

A Quality Index for New Job Hires

Brad Hershbein

W.E. Upjohn Institute for Employment Research

May 2017

Introduction

- Much timely (monthly) government data on number of new jobs
 - But by industry, not occupation or much else
- There are also considerable wage data for workers
 - But almost always for incumbents, not new hires
- The result is that we don't know much about the “quality” of new jobs

Motivation

- Understanding characteristics of new jobs, and workers in them, of key concern
 - An important coincident, and perhaps leading, indicator
 - Provides insight into cyclical labor markets
 - Can shed light on structural changes in skill demand
- But “quality” is always hard to define
 - Wage is often a useful summary statistic, but...
 - Other nuances important, especially volume of new hires
 - How much detail is possible? Useful?
- **Goal: Create a new index of job hires quality**

Occupations vs. Industries

- Economic literature has long recognized that what one does affects compensation more than where one does it
 - Roy (1951); Houty (1958, 1961); Groshen (1991)
 - And now task-based models of human capital: Spitz-Oener (2006); Gathmann & Schoenberg 2010; Acemoglu & Autor (2011); Autor (2013)
- Mincer-style wage regressions show that occupations explain 2–4 times the variance of industries, even with additional controls
- Despite this, armchair analysis on wages of new jobs is often based on industry, not occupation
 - Unlike for industries, no high-frequency occupation-level releases...
 - Result is lamp-post inference



National Employment Law Project

DATA BRIEF

April 2014

The Low-Wage Recovery:

Industry Employment and Wages Four Years into the Recovery

Most jobs added in Boston since recession called low-paying

By Katie Johnston | GLOBE STAFF SEPTEMBER 22, 2015

Many of the job gains have come in low-paying sectors such as food service, home health care, and janitorial services, while higher-paying fields such as information services, have not grown as rapidly.

The Washington Post

The ‘low-wage recovery’ is a myth

To do this, she divided businesses into three groups by their pay. Today’s average hourly pay is \$25. Low-paying employment is dominated by restaurant and hotel jobs (2015 average hourly rate: \$14.12) and retail jobs (\$17.21). Midlevel jobs include manufacturing (\$23.90), health care and education (\$24.97) and construction (\$26.91). Finally, high-paying jobs included professional and business services (\$29.59), finance (\$31.10) and utilities (\$36.02).

NELP

National Employment Law Project

DATA BRIEF

April 2014

The Low-Wage Recovery:

Industry Employment and Wages Four Years into the Recovery

Most jobs added in Boston since recession called low-paying

By Katie Johnston | GLOBE STAFF SEPTEMBER 22, 2015

Many of the job gains have come in low-paying sectors such as food service, home health care, and janitorial services, while higher-paying fields such as information services, have not grown as rapidly.

The Washington Post

The ‘low-wage recovery’ is a myth

To do this, she divided businesses into three groups by their pay. Today’s average hourly pay is \$25. Low-paying employment is dominated by restaurant and hotel jobs (2015 average hourly rate: \$14.12) and retail jobs (\$17.21). Midlevel jobs include manufacturing (\$23.90), health care and education (\$24.97) and construction (\$26.91). Finally, high-paying jobs included professional and business services (\$29.59), finance (\$31.10) and utilities (\$36.02).

The New York Times

The Upshot

EVERYDAY ECONOMICS

Justin Wolfers @JustinWolfers

There are many highly paid managers working in the low-paid retail trade sector, just as there are many low-paid janitors working in the high-paid professional services sector

Figuring out whether the recovery is creating “good jobs” or “bad jobs” requires looking deeply into skill levels and job responsibilities

NELP

National Employment Law Project

DATA BRIEF

April 2014

The Low-Wage Recovery:

Industry Employment and Wages Four Years into the Recovery

Most jobs added in Boston since recession called low-paying

By Katie Johnston | GLOBE STAFF | SEPTEMBER 22, 2015

Many of the job gains Comments have come in low-paying sectors such as food service, home health care, and janitorial services, while higher-paying fields such as information services, have not grown as rapidly.

The Washington Post

The ‘low-wage recovery’ is a myth

To do this, she divided businesses into three groups by their pay. Today’s average hourly pay is \$25. Low-paying employment is dominated by restaurant and hotel jobs (2015 average hourly rate: \$14.12) and retail jobs (\$17.21). Midlevel jobs include manufacturing (\$23.90), health care and education (\$24.97) and construction (\$26.91). Finally, high-paying jobs included professional and business services (\$29.59), finance (\$31.10) and utilities (\$36.02).

The New York Times

The Upshot

EVERYDAY ECONOMICS

Justin Wolfers @JustinWolfers

There are many highly paid managers working in the low-paid retail trade sector, just as there are many low-paid janitors working in the high-paid professional services sector

Figuring out whether the recovery is creating “good jobs” or “bad jobs” requires looking deeply into skill levels and job responsibilities

Exactly. Why not do this?

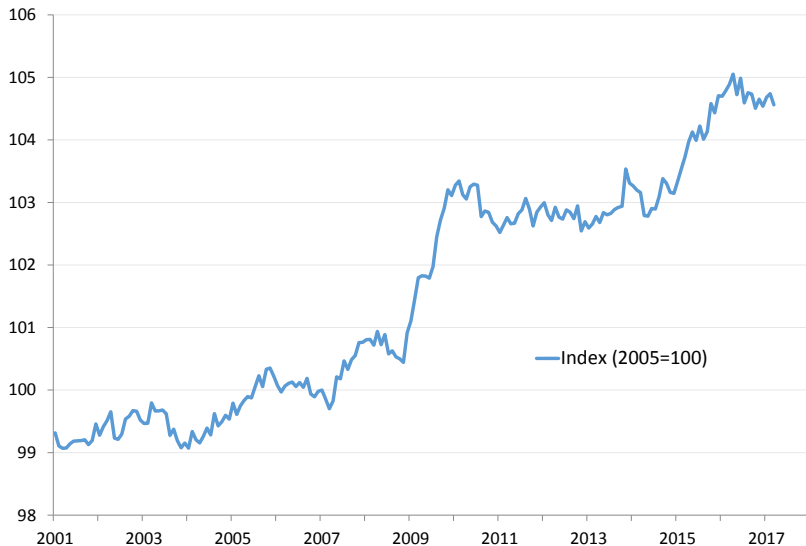
Upjohn Institute New Hires Quality Index (NHQI)

- New monthly index tracks “quality” of new job hires (2001 →)
- Uses CPS to identify new hires: those switching in adjacent months from non-employment to employment or changing employers
- Detailed occupation in CPS merged with OES occupational wage data via SOC crosswalks
 - Overcomes some weaknesses of self-reported CPS wage data
 - Automatically adjusts for inflation
- Resulting index shows change in realized skill demand through *changes in occupation mix*
 - Adjust for new-hire demographics, but not within-occupation skill changes
 - compare with self-reported wages to understand differences
 - Also yields hire volume, and index for many subgroups

