
What Do We Really Know About Changes in Wage Inequality?

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Motivation

- Very well known that wage inequality has grown substantially since the 1970s
 - Series of influential papers published in the early 1990s, in particular Katz and Murphy (1992) and Juhn, Murphy and Pierce (1993) laid down the main facts and possible explanations.
 - What have been the main changes in wage inequality since then?
 - What does this tell us about possible explanations for inequality changes?
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Some specific issues addressed here

- Measurement issues, do they matter?
 - March vs. MORG CPS
 - Growing importance of wage allocation (35% in latest MORG)
 - Top-coding
 - Explanations: Demand and supply and beyond
 - Secular changes in the nature of relative demand
 - Wage setting institutions (minimum wage, unions, performance pay)
 - What is going on at the very top end (rent-extraction, social norms, market of executives, etc.)
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Data issues

- I mostly focus on the May/MORG instead of the March CPS for a number of reasons
 - Larger samples (12 rotation groups vs. 4 plus oversamples)
 - Direct measure of hourly wages for workers paid by the hour (over half of workforce)
 - Cannot go back as far (1960s in March vs 1973 in May) but less important with the passage of time
 - Union status available in most years
 - May-ORG vs. March mostly matters of within-group inequality
 - Does not increase in the 1970s
 - Plays less of a role in the overall inequality growth
 - But all other findings highly robust
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Data choices for most of the paper

- Use May (1973-78) and ORG (1979-2006) supplements of the CPS
 - Hourly wage rate: direct measure for hourly workers, earnings divided by hours for others
 - No wage allocation in 1973-78, throw out allocators in ORG to be consistent (but lose 1994 and 8 months of 1995)
 - 1.4 adjustment for top-coded observations
 - Weight by hours of work times CPS weight
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Measurement model

- Cut the data in 6 education groups and 22 two-years experience cells (separately for men and women)
 - Can control for composition effects by holding shares in each cell at its average value for the whole 1973-2006 period
 - Within (dispersion around cell mean) and between (dispersion across cell means) straightforward to compute
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Computing Wage Differentials

- Hold composition constant when computing education and experience wage differentials (and between-group variance, a relevant summary measure)
 - Education differentials: Hold experience distribution constant to its average 1973-2006 value (for all education groups)
 - Experience differentials: Hold education distribution constant to its average 1973-2006 value (for all experience groups)
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Price and Quantity Effects

- Quantity effects estimated by comparing actual dispersion to dispersion with constant shares of workforce in each cell (reweighted)
 - (Observable) price effects obtained by setting mean cell wage to average 1973-2006 level
 - Natural procedure when looking at the variance, not completely clear what to do with percentiles
 - Similar in spirit to Juhn, Murphy and Pierce decomposition
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Descriptive Facts

- Variance: within and between decomposition
 - Wage differentials
 - Within-group variance by education and experience
 - Beyond the variance: analysis by percentiles and JMP decomposition for 50-10 and 90-50 gap
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Figure 1a: Total Variance, Men

