5564 172
## 567 US 1995 female 65 591
## 568 US 1996 female 65 541
## 569 US 1997 female 65 514
## 570 US 1998 female 65 492
## 571 US 1999 female 65 444
## 572 US 2000 female 65 396
## 573 US 2001 female 65 389
## 574 US 2002 female 65 370
## 575 US 2003 female 65 354
## 576 US 2004 female 65 344
## 577 US 2005 female 65 322
## 578 US 2006 female 65 303
## 579 US 2007 female 65 308
## 580 US 2008 female 65 300
## 609 US 2008 female u 0
```

Answer the question:

```
library(ggplot2)
ggplot(tb.sep, aes(x = year, y = count, color = sex)) + geom_line() + facet_wrap(~age)
```



In all age groups with non-negligible counts, men have higher TB rates than women.

10.2 Baseball

```
# install.packages('Lahman')
library(Lahman)
```

Batting data:

```
# Batting = as_tibble(Batting)
dim(Batting)
```

```
## [1] 101332    22
```

```
names(Batting)
```

```
## [1] "playerID" "yearID"   "stint"    "teamID"   "lgID"     "G"
## [7] "AB"       "R"        "H"        "X2B"      "X3B"      "HR"
## [13] "RBI"      "SB"       "CS"       "BB"       "SO"       "IBB"
## [19] "HBP"      "SH"       "SF"       "GIDP"
```

Our goal is to predict a player's 2015 statistics based on 2012 to 2014 statistics. Get 2012–2015 data:

```
pred.year = 2015
B = subset(Batting, yearID >= pred.year - 3 & yearID <= pred.year)
```

Add a variable for “plate appearances”:

```
B = transform(B, PA = AB + BB + HBP + SF + SH)
```

Aggregate by year. We'll use `ddply()` from the `plyr` library (there are also functions to do this in `dplyr` but I haven't learned them.)

```
stats = c("PA", "AB", "R", "H", "X2B", "X3B", "HR", "RBI", "SB", "CS", "BB",
          "SO", "IBB", "HBP", "SH", "SF", "GIDP")
B = ddply(B[, c("playerID", "yearID", stats)], ~playerID + yearID, summarise,
          PA = sum(PA), AB = sum(AB), R = sum(R), H = sum(H), X2B = sum(X2B), X3B = sum(X3B),
          HR = sum(HR), RBI = sum(RBI), SB = sum(SB), CS = sum(CS), BB = sum(BB),
          SO = sum(SO), IBB = sum(IBB), HBP = sum(HBP), SH = sum(SH), SF = sum(SF),
          GIDP = sum(GIDP))
dim(B)
```