Summary of findings

- 1 Hourly wage index is up nearly 5 percent from 2005

New Hires Quality Index: Hourly Wages



SOURCE: Upjohn Institute New Hires Quality Index

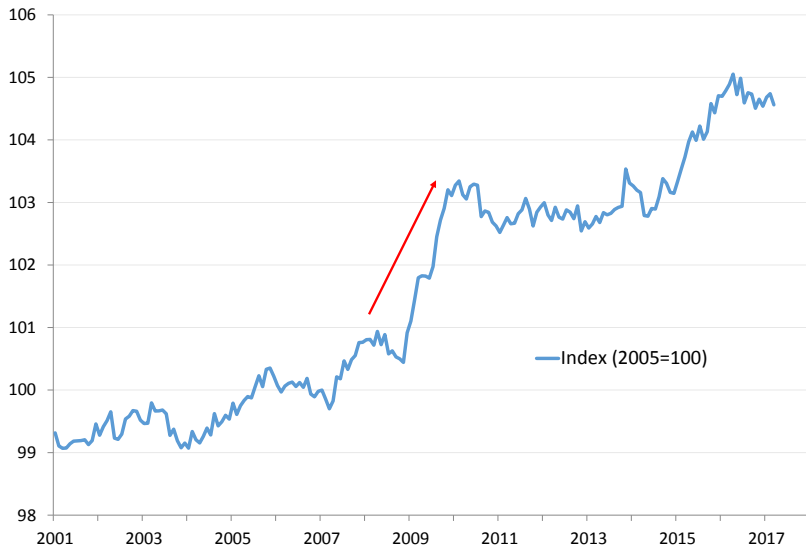
NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

Summary of findings

- 1 Hourly wage index is up nearly 5 percent from 2005
- 2 Occupational mix rose sharply during recession, was flat during recovery, and rose again from mid-2014 through 2015

New Hires Quality Index: Hourly Wages

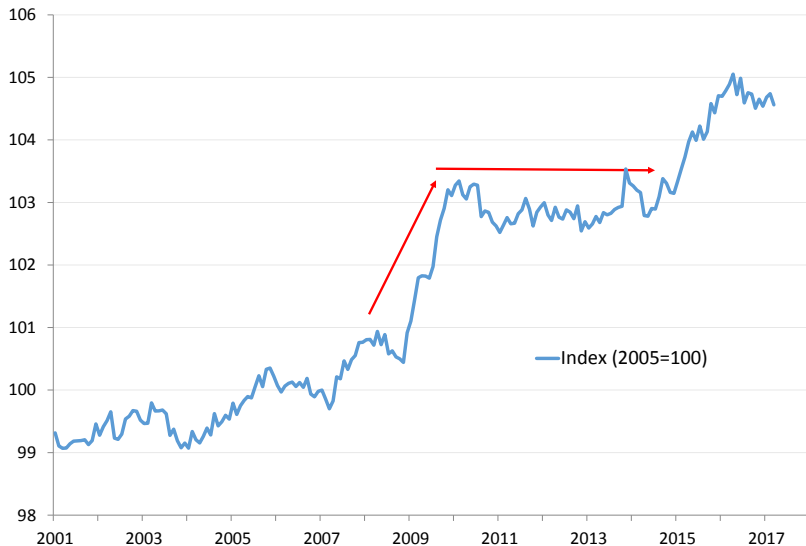


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

New Hires Quality Index: Hourly Wages

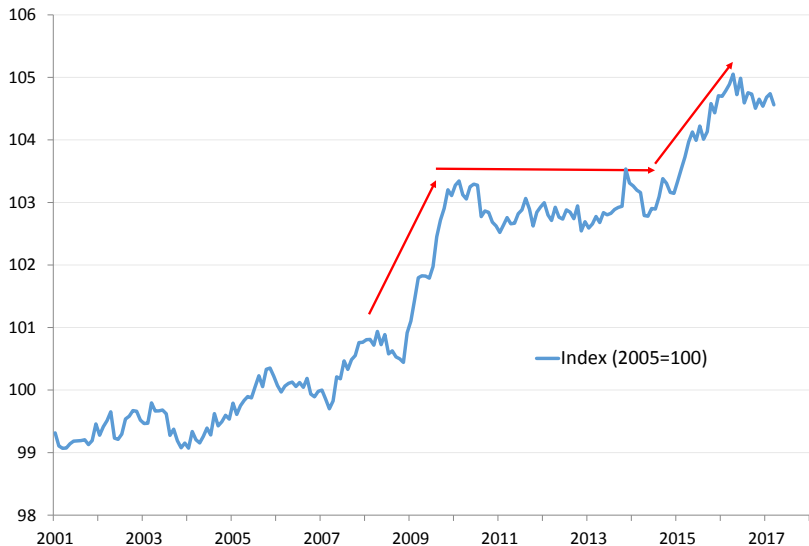


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

New Hires Quality Index: Hourly Wages



SOURCE: Upjohn Institute New Hires Quality Index

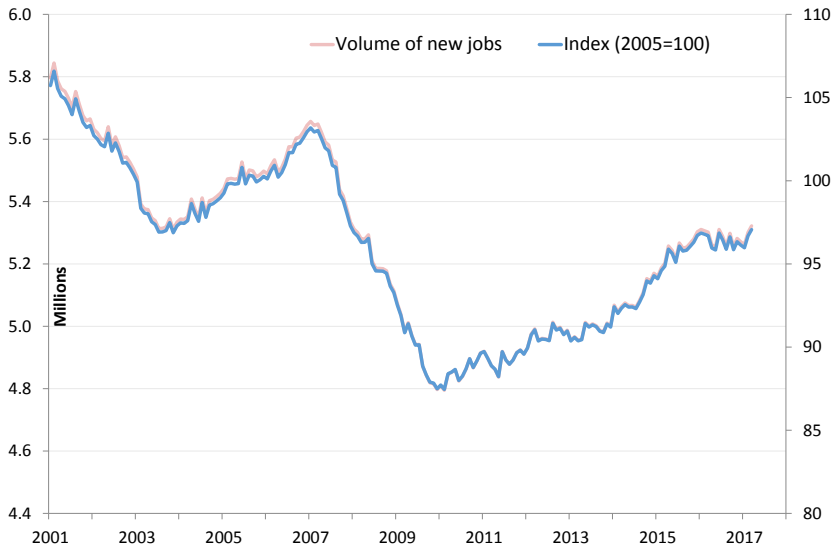
NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

Summary of findings

- 1 Hourly wage index is up nearly 5 percent from 2005
- 2 Occupational mix rose sharply during recession, was flat during recovery, and rose again from mid-2014 through 2015
- 3 Volume of new hires has not recovered; wage bill has just barely; hires/person not at all

