The Rise in Returns to Skill? A Modern Regression Analysis of Wage Inequality in the Current Population Survey (CPS)

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Overview

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 - Decomposition
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- What is driving this rise in inequality?
- Can non-standard statistical methods explain rising inequality?

Figure: Indexed Real Hourly Wage by Percentile, 1980-2016.

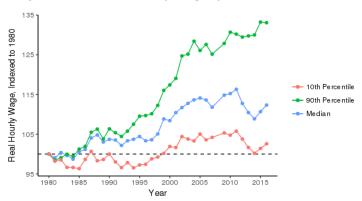
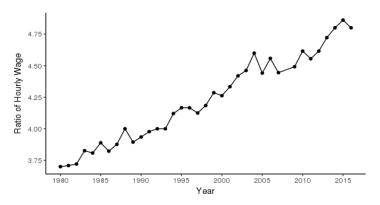


Figure: Ratio of Wage Between 90th and 10th Percentiles, 1980-2016.



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- Attribute most of the rise in inequality to 'rise in return to skill

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- Expose the decomposition's sensitivity to specific prediction methods used

Data - March CPS

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Current Population Survey

- Current Population Survey (CPS)
 - Serves as the nation's primary source of statistics on labor force characteristics.
 - The Annual Social and Economic Supplement provides the official annual statistics on the nation's poverty levels as well as statistics on income, health insurance coverage, marital status, educational attainment, employee benefits, work schedules, school enrollment, noncash benefits and migration.

CPS Table Creator

The CPS Table Creator gives you the ability to create customized tables from the Current Population Survey's Annual Social and Economic Supplement (CPS ASEC).



Standard linear model for predicting a measure of wages individual years:

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Note: $\bar{\beta} \neq \beta_t$, for $t = 1980, \dots, 2016$

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 (predicted income distribution, under fixed returns)
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$$(\text{observed income distribution}) \\ Y_{it} = \\ (\text{predicted income distribution, under fixed returns}) \\ \mathbf{X}_{it}\bar{\beta} \\ + \\ (\text{difference to predicted income distribution, under variable returns}) \\ \mathbf{X}_{it}(\beta_t - \bar{\beta}) \\ + \\ (\text{residuals, unexplained factors}) \\ \varepsilon_{it}$$

So that the distribution of income is the sum of three components:

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- \Rightarrow This approach relies heavily on the variables in \mathbf{X}_{it} or even prediction method used.
- \Rightarrow Perhaps we could use more than OLS?

Prediction Methods

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Mincer Wage equation, dating to Mincer (1954, 1972):

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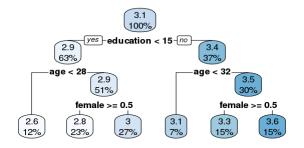
Adjusted Mincer wage equation, proposed by Lemieux (2006):

$$Y_{it} = Y_0 + \rho_{1t} s_{it} + \rho_{2t} s_{it}^2 + \beta_{1t} x_{it} + \beta_{2t} x_{it}^2 + \beta_{3t} x_{it}^3 + \beta_{4t} x_{it}^4 + \varepsilon_{it}$$
 (5)

Novel prediction methods, tree-based methods.

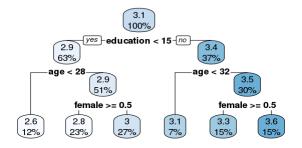
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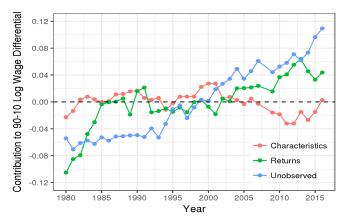


Random forest prediction uses random sampling and building a forest of decision trees to avoid over-fitting.

Results - Mincer Wage Equation

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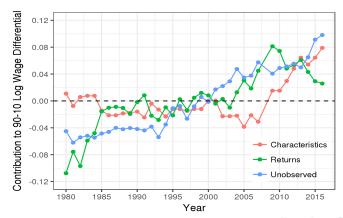
Figure: Components of 90-10 Percentile Log Wage Differential.



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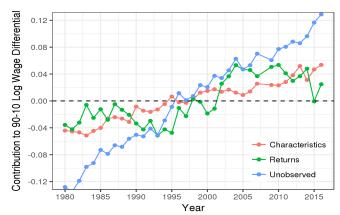
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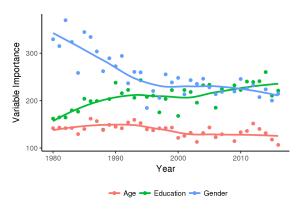
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Results - Variable Importance

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Figure: Variable Importance in Random Forest Prediction, 1980–2016



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 Education is more and more important in random forest prediction of wages, while gender's role reduces.

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Why just random forests?

Overview of Results
Discussion and Limitations

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Any questions?