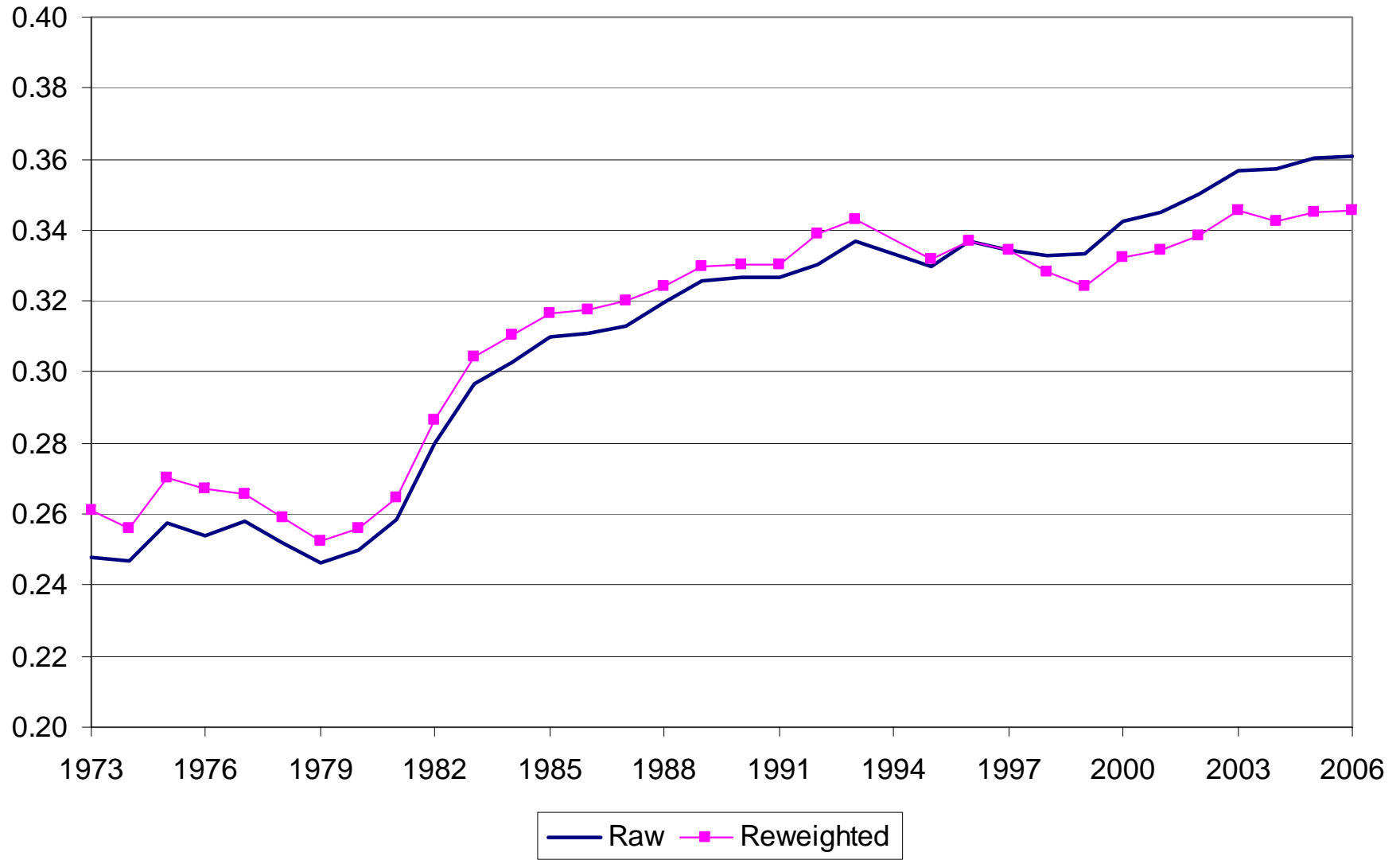


Figure 1b: Total Variance, Women

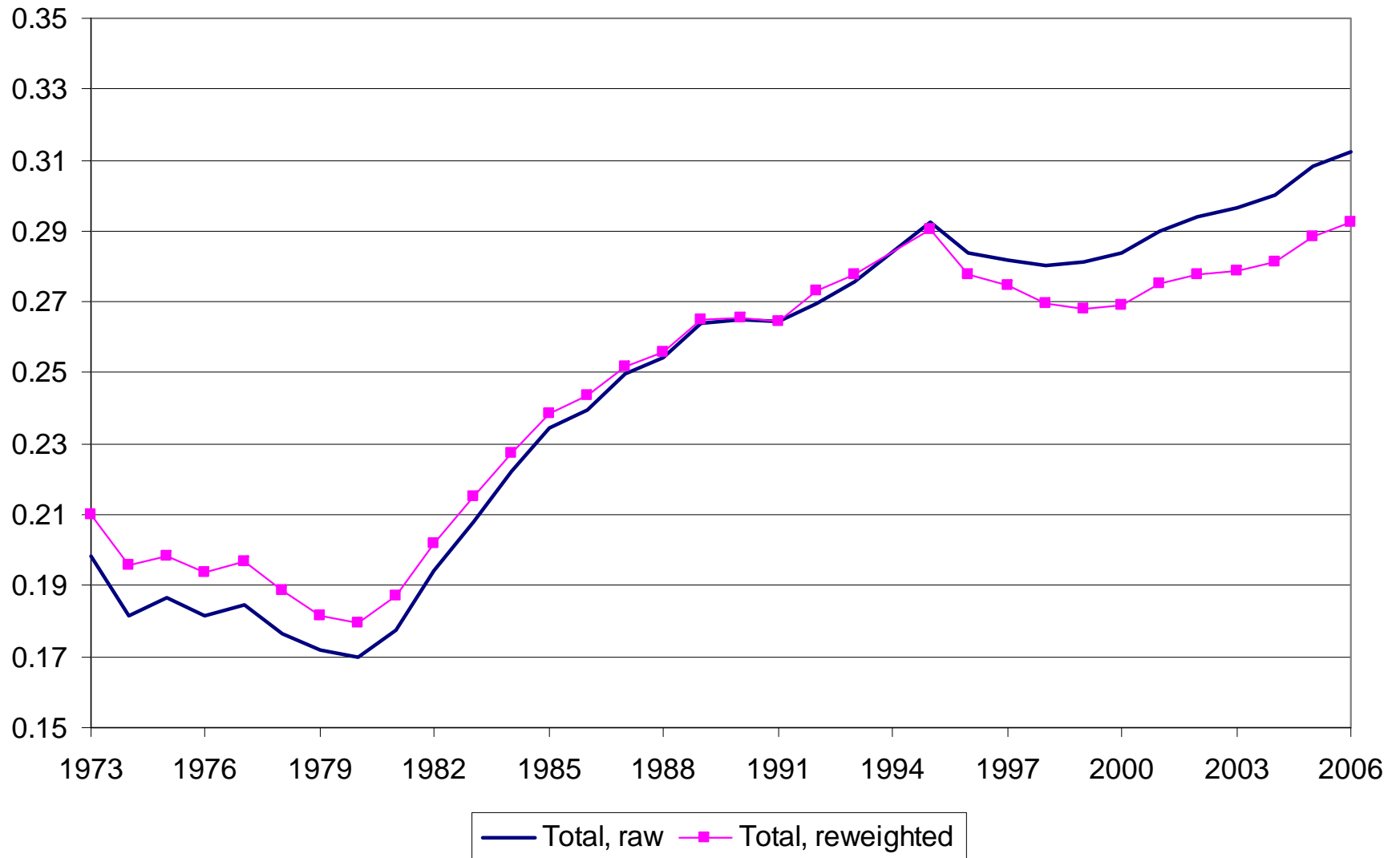


Figure 2a: Within- and Between-group Variances, Men

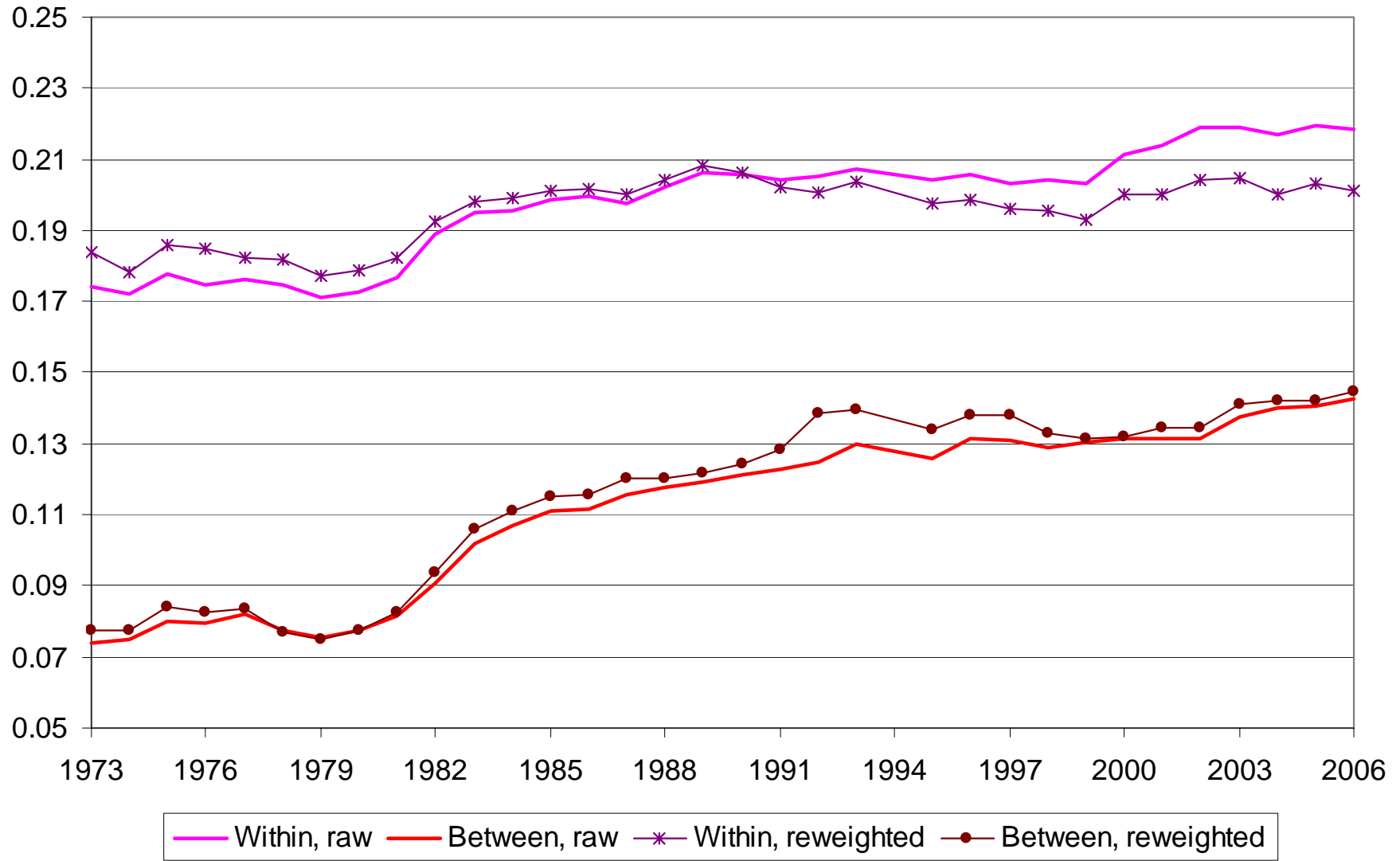


Figure 2b: Within- and Between-group Variances, Women