New Hires Quality Index: Monthly Volume



SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data



New Hires Quality Index: Monthly Wage Bill

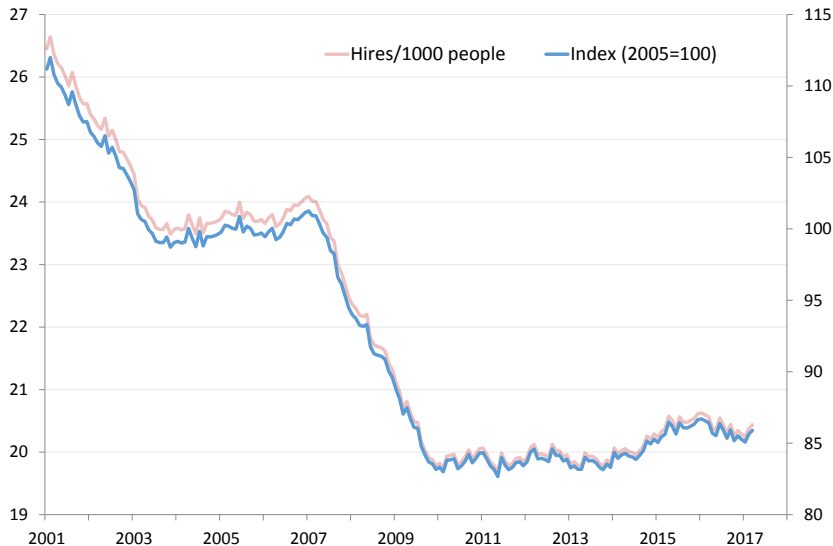


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data



New Hires Quality Index: Hires per capita



SOURCE: Upjohn Institute New Hires Quality Index

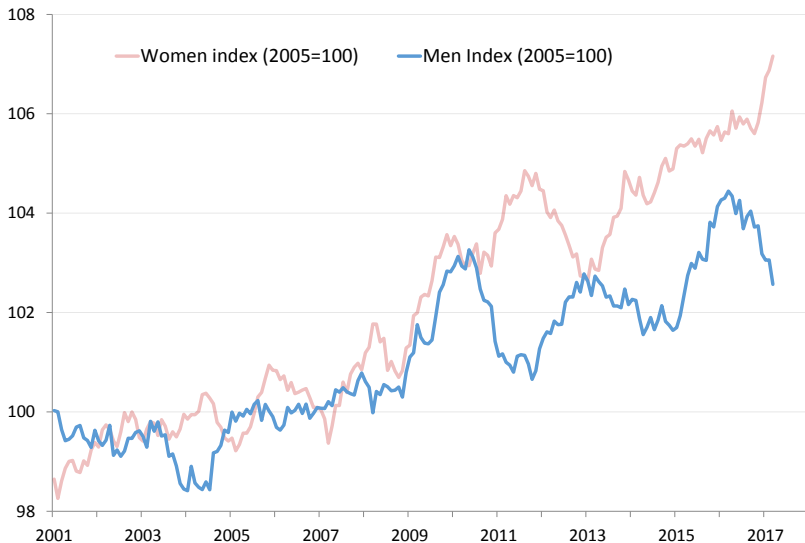
NOTE: Wage index is based on a 12-month lagged moving average of monthly data



Summary of findings

- 1 Hourly wage index is up nearly 5 percent from 2005
- 2 Occupational mix rose sharply during recession, was flat during recovery, and rose again from mid-2014 through 2015
- 3 Volume of new hires has not recovered; wage bill has just barely; hires/person not at all
- 4 Women have had a stronger recovery than men

New Hires Quality Index: Women and Men

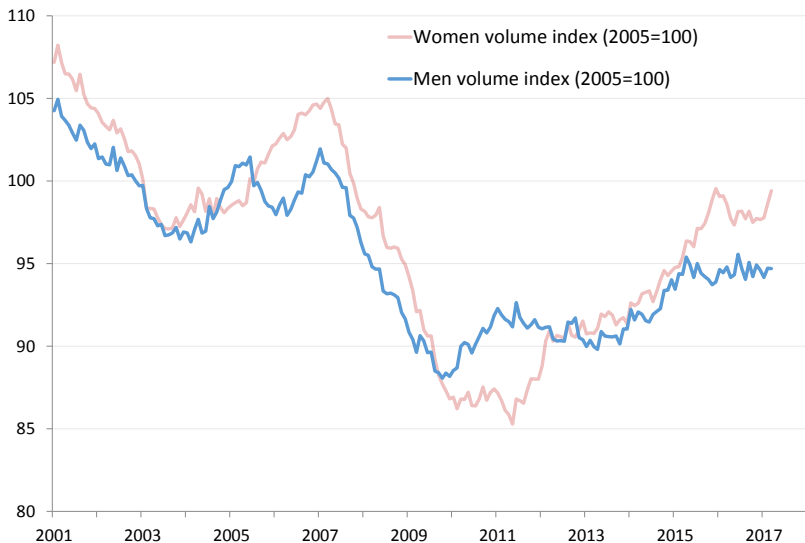


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data



New Hires Quality Index: Women and Men, volume



SOURCE: Upjohn Institute New Hires Quality Index

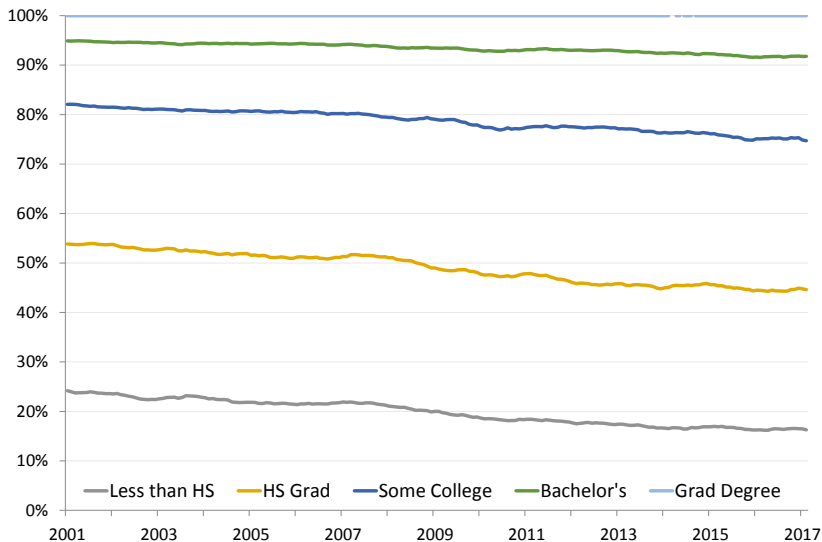
NOTE: Wage index is based on a 12-month lagged moving average of monthly data



Summary of findings

- 1 Hourly wage index is up nearly 5 percent from 2005
- 2 Occupational mix rose sharply during recession, was flat during recovery, and rose again from mid-2014 through 2015
- 3 Volume of new hires has not recovered; wage bill has just barely; hires/person not at all
- 4 Women have had a stronger recovery than men
- 5 In 2005, college graduates accounted for one-fifth of all hires; in 2016, they accounted for one-fourth

New Hires Quality Index: Volume by education

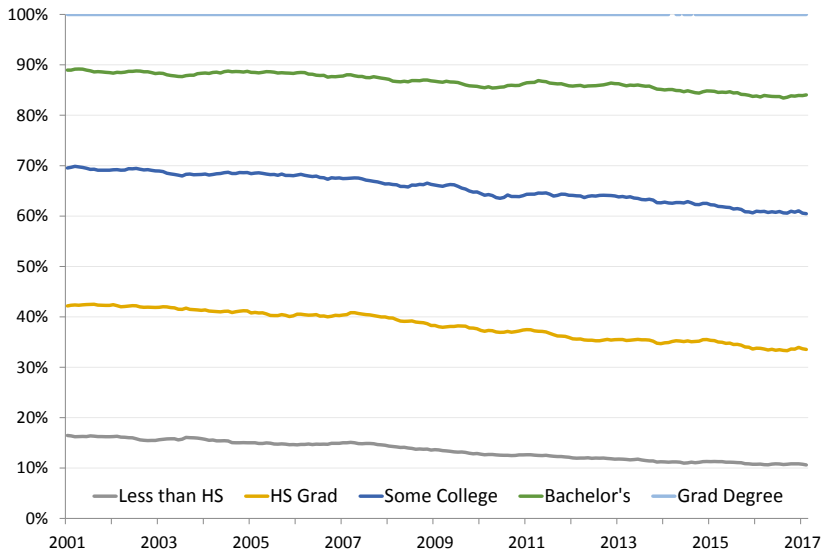


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data



New Hires Quality Index: Wage bill by education



SOURCE: Upjohn Institute New Hires Quality Index

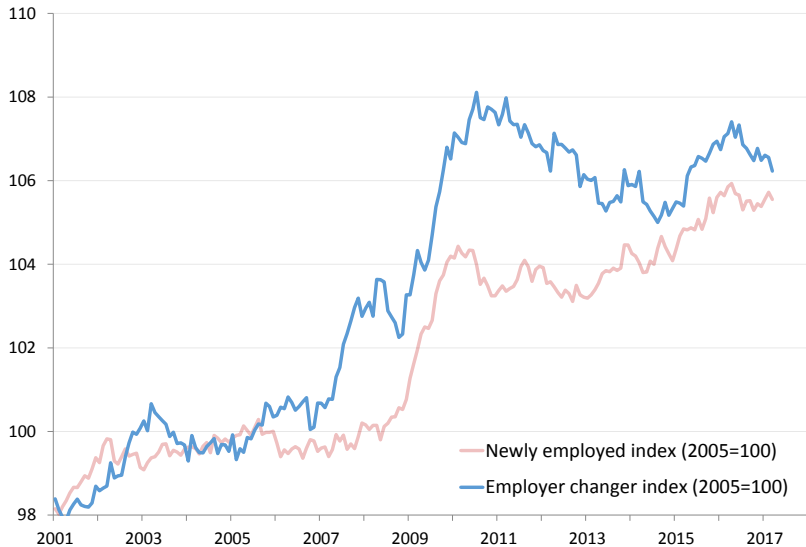
NOTE: Wage index is based on a 12-month lagged moving average of monthly data