```
## [1] 5256 19
```

```
summary(B)
```

```
##      playerID      yearID      PA      AB
## Length:5256      Min.   :2012      Min.   : 0.0      Min.   : 0
## Class :character  1st Qu.:2013      1st Qu.: 0.0      1st Qu.: 0
## Mode  :character  Median :2014      Median : 20.0     Median : 18
##                               Mean  :2014      Mean  :140.1     Mean  :126
##                               3rd Qu.:2015      3rd Qu.:225.0     3rd Qu.:204
##                               Max.   :2015      Max.   :740.0     Max.   :684
##      R      H      X2B      X3B
## Min.   : 0.00      Min.   : 0.00      Min.   : 0.0000      Min.   : 0.0000
## 1st Qu.: 0.00      1st Qu.: 0.00      1st Qu.: 0.0000      1st Qu.: 0.0000
## Median : 1.00      Median : 3.00      Median : 0.0000      Median : 0.0000
## Mean   : 15.54      Mean   : 31.94      Mean   : 6.252      Mean   : 0.6634
## 3rd Qu.: 22.00      3rd Qu.: 48.00      3rd Qu.: 9.000      3rd Qu.: 0.0000
## Max.   :129.00      Max.   :225.00      Max.   :55.000      Max.   :15.0000
##      HR      RBI      SB      CS
## Min.   : 0.000      Min.   : 0.00      Min.   : 0.000      Min.   : 0.0000
## 1st Qu.: 0.000      1st Qu.: 0.00      1st Qu.: 0.000      1st Qu.: 0.0000
## Median : 0.000      Median : 1.00      Median : 0.000      Median : 0.0000
## Mean   : 3.556      Mean   : 14.78      Mean   : 2.129      Mean   : 0.8071
## 3rd Qu.: 4.000      3rd Qu.: 20.00      3rd Qu.: 1.000      3rd Qu.: 1.0000
## Max.   :53.000      Max.   :139.00      Max.   :64.000      Max.   :23.0000
##      BB      SO      IBB      HBP
## Min.   : 0.00      Min.   : 0.00      Min.   : 0.0000      Min.   : 0.000
## 1st Qu.: 0.00      1st Qu.: 0.00      1st Qu.: 0.0000      1st Qu.: 0.000
## Median : 1.00      Median : 6.00      Median : 0.0000      Median : 0.000
## Mean   : 10.93      Mean   : 28.16      Mean   : 0.7627      Mean   : 1.196
## 3rd Qu.: 15.00      3rd Qu.: 45.00      3rd Qu.: 0.0000      3rd Qu.: 1.000
## Max.   :143.00      Max.   :222.00      Max.   :29.0000      Max.   :30.000
##      SH      SF      GIDP
## Min.   : 0.000      Min.   : 0.000      Min.   : 0.000
## 1st Qu.: 0.000      1st Qu.: 0.000      1st Qu.: 0.000
## Median : 0.000      Median : 0.000      Median : 0.000
## Mean   : 1.028      Mean   : 0.942      Mean   : 2.796
## 3rd Qu.: 1.000      3rd Qu.: 1.000      3rd Qu.: 4.000
## Max.   :17.000      Max.   :12.000      Max.   :31.000
```

Pick out years:

```
B.short = B[, c(1, 2, 3, 4, 5, 6, 9, 10, 13, 14)]
names(B.short)
```

```
## [1] "playerID" "yearID" "PA" "AB" "R" "H"
## [7] "HR" "RBI" "BB" "SO"
```

```
B.2012 = B.short[B.short$yearID == 2012, ]
B.2013 = B.short[B.short$yearID == 2013, ]
B.2014 = B.short[B.short$yearID == 2014, ]
B.2015 = B.short[B.short$yearID == 2015, ]
```

Add names to things:

```
names(B.2012) = c("playerID", "yearID.1", "PA.1", "AB.1", "R.1", "H.1", "HR.1",
  "RBI.1", "BB.1", "SO.1")
names(B.2013) = c("playerID", "yearID.2", "PA.2", "AB.2", "R.2", "H.2", "HR.2",
  "RBI.2", "BB.2", "SO.2")
names(B.2014) = c("playerID", "yearID.3", "PA.3", "AB.3", "R.3", "H.3", "HR.3",
  "RBI.3", "BB.3", "SO.3")
names(B.2015) = c("playerID", "yearID.4", "PA.4", "AB.4", "R.4", "H.4", "HR.4",
  "RBI.4", "BB.4", "SO.4")
```

10.2.1 Merging things

For easy problems `cbind()` and `rbind()` may suffice, but this is harder. Instead, we'll `merge()` everything:

```
B2 = merge(B.2012, B.2013, by = "playerID")
B3 = merge(B2, B.2014, by = "playerID")
B4 = merge(B3, B.2015, by = "playerID")
```

Only keep the complete cases:

```
B.complete = B4[complete.cases(B4), ]
PAs = (B.complete$PA.1 > 0) * (B.complete$PA.2 > 0) * (B.complete$PA.3 > 0) *
  (B.complete$PA.4 > 0)
B.all = B.complete[PAs == 1, ]
```

What if we wanted to include birth year? (We don't need it here, but age would be the next variable that I'd try to include in a model.) Let's add that in from the `Master` data frame:

```
birth = Master[, c("playerID", "birthYear")]
B.merged = merge(B.all, birth, by = "playerID")
```

10.2.2 Making a model

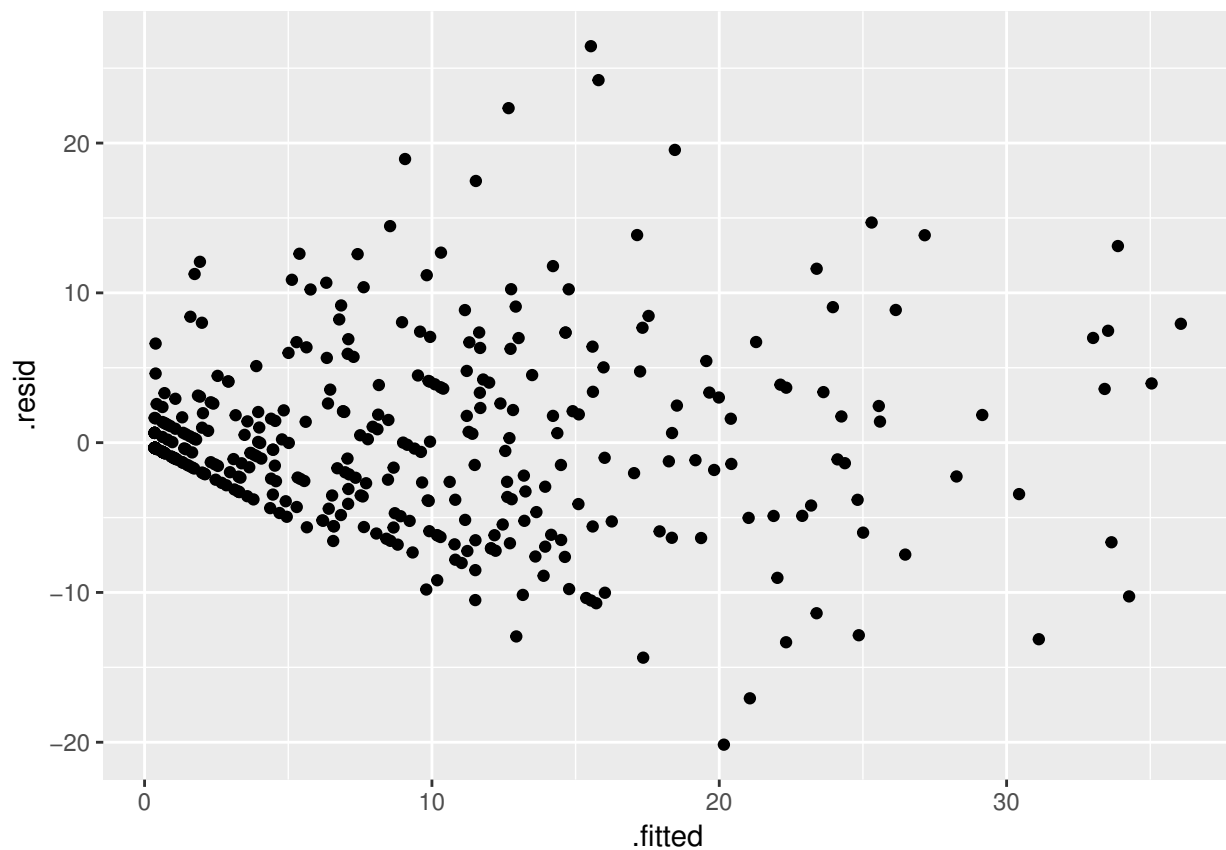
Let's predict home runs based on their past three years of home runs:

```
homerun.lm = lm(HR.4 ~ HR.1 + HR.2 + HR.3, data = B.merged)
library(broom)
tidy(homerun.lm)
```