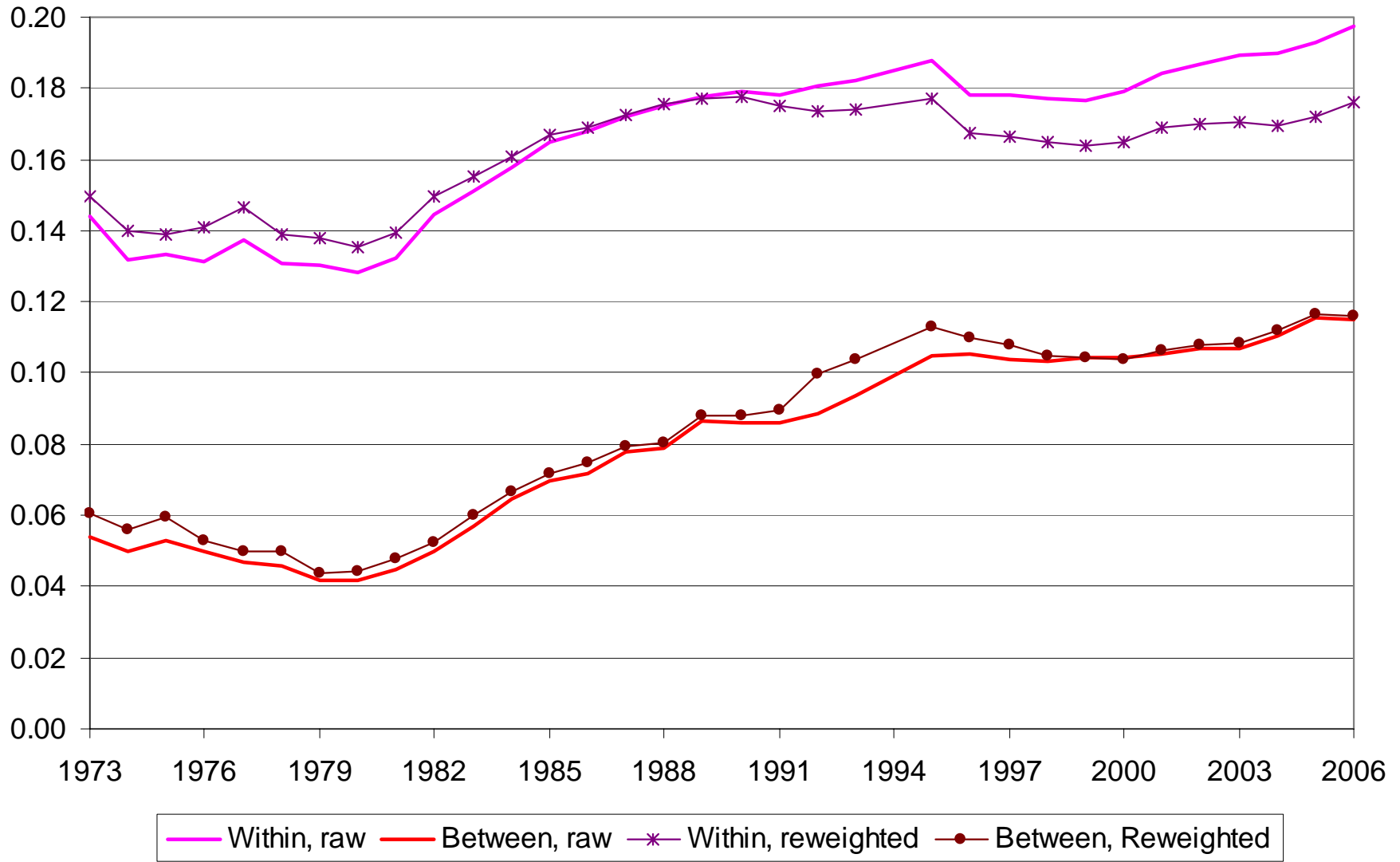


Figure 3a: Education Wage Differentials (Relative to High School Graduates), Men

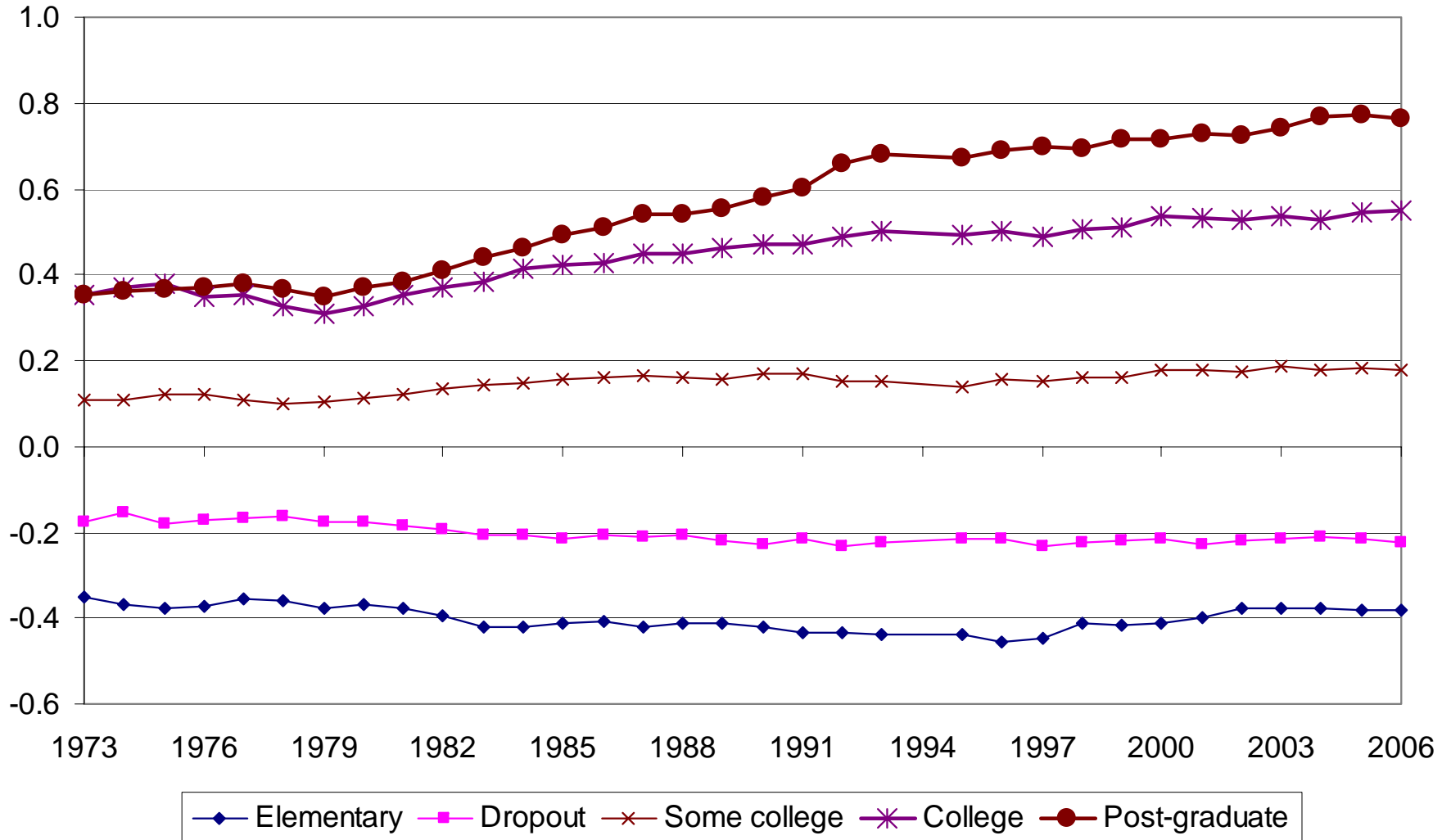


Figure 3b: Education Wage Differentials (Relative to High School Graduates), Women