Summary of findings

- 1 Hourly wage index is up nearly 5 percent from 2005
- 2 Occupational mix rose sharply during recession, was flat during recovery, and rose again from mid-2014 through 2015
- 3 Volume of new hires has not recovered; wage bill has just barely; hires/person not at all
- 4 Women have had a stronger recovery than men
- 5 In 2005, college graduates accounted for one-fifth of all hires; in 2016, they accounted for one-fourth
- 6 Wage index gains have been comparable for newly employed and employer changers, but volume growth of former vastly outpaces that of latter

New Hires Quality Index: Index by Hire Type

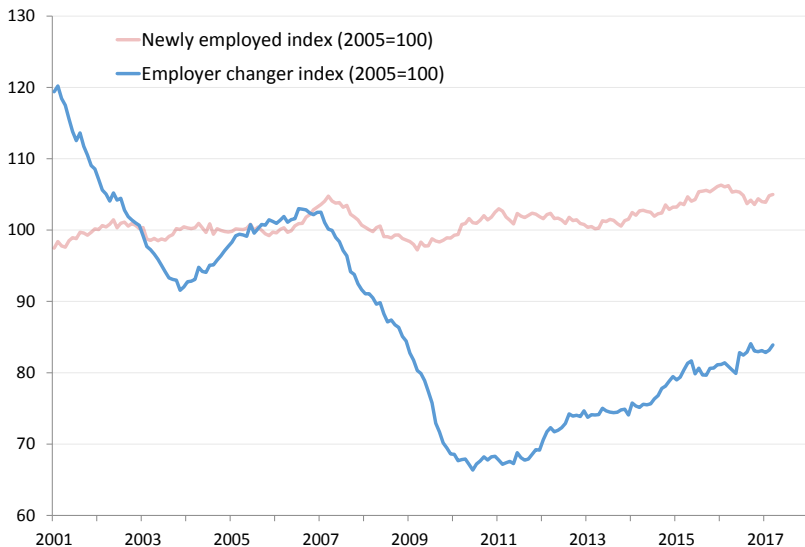


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

New Hires Quality Index: Volume by Hire Type



SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

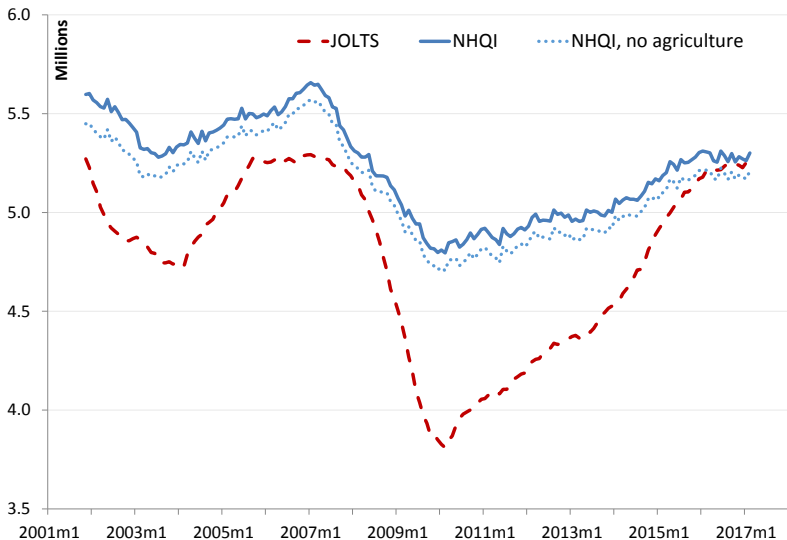


- 1 Methodology
- 2 Robustness
- 3 What about actual reported wages?
- 4 Subgroups
- 5 Conclusions

- Longitudinally link CPS (adult civilian) respondents in adjacent months (Madrian and Lefgren 2000)
- In theory, can do this for $\frac{3}{4}$ of sample (rotation groups 1–3 and 5–7)
- Will necessarily miss individuals who leave the household or move (or die)
 - Drew, Flood, and Warren (2014) show match rates of 95% of theoretical max
 - 96% successful links; 1 pp don't match age/race/sex
- Still, will check SIPP(?) to gauge magnitude of new hires who change residences
 - Probably positively selected...

- How to identify new hires (excluding self-employed)?
- For NE \rightarrow E transitions, straightforward to observe change from unemployed/NILF to employed using labor recode
- For E \rightarrow E new job transitions, exploit post-1994 variable (puiodp1) on whether employer is same as last month's
 - Will not count occupation changes with same employer (ignore internal labor market); too arbitrary and problematically measured
- Weighted aggregates compare reasonably well with JOLTS, but less cyclical
 - Conceptual differences, and JOLTS undercounts relative to QWI

CPS New Hires Volume vs JOLTS



SOURCE: Upjohn Institute New Hires Quality Index; JOLTS (BLS)

NOTE: Both measures based on a 12-month lagged moving average of NSA monthly data



- How to identify new hires (excluding self-employed)?
- For NE \rightarrow E transitions, straightforward to observe change from unemployed/NILF to employed using labor recode
- For E \rightarrow E new job transitions, exploit post-1994 variable (puiodp1) on whether employer is same as last month's
 - Will not count occupation changes with same employer (ignore internal labor market); too arbitrary and problematically measured
- Weighted aggregates compare reasonably well with JOLTS, but less cyclical
- Simple correlation is 0.92, and some conceptual differences in samples (reference period, unpaid leave, informal work)

- Need to harmonize occupation codes over time
 - From 1994 through 2002 → 1990 Census codes
 - From 2003 through 2010 → 2000 Census codes
 - From 2011 to current → 2010 Census codes
- Goal is to map to 2010 SOC codes (what OES now uses)
- For 2010 Census codes, Census crosswalk maps 532 occ codes to 532 SOC codes (out of 820)
 - Occ codes are coarser than SOCs, so some occs maps to 4- or 5-digit SOCs
- The 2011 → period is straightforward...