```
##      term estimate std.error statistic    p.value
## 1 (Intercept) 0.3487871 0.36773423  0.9484760 3.434078e-01
## 2      HR.1 0.0358215 0.04276830  0.8375714 4.027262e-01
## 3      HR.2 0.2732422 0.05269861  5.1849986 3.300467e-07
## 4      HR.3 0.6869728 0.04902130 14.0137630 3.751616e-37
```

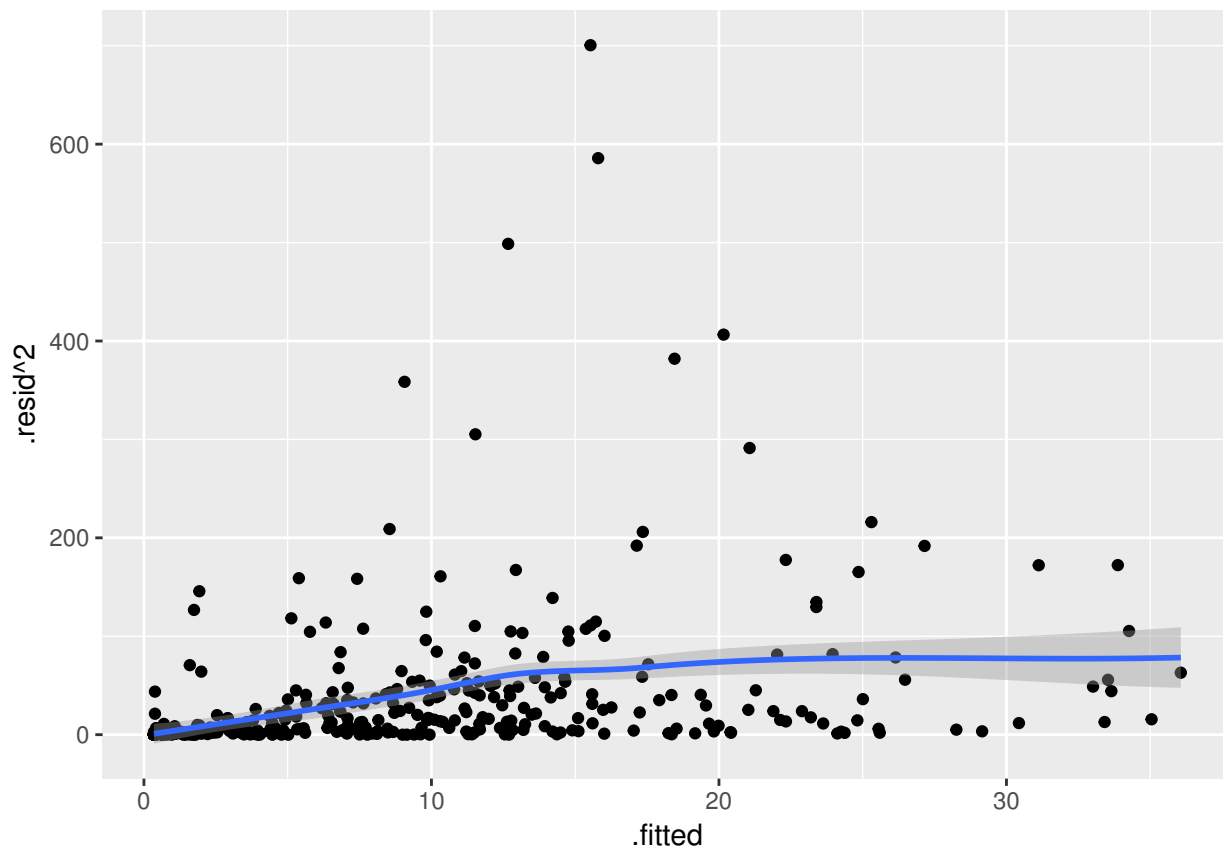
The coefficients get smaller for older years, which makes sense. Look at the residuals:

```
homerun.df = augment(homerun.lm)
ggplot(homerun.df, aes(x = .fitted, y = .resid)) + geom_point()
```



It's heteroskedastic. How do the squared residuals change?