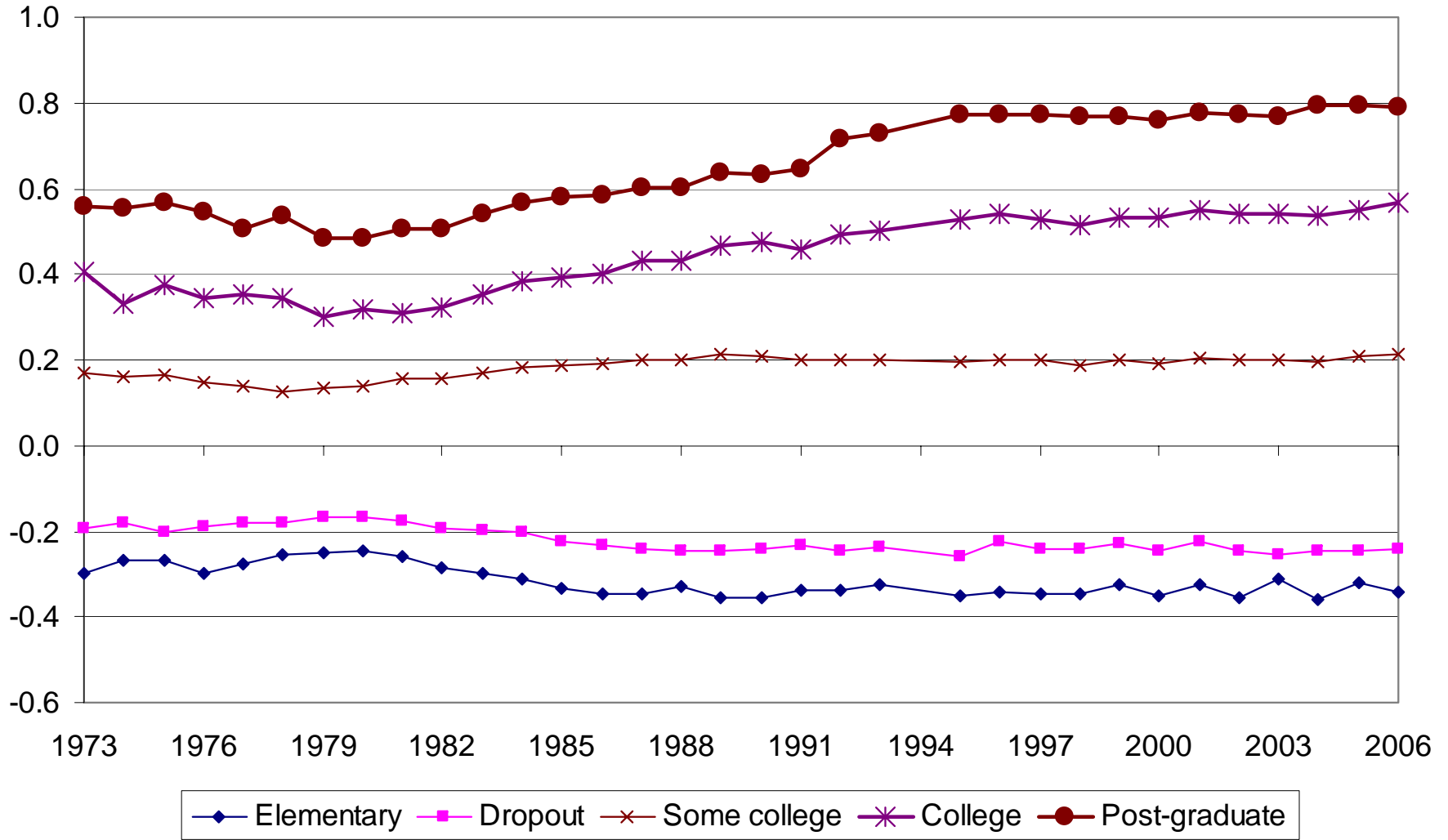


Figure 4a: Experience Wages Differentials (Relative to 20-29 Years of Experience), Men