Methodology: CPS4

- For 2000 Census codes, IPUMS crosswalk maps 505 occ codes to 505 SOC codes (out of 801)
 - Again, occ codes are coarser than SOCs, so some occs maps to 4- or 5-digit SOCs
- But need to map 2000 SOCS → 2010 SOCS
 - Some simple 1:1 recodes or combinations, but also several splits
 - For splits, randomly assign based on empirical shares from ACS over 2010–2012
- These adjustments are minor, as most splits are into similarly paid occupations
 - Many splits into same 5-digit SOC, a few into same 4-digit SOC

- For pre-2003 period (1990 Census codes), crosswalking is a problem
- 1990 to 2000 change was very significant, reflected evolution to service-based economy
 - Census “crosswalks” show almost every occupation split into others in *both* directions
- IPUMS provides crosswalk between 1990 and 2010 occ codes
 - But it uses majority-split rule, not stochastic assignment
 - As a result, 499 1990 occ codes are mapped to only 352 2010 occ codes
- Partial solution: CPS extracts
 - BLS-released 2000–2002 files with 2000 Census codes
- Thus, focus on 2000 → period

- Also need to harmonize industries, but only at 2-digit level
- Much easier than trying detailed NAICS crosswalk
- Census industry codes map into 3-digit NAICS easily in 2003 → period
- In pre-2003 period, mapping isn't exact, but still quite good
 - And CPS extracts solve 2000–2002 period

Methodology: OES

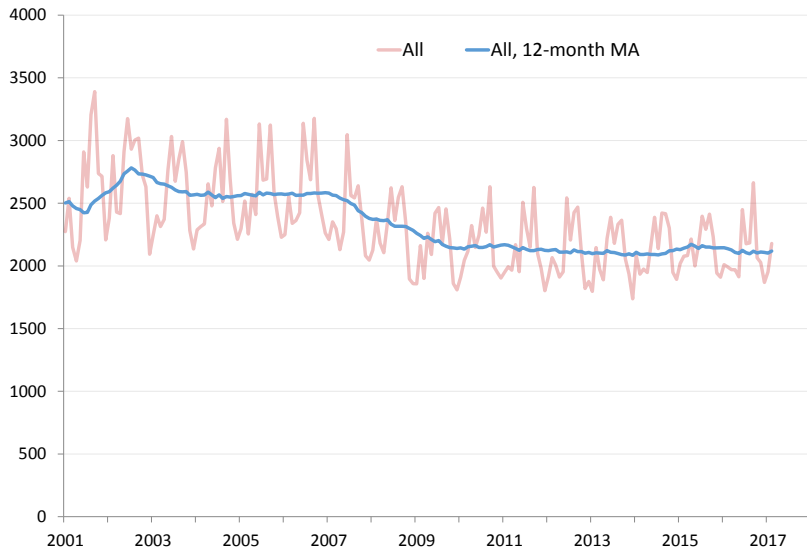
- Occupational Employment Statistics (OES) provides annual occupation-level wage data
 - At national level, available at cross of 6-digit SOC and 2-6 digit NAICS
 - Also available at MSA, state, and some state-industry levels
- Provides wage distribution (hourly or annual) at key quantiles and mean
- Merge 25th percentile occupational wages using SOC to CPS new hires
 - This quantile better approximates wages of new hires
- Merge on 6-digit SOC by 2-digit NAICS
 - Hierarchical process; use coarser SOC for unsuccessful matches

Methodology: Demographic Adjustment

- Assigning wages by occupation means wages will be the same for a 20-year-old LPN on her first job as for a 35-year-old LPN switching hospitals
- Desirable to adjust for these types of demographic differences in new hires, within occupation
- Use data on actual, valid self-reported (log hourly) wages to estimate adjustment factors
 - 1st: regress wages on non-demographics (time, worker type, hire type, occupation, industry)
 - 2nd: regress residuals, separately by 4-digit SOC, on sex, race, education, and quartic in age
 - 3rd: Use predicted values to adjust OES wages

- Calculate means, overall and for subgroups, each month
- To smooth out noise and seasonals, take 12-month lagged moving average
 - Straightforward, intuitive, and easy to implement
 - Generally yields results similar to X-13 ARIMA SA process or HP filter

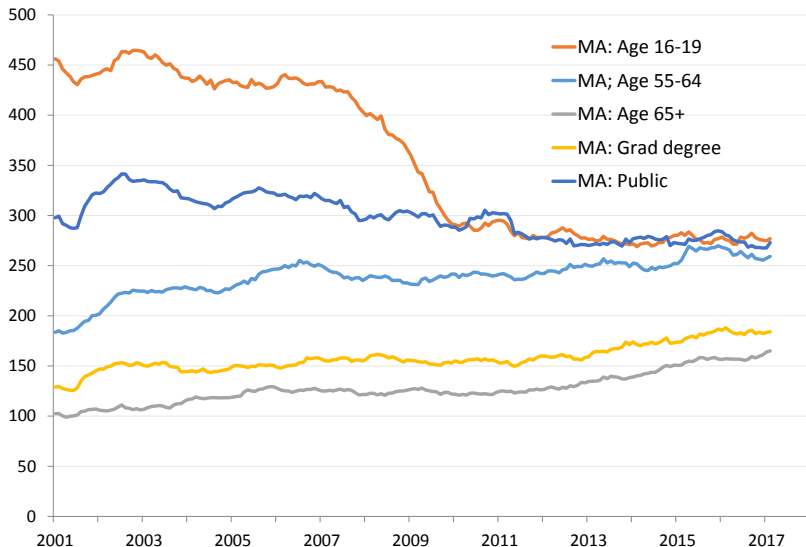
New Hires Quality Index: Sample Size Over Time



Source: Upjohn Institute New Hires Quality Index; CPS

W.E. UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

New Hires Quality Index: Sample Size, by group



Source: Upjohn Institute New Hires Quality Index; CPS

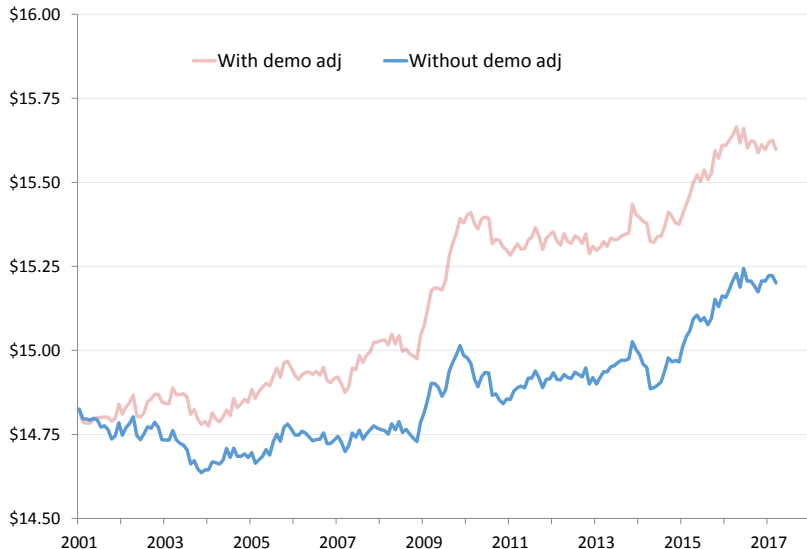
WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

- Calculate means, overall and for subgroups, each month
- To smooth out noise and seasonals, take 12-month lagged moving average
 - Straightforward, intuitive, and easy to implement
 - Generally yields results similar to X-13 ARIMA SA process or HP filter
- Taking means weights right-tail occupations more heavily
 - Could look at quantiles, too

Robustness: Demographic adjustment

- Adjustment is mostly a level shift up, overall, and again after recession
- Hires in highly paid occupations are older and more educated
- Also permanent(?) shift in hiring demographics after GR (Hershbein and Kahn 2017)