```
homerun.df = augment(homerun.lm)
ggplot(homerun.df, aes(x = .fitted, y = .resid^2)) + geom_point() + geom_smooth(method.args = list(degre
```

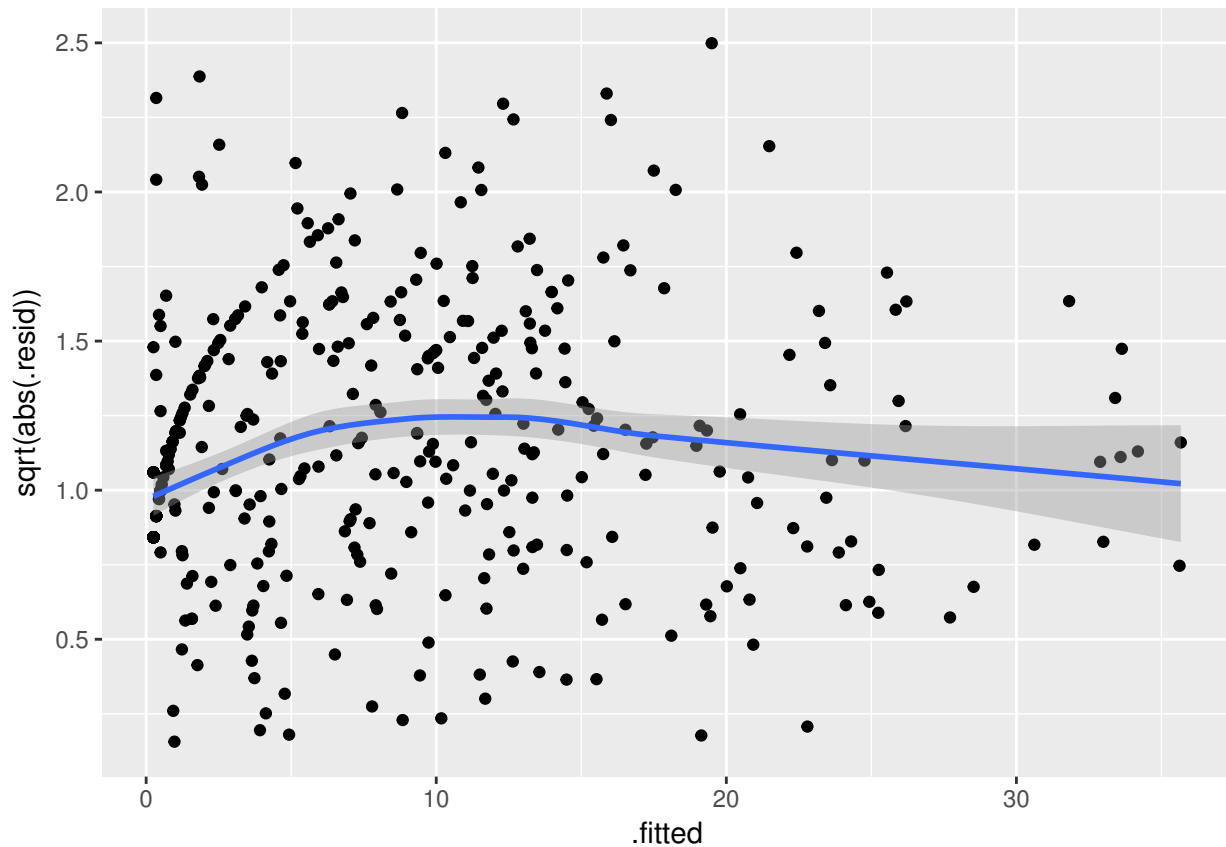


An overdispersed Poisson might work okay. It doesn't make much sense to predict on the log scale, so we fit on the original scale instead.

```
homerun.quasi = glm(HR.4 ~ HR.1 + HR.2 + HR.3, start = coef(homerun.lm), family = quasipoisson(link = "log",
  data = B.merged)
tidy(homerun.quasi)
```

```
##           term  estimate std.error statistic    p.value
## 1 (Intercept) 0.25183823 0.09422345  2.672777 7.802190e-03
## 2           HR.1 0.09569843 0.04704579  2.034155 4.253523e-02
## 3           HR.2 0.24533197 0.05891869  4.163907 3.765670e-05
## 4           HR.3 0.66236049 0.06091515 10.873493 1.491511e-24
```