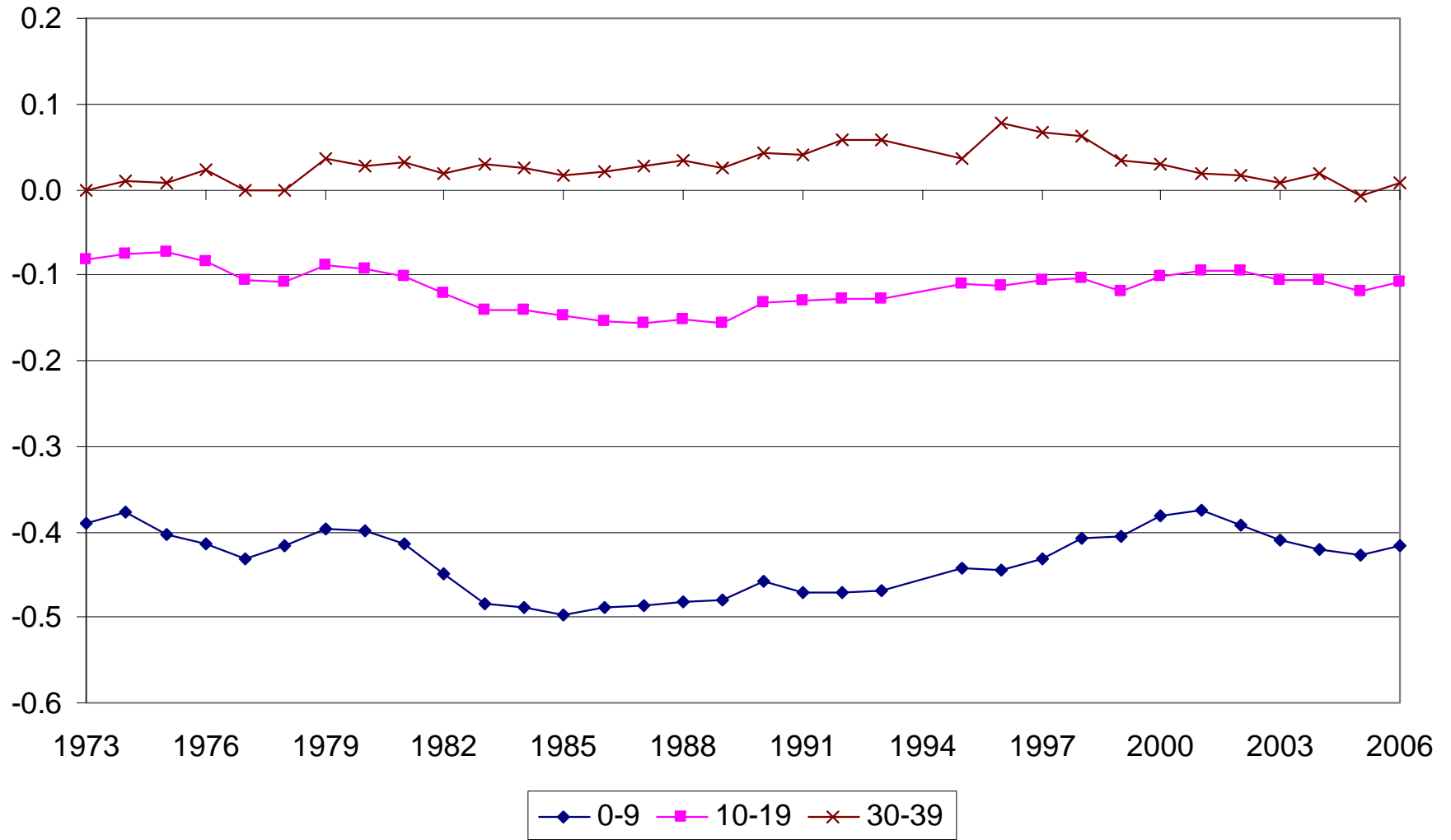


Figure 4b: Experience Wage Differentials (Relative to 20-29 Years of Experience), Women