NHQI: Robust to Demographic Adjustment

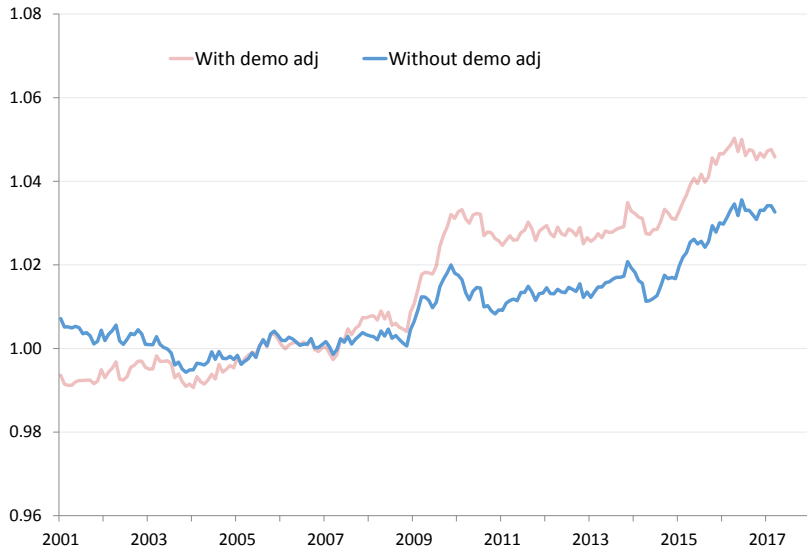


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data



Robustness: Robust to Demographic Adjustment



SOURCE: Upjohn Institute New Hires Quality Index

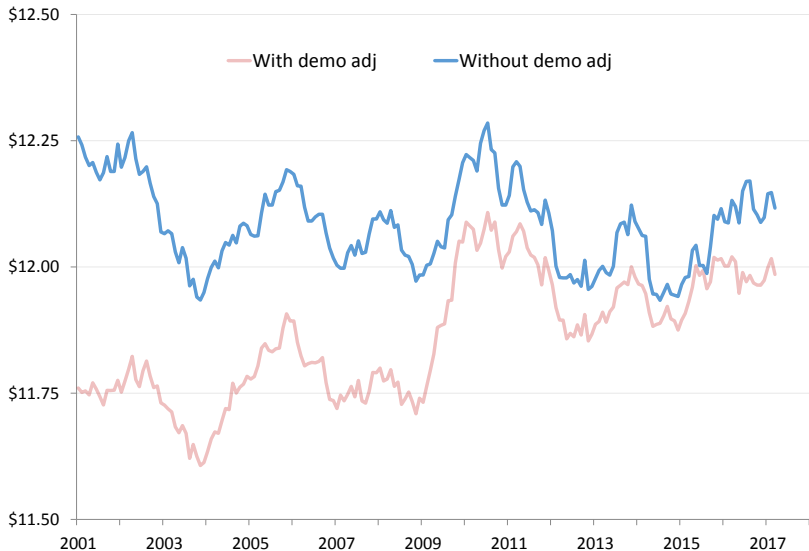
NOTE: Wage index is based on a 12-month lagged moving average of monthly data



Robustness: Median vs Mean

- Adjustment is mostly a level shift up, overall, and again after recession
 - Hires in highly paid occupations are older and more educated
 - Also permanent(?) shift in hiring demographics after GR (Hershbein and Kahn 2017)
- Can also take *median* instead of mean of new hires
 - Without demo adjustment, not very interesting...
 - Captures **only** change in median occupation hired
 - Even with demo adjustment, misses rest of distribution

Robustness: Median



SOURCE: Upjohn Institute New Hires Quality Index

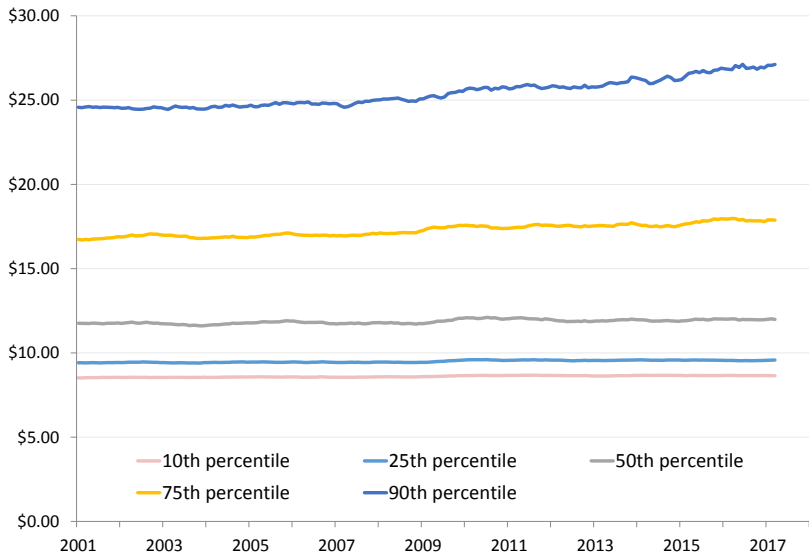
NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

Robustness: Quantiles

- Adjustment is mostly a level shift up, overall, and again after recession
 - Hires in highly paid occupations are older and more educated
 - Also permanent(?) shift in hiring demographics after GR (Hershbein and Kahn 2017)
- Can also take *median* instead of mean of new hires
 - Without demo adjustment, not very interesting...
 - Captures **only** change in median occupation hired
 - Even with demo adjustment, misses rest of distribution
- Growth is concentrated in right-tail occupations

Robustness: Quantiles

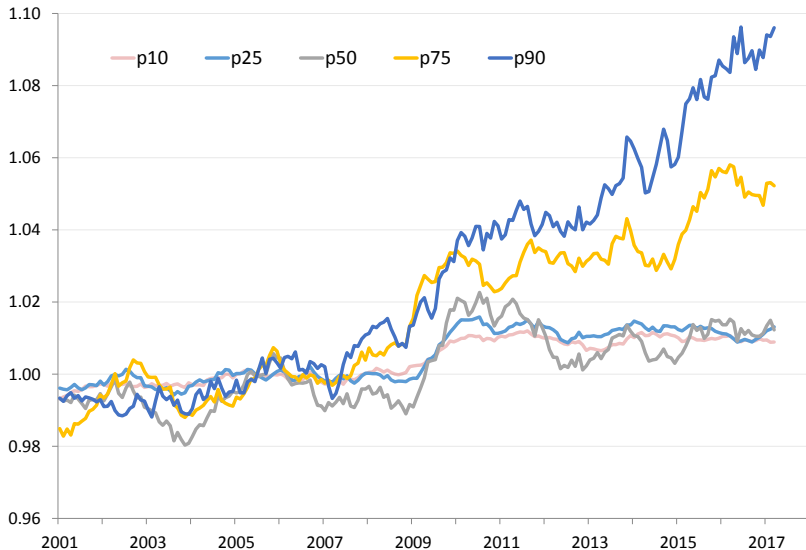


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data



Robustness: Quantiles (Index: 2005=1)

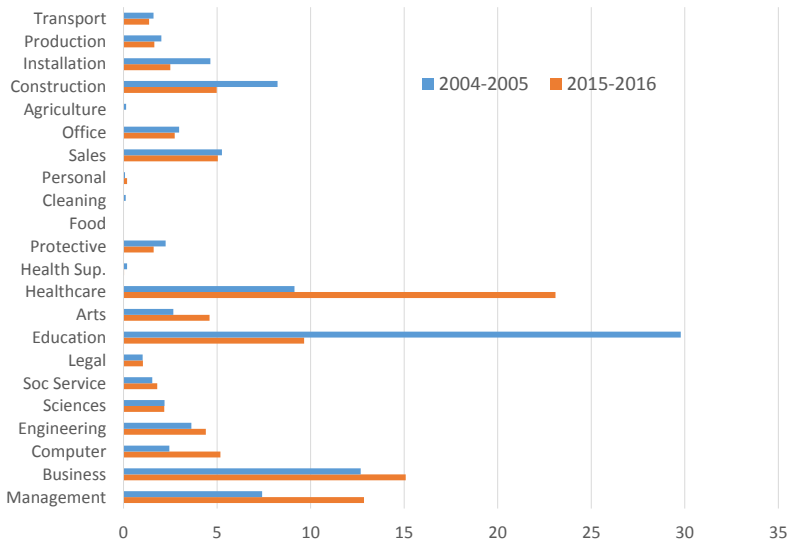


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

Occupational Distribution at 90th percentile



SOURCE: Upjohn Institute New Hires Quality Index; CPS

NOTE: Data are for 89th–91st percentile of wage index for years shown.