```
homerun.quasi.df = augment(homerun.quasi)
ggplot(homerun.quasi.df, aes(x = .fitted, y = sqrt(abs(.resid)))) + geom_point() +
  geom_smooth(method.args = list(degree = 1))
```



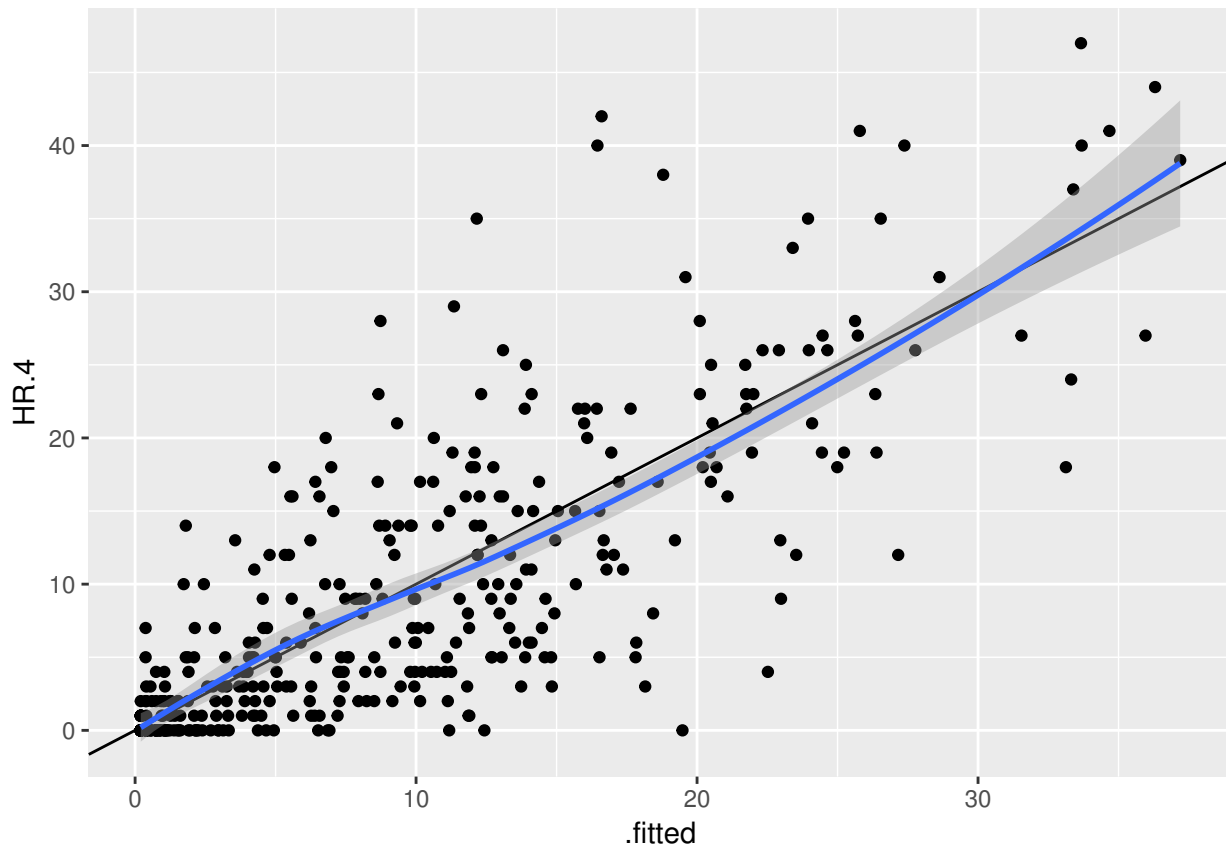
The coefficients are a bit different and the standard errors are a bit bigger. The deviance residuals aren't perfect but they could be worse.

Alternatively, try a negative binomial, which can give probabilistic predictions:

```
library(MASS)
homerun.nb = glm.nb(HR.4 ~ HR.1 + HR.2 + HR.3, data = B.merged, start = coef(homerun.lm),
  link = identity)
tidy(homerun.nb)
```

```
##           term estimate std.error statistic    p.value
## 1 (Intercept) 0.2018393 0.04744128  4.254507 2.095097e-05
## 2           HR.1 0.1729842 0.04796215  3.606681 3.101384e-04
## 3           HR.2 0.1971358 0.05634843  3.498515 4.678576e-04
## 4           HR.3 0.6656151 0.06987018  9.526455 1.627453e-21
```

```
homerun.nb.df = augment(homerun.nb, type.residuals = "response")
ggplot(homerun.nb.df, aes(x = .fitted, y = HR.4)) + geom_point() + geom_abline() +
  geom_smooth()
```



10.3 Catterplots

These are self-explanatory I think.

```
# library(devtools) install_github('Gibbsdavidl/CatterPlots')
library(CatterPlots)
cat.random = multicat(xs = runif(42), ys = runif(42), cat = 1:11, catcolor = list(c(0,
0, 0, 1)), main = "42 random cats")
```