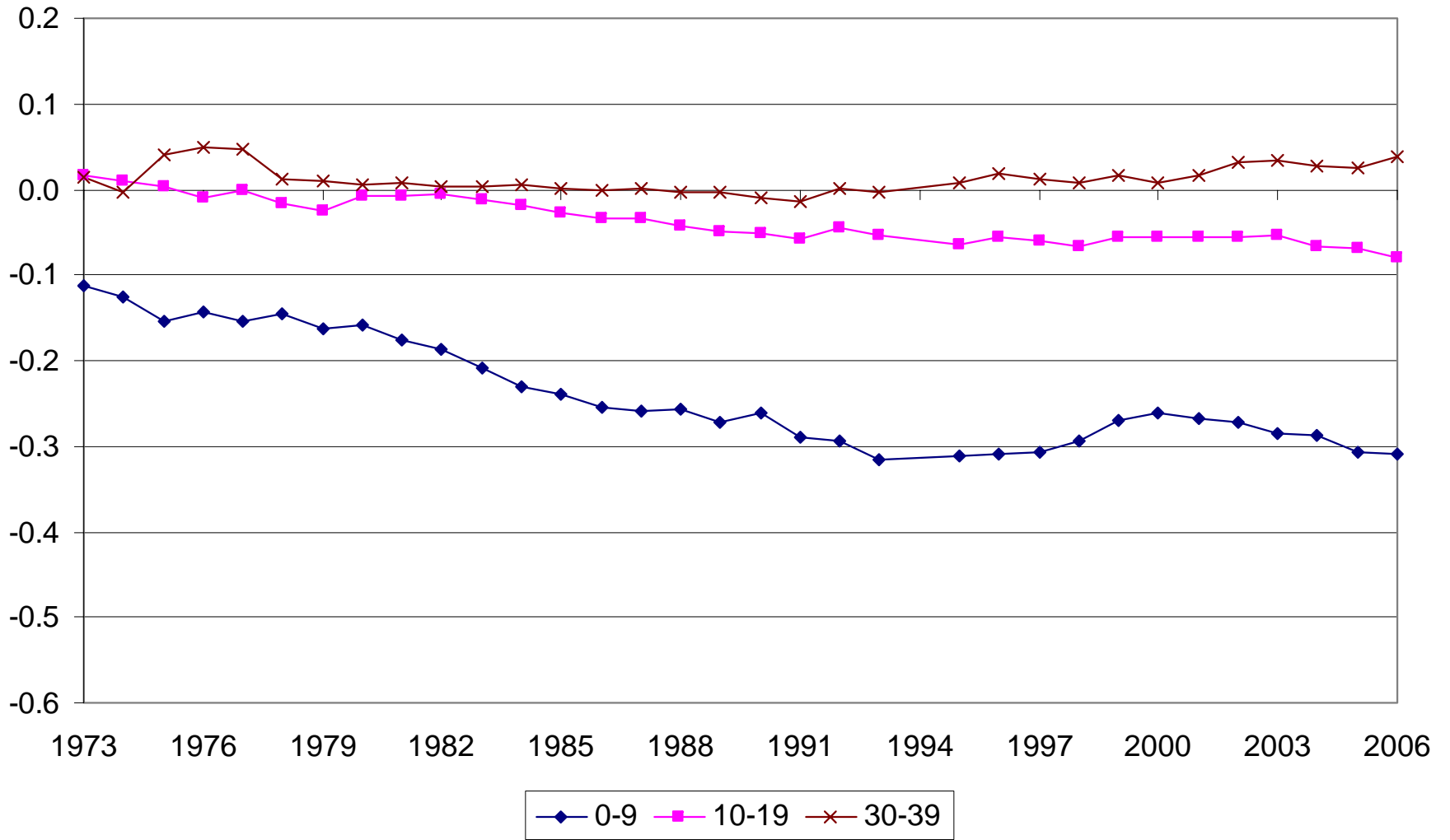


Figure 5a: Within-Group Variance by Education Groups, Men

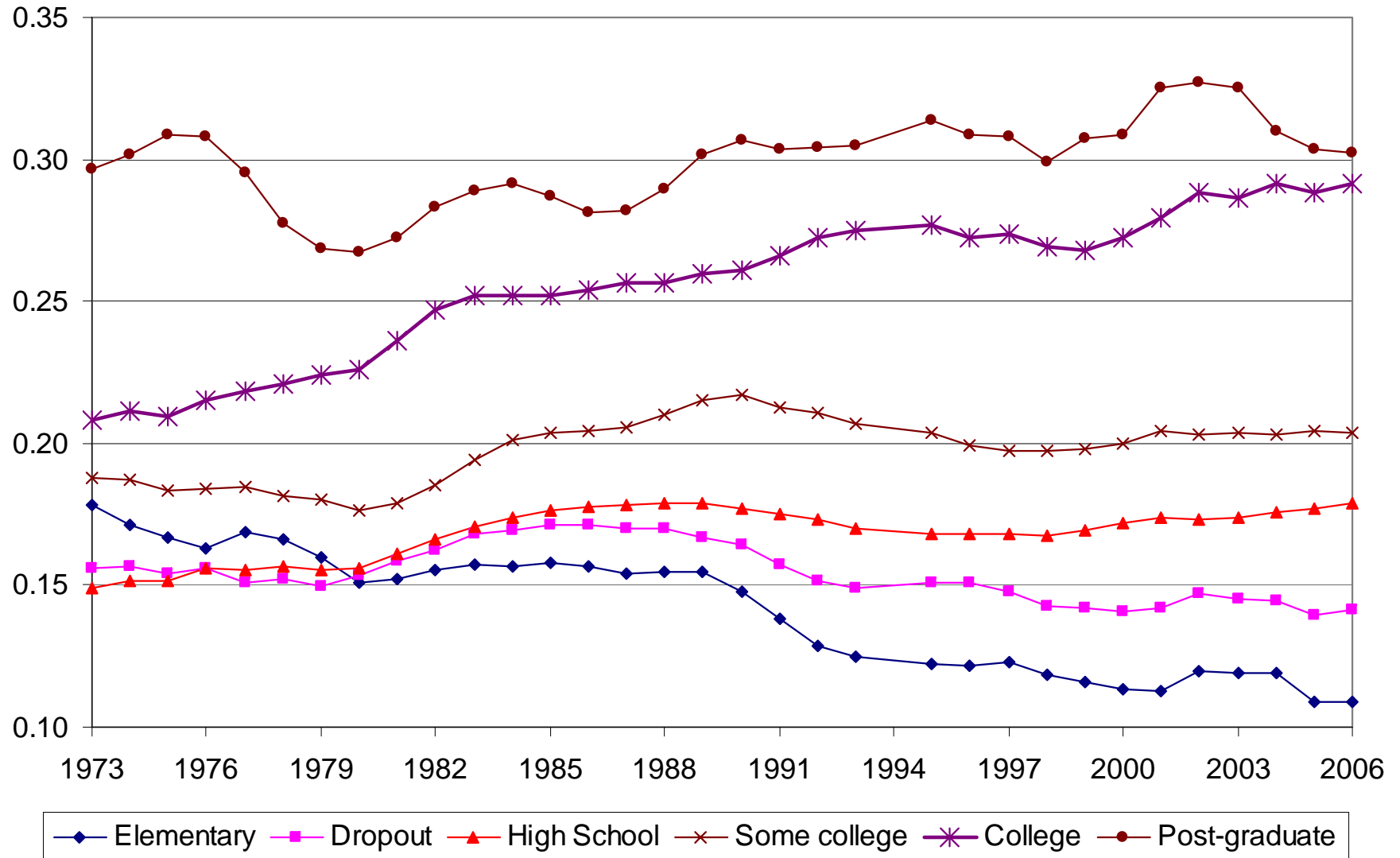


Figure 5b: Within-Group Variance by Education Groups, Women