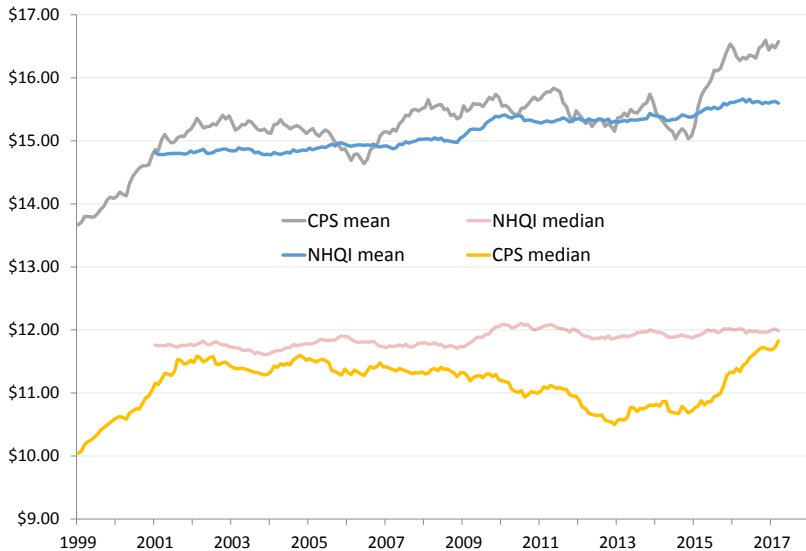
What about self-reported wages?

- Individuals report hourly (weekly) wages in ORG months... and consistent since 1994... why not use them?
- Three issues:
 - 1 Much smaller sample size: ORG restriction cuts to 1/4 size, from about 2,500 to 625 per month
 - 2 Growing imputation problem: Imputed share of wages rises from 1/4 in 1998 to 2/5 by 2016, lowering sample size to ≈ 400 today
 - 3 Composition and selection: Imputation may cause valid wages to cover different population than all new hires
- But also conceptual difference: X s vs. β s

What about self-reported wages?

- Reduced sample sizes, when averaged, sufficient for index...
- ... but not so much for subgroups
- ... and overall index still volatile, even when averaged

NHQI and CPS self-reports

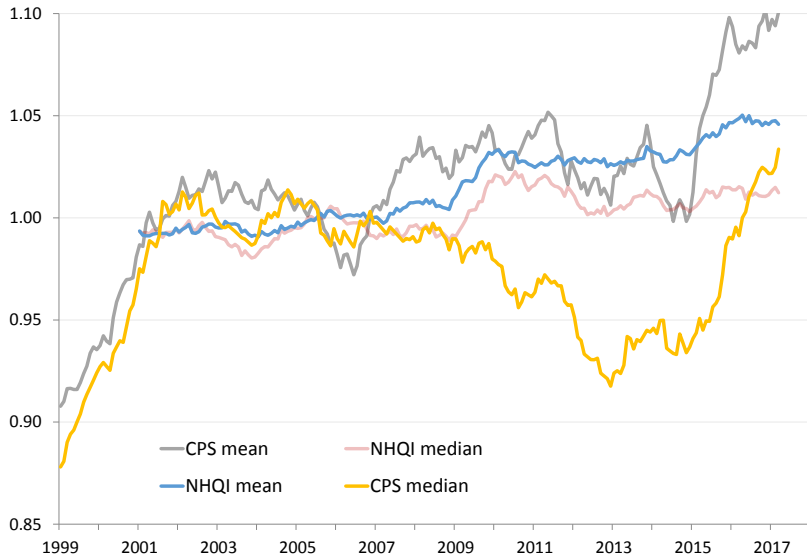


SOURCE: Upjohn Institute New Hires Quality Index; CPS

NOTE: Wage index is based on a 12-month lagged moving average of monthly data



NHQI and CPS self-reports (2005=1)



SOURCE: Upjohn Institute New Hires Quality Index; CPS

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

What about self-reported wages?

- Reduced sample sizes, when averaged, sufficient for index...
 - ... but not so much for subgroups
 - ... and overall index still volatile, even when averaged
- Strong real wage growth before 2002 (well known) and over 2015–2016 (not well known)
- Wages flat or falling even as positive occupation shift during GR
- Ocular evidence suggests roles for within-occupation and cross-occupation change at different times
 - But need to address composition bias

Composition Bias: All new hires and valid wages

	1999			2007			2016		
	All	Wage	Diff	All	Wage	Diff	All	Wage	Diff
Age	33.4	32.4	-1.1	35.2	34.0	-1.2	36.9	35.6	-1.3
Race									
White	0.682	0.695	0.012	0.628	0.651	0.024	0.572	0.588	0.016
Black	0.137	0.123	-0.014	0.130	0.108	-0.023	0.140	0.119	-0.021
Asian	0.038	0.036	-0.002	0.047	0.042	-0.005	0.058	0.056	-0.002
Hispanic	0.134	0.137	0.003	0.175	0.177	0.002	0.205	0.209	0.005
Education									
< HS	0.242	0.252	0.010	0.212	0.219	0.006	0.165	0.163	-0.002
HS grad	0.307	0.297	-0.010	0.300	0.296	-0.004	0.284	0.278	-0.006
Some college	0.279	0.291	0.012**	0.282	0.281	-0.002	0.304	0.318	0.014
Bachelor's	0.124	0.116	-0.008**	0.143	0.142	-0.001	0.165	0.162	-0.003
Grad degree	0.047	0.044	-0.003	0.062	0.062	0.000	0.082	0.079	-0.003
Sector									
Goods	0.214	0.209	-0.005	0.194	0.194	-0.000	0.167	0.156	-0.010
Services	0.786	0.791	0.005	0.806	0.806	0.000	0.833	0.843	0.010
Hire type									
Newly employed	0.581	0.560	-0.021	0.646	0.614	-0.032	0.675	0.617	-0.058
Change employer	0.419	0.440	0.021	0.354	0.386	0.032	0.325	0.383	0.058

SOURCE: Upjohn Institute New Hires Quality Index; CPS

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

Composition bias

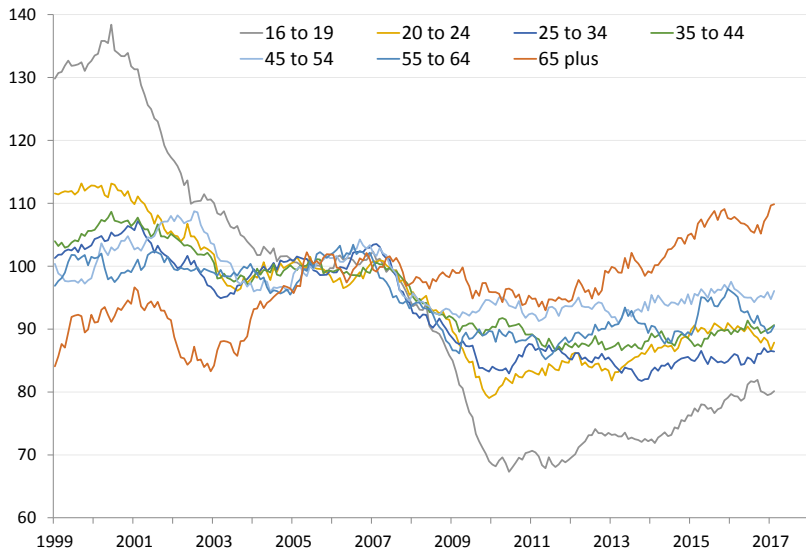
- On most observables, seems small
 - Valid-wage sample is younger, less Black, and more $E \rightarrow E$
- Tried reweighting valid wage sample to all new hires, but did not work well
 - Insufficient predictors
- Can back out expected bias (from observables)
 - Run (valid) wage regression on \mathbf{X} and adjust for $\Delta\mathbf{X}$
 - Results imply about 1% negative bias, mostly from age
 - Adding occupation and industry to \mathbf{X} increases bias slightly, to 2.7%
 - Stable over time, for offsetting reasons

Self-reported wages?

