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New Hires Quality Index

Brad J. Hershbein W.E. Upjohn Institute for Employment Research, hershbein@upjohn.org

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A Quality Index for New Job Hires

Brad Hershbein

W.E. Upjohn Institute for Employment Research

May 2017

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 - 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

- 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

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 - 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)

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- 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

Examples of New Hire "Job Quality" Lamp-post Inference

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

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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 "goodjobs" or "bad jobs" requires looking deeply into skill levels and job responsibilities

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Exactly. Why not do this?

• New monthly index tracks "quality" of new job hires (2001 ightarrow)

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 - 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

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Summary of findings

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Hershbein New Hires Quality Index

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New Hires Quality Index: Monthly Volume



Hershbein New Hires Quality Index

New Hires Quality Index: Monthly Wage Bill



Hershbein New Hires Quality Index

New Hires Quality Index: Hires per capita



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New Hires Quality Index 18/ 66

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New Hires Quality Index: Women and Men



Hershbein New Hires Quality Index

New Hires Quality Index: Women and Men, volume



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New Hires Quality Index: Volume by education



New Hires Quality Index: Wage bill by education



Hershbein

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- In 2005, college graduates accounted for one-fifth of all hires; in 2016, they accounted for one-fourth
- 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




New Hires Quality Index: Volume by Hire Type



Hershbein New Hires Quality Index

- Methodology
- Ø Robustness
- O What about actual reported wages?
- Subgroups
- Onclusions

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- Still, will check SIPP(?) to gauge magnitude of new hires who change residences
 - Probably positively selected...

• How to identify new hires (excluding self-employed)?

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- Weighted aggregates compare reasonably well with JOLTS, but less cyclical
 - Conceptual differences, and JOLTS undercounts relative to QWI

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CPS New Hires Volume vs JOLTS





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- Simple correlation is 0.92, and some conceptual differences in samples (reference period, unpaid leave, informal work)

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- Need to harmonize occupation codes over time
 - $\bullet\,$ From 1994 through 2002 \rightarrow 1990 Census codes
 - $\bullet\,$ From 2003 through 2010 \rightarrow 2000 Census codes
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- The 2011 \rightarrow period is straightforward...

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- $\bullet~$ But need to map 2000 SOCS \rightarrow 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

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 - BLS-released 2000–2002 files with 2000 Census codes
- $\bullet\,$ Thus, focus on 2000 $\rightarrow\,$ period

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

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- Merge 25th percentile occupational wages using SOC to CPS new hires
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- Merge on 6-digit SOC by 2-digit NAICS
 - Hierarchical process; use coarser SOCs for unsuccessful matches

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Methodology: Demographic Adjustment

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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

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Process

• Calculate means, overall and for subgroups, each month

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- To smooth out noise and seasonals, take 12-month lagged moving average
 - Straightforward, intutive, 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



Hershbein New Hires Quality Index

New Hires Quality Index: Sample Size, by group



Hershbein New Hires Quality Index

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- To smooth out noise and seasonals, take 12-month lagged moving average
 - Straightforward, intutive, 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



Hershbein New Hires Quality Index

Robustness: Robust to Demographic Adjustment



Hershbein New Hires Quality Index

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





Robustness: Quantiles

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 - Captures only change in median occupation hired
 - Even with demo adjustment, misses rest of distribution
- Growth is concentrated in right-tail occupations

Robustness: Quantiles



Hershbein New Hires Quality Index

Robustness: Quantiles (Index: 2005=1)



Hershbein

New Hires Quality Index

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. W.E. UPJOHN INSTITUTE FOR EMPLOYMENT RESEARCH

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 - Omposition and selection: Imputation may cause valid wages to cover different population than all new hires
- But also conceptual difference: Xs vs. β s

- 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





NHQI and CPS self-reports (2005=1)



Hershbein New Hires Quality Index

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- Reduced sample sizes, when averaged, sufficient for index...
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 - ... 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

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- On most observables, seems small
 - $\bullet\,$ Valid-wage sample is younger, less Black, and more $E{\rightarrow}E$

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 - Insufficient predictors
- Can back out expected bias (from observables)
 - Run (valid) wage regression on ${\boldsymbol X}$ and adjust for $\Delta {\boldsymbol X}$
 - Results imply about 1% negative bias, mostly from age
 - Adding occupation and industry to ${f X}$ increases bias slightly, to 2.7%
 - Stable over time, for offsetting reasons

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NHQI heterogenity

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 - Sex, age, education, sector, region, hire type
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- 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)



NHQI: Index, by age (2005=100)



Hershbein New Hires Quality Index

NHQI: Index, by age (2005=100)



Hershbein New Hires Quality Index

NHQI: Index, by education (2005=100)



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New Hires Quality Index

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



Hershbein New Hires Quality Index

Conclusion

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 - 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