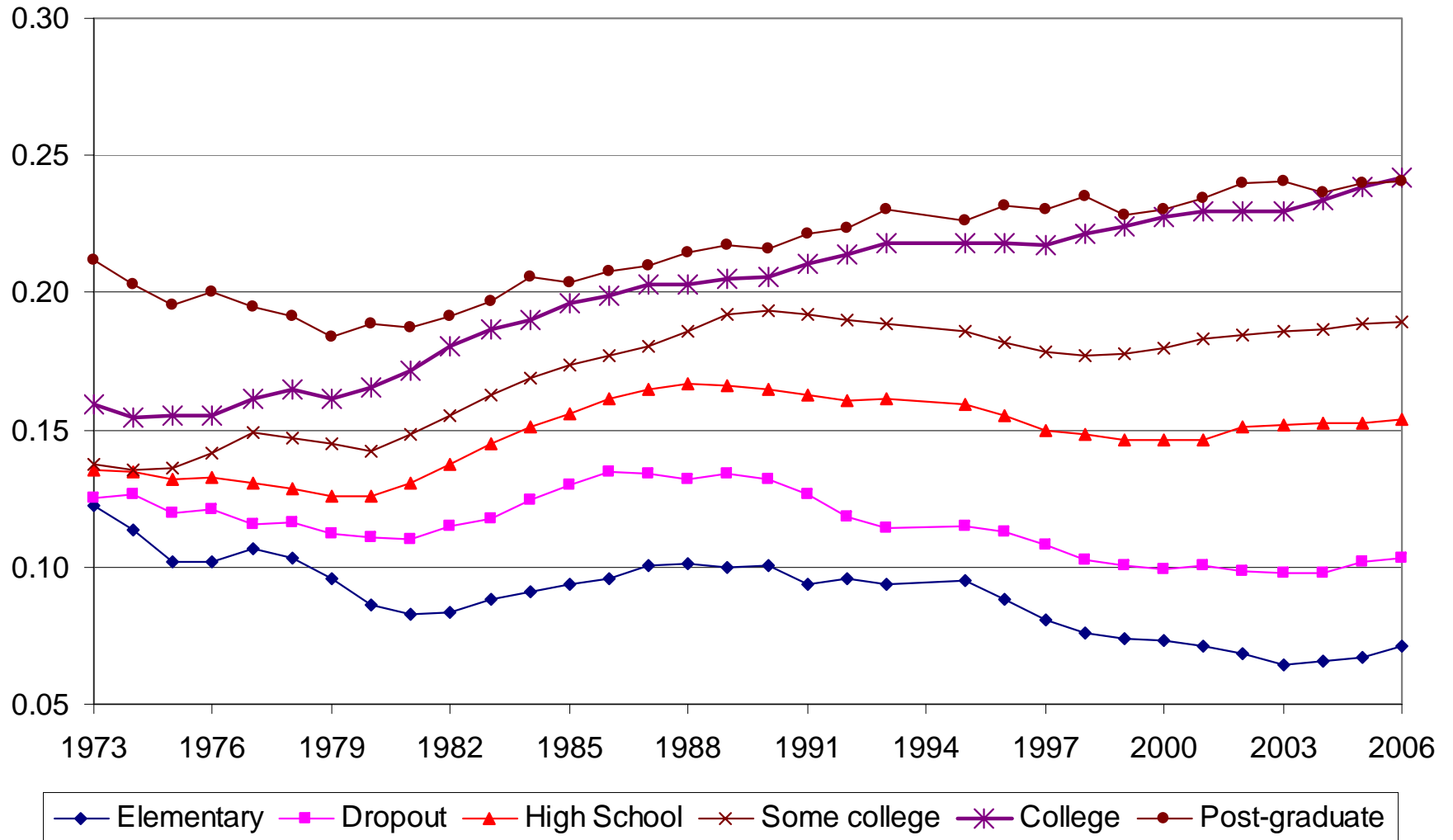


Figure 6a: Variance by Experience Groups, Men

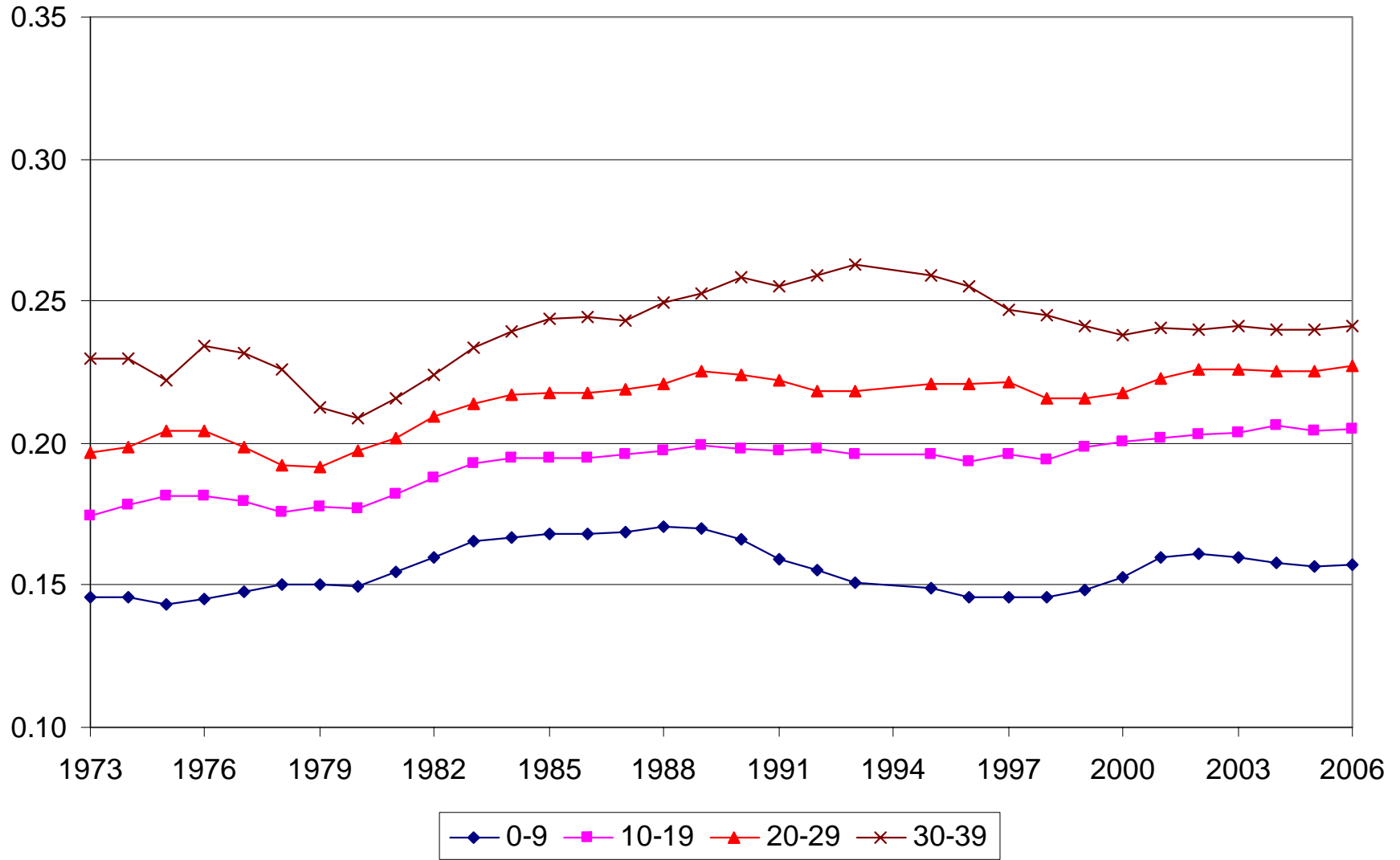


Figure 6b: Variance by Experience Groups, Women

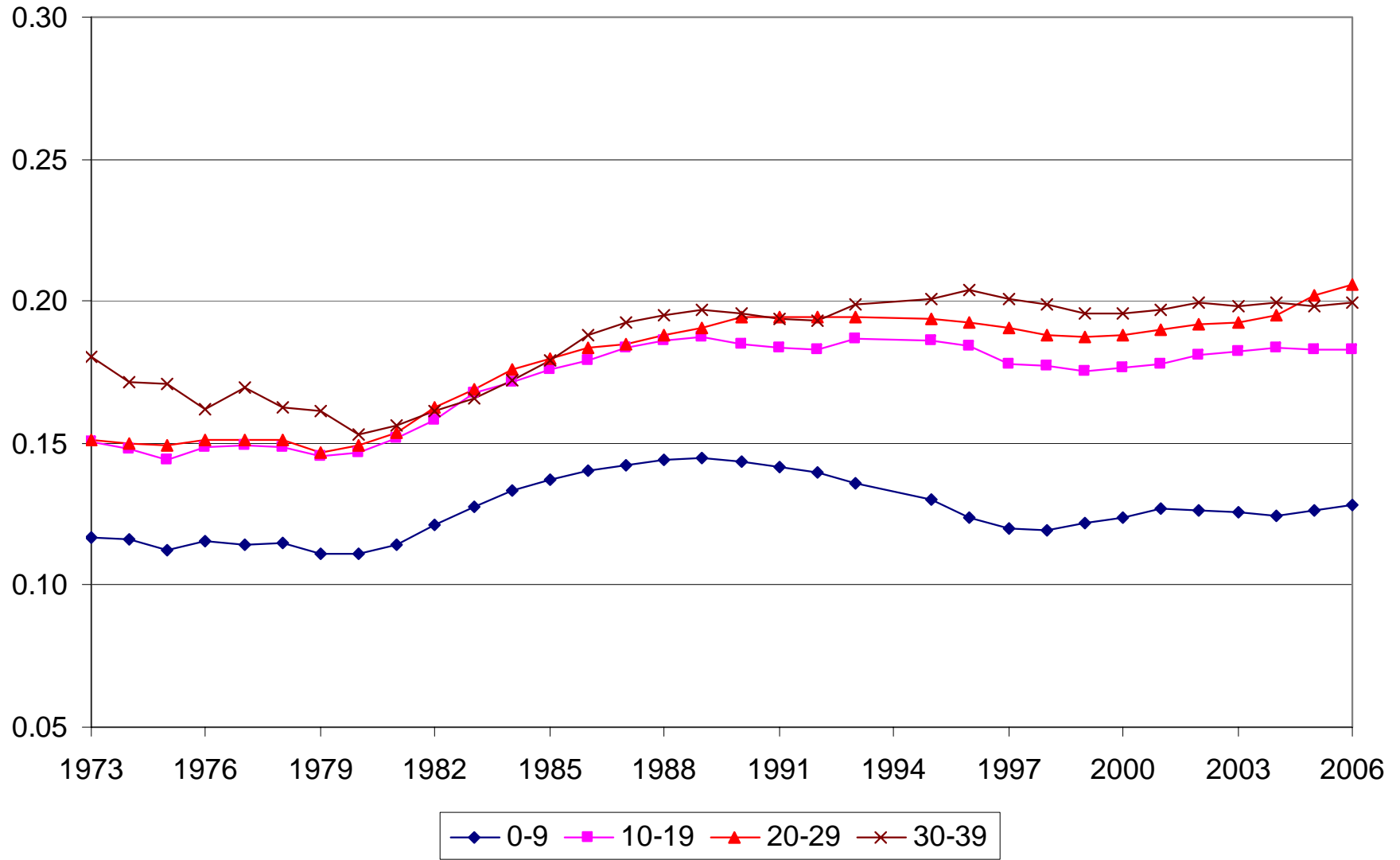


Figure 7a: Change in Real Wages by Percentile, Men

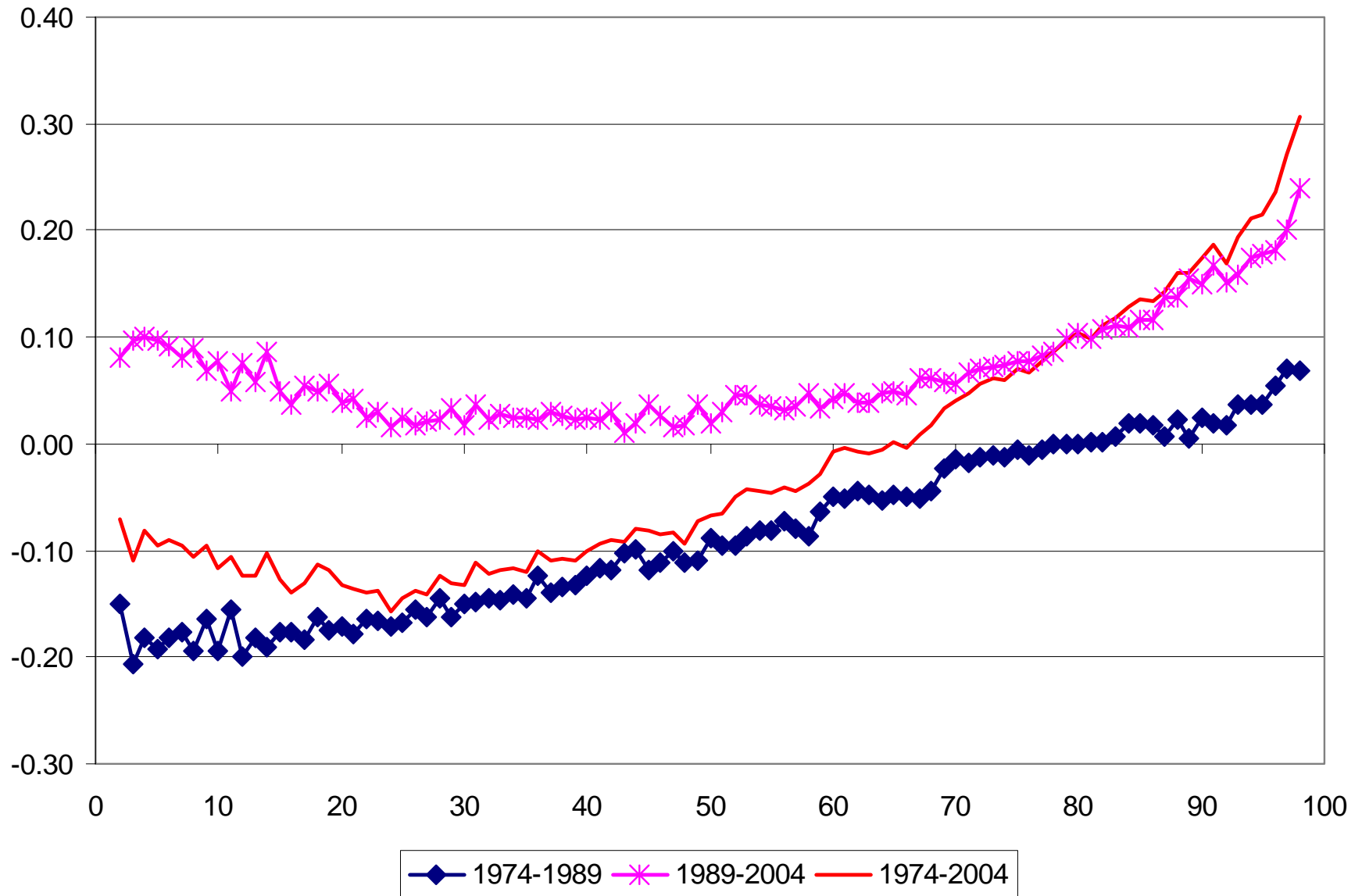


Figure 7b: Change in Real Wages by Percentile, Women

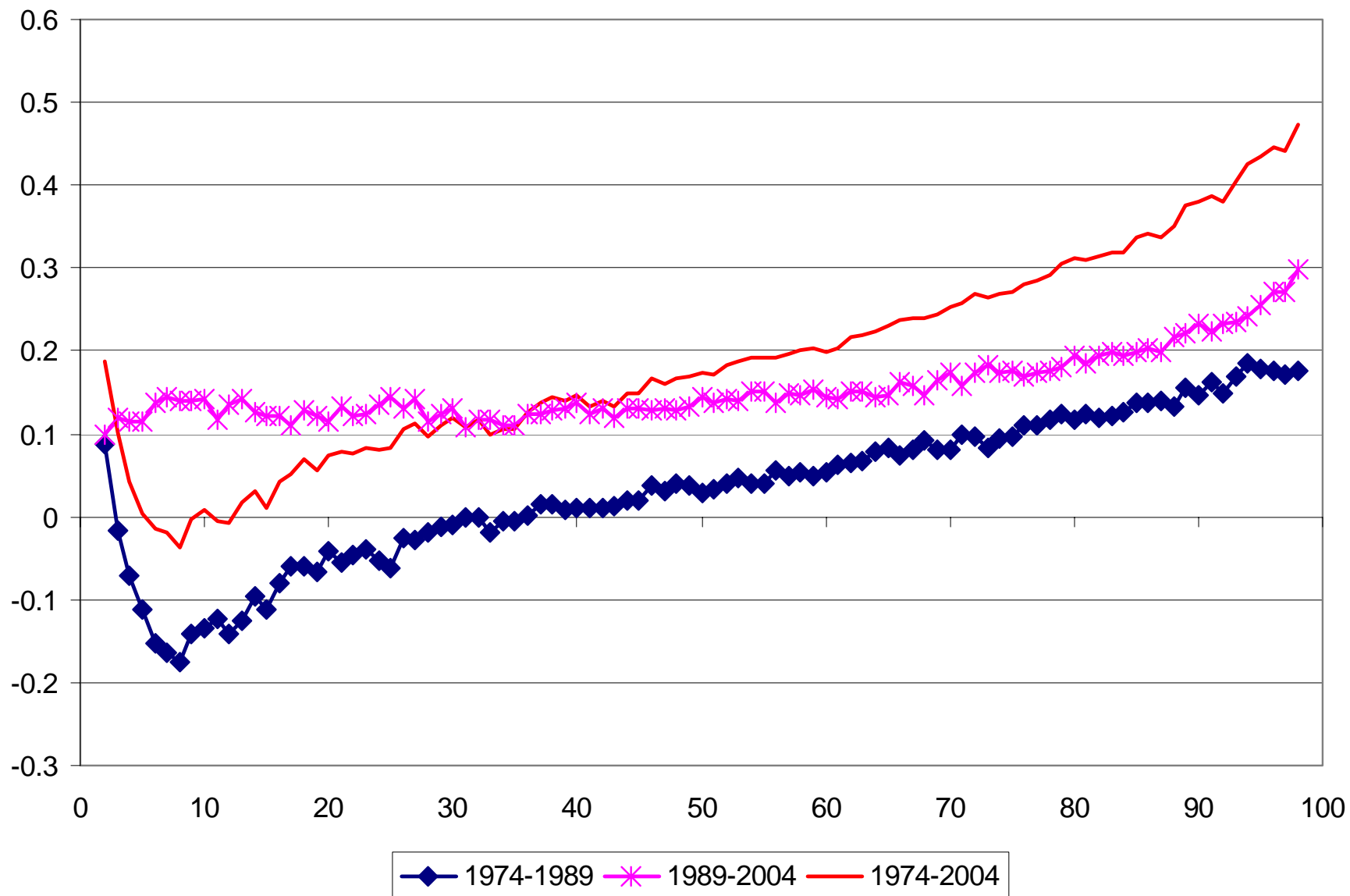


Figure 8a: Change in Wages Residuals by Percentile, Men