- Could potentially use for index
 - Bias is apparently small
- But n is too small for subgroups, even when averaging
- Tradeoff between simplicity and breadth of applicability

- Index is currently calculated for 26 subgroups
 - Sex, age, education, sector, region, hire type
- But could do for others:
 - Ethnicity, marital status, occupation or industry groups
- In each case, calculate level and index of wage, volume, and wage bill
- For age, also calculate per-capita volume

NHQI: Per-capita volume, by age (2005=100)

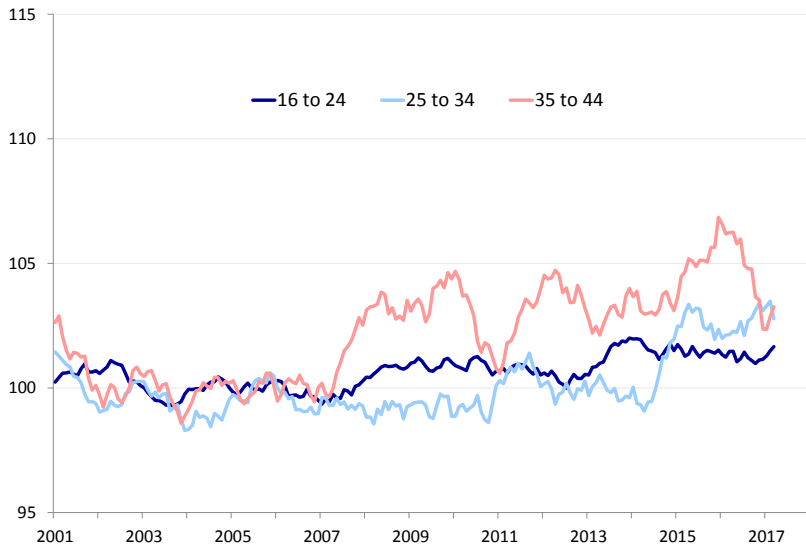


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Per-capita volume is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

NHQI: Index, by age (2005=100)

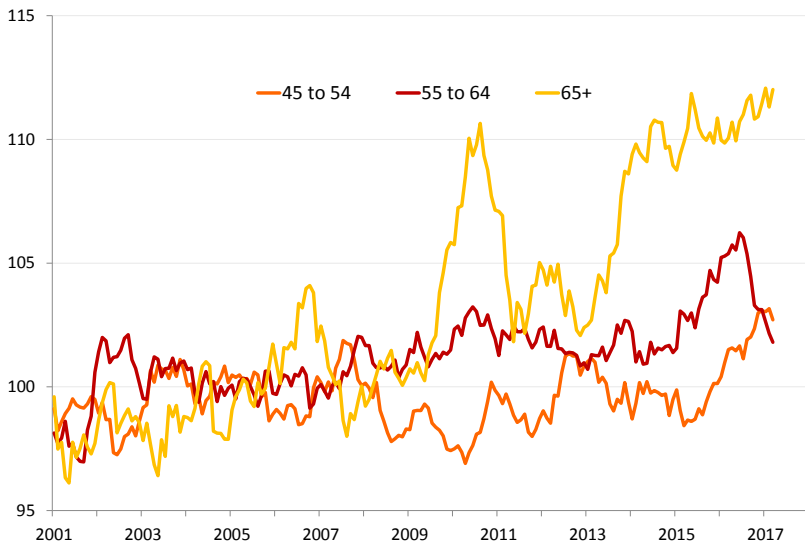


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data



NHQI: Index, by age (2005=100)

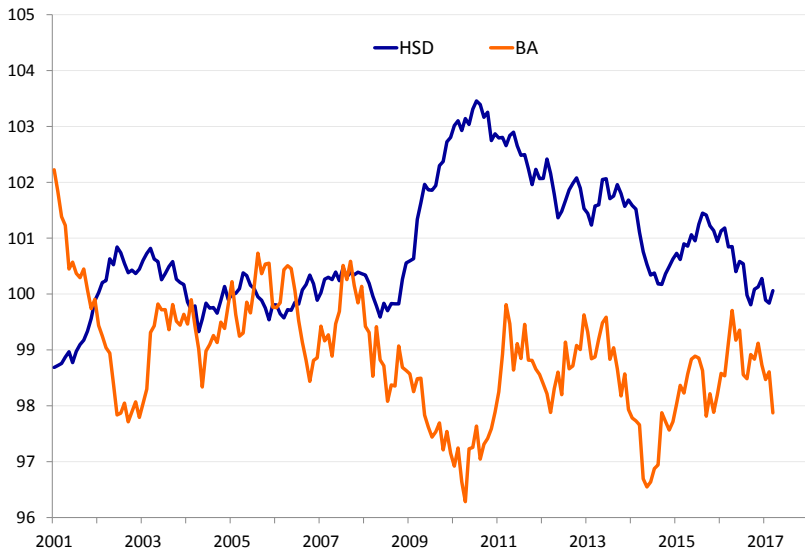


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

NHQI: Index, by education (2005=100)

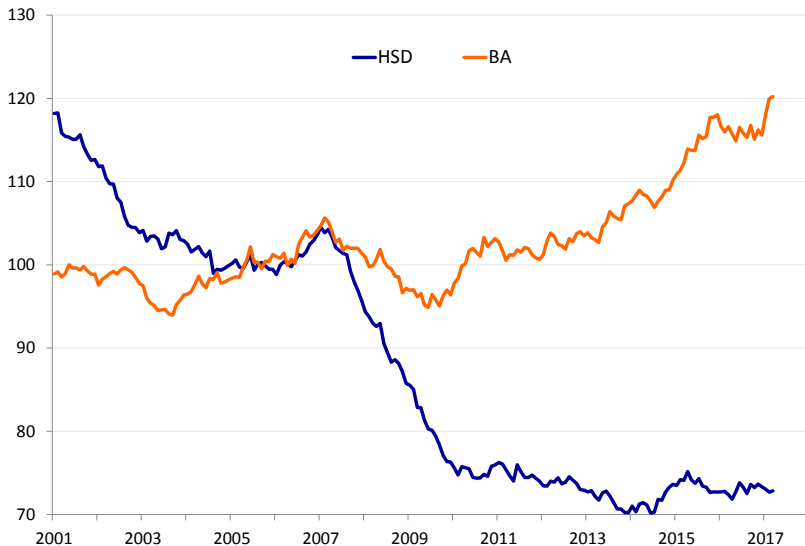


SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Wage index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

NHQI: Volume index, by education (2005=100)



SOURCE: Upjohn Institute New Hires Quality Index

NOTE: Volume index is based on a 12-month lagged moving average of monthly data

WE UPJOHN INSTITUTE
FOR EMPLOYMENT RESEARCH

Conclusion

- Monthly index of new hires is possible with CPS
- Can easily create metrics for volume, overall and for subgroups
- Hourly wage is also possible, with more caveats
 - OES-occupation wages allow matches for all new hires per month (→ subgroup trends), but miss within-occupation changes and available only from 2001
 - Self-reported wages capture total wage change and available longer, but smaller sample sizes limit subgroups and stability
- Both measures show increases since 2005 and sharply since 2015, but demographics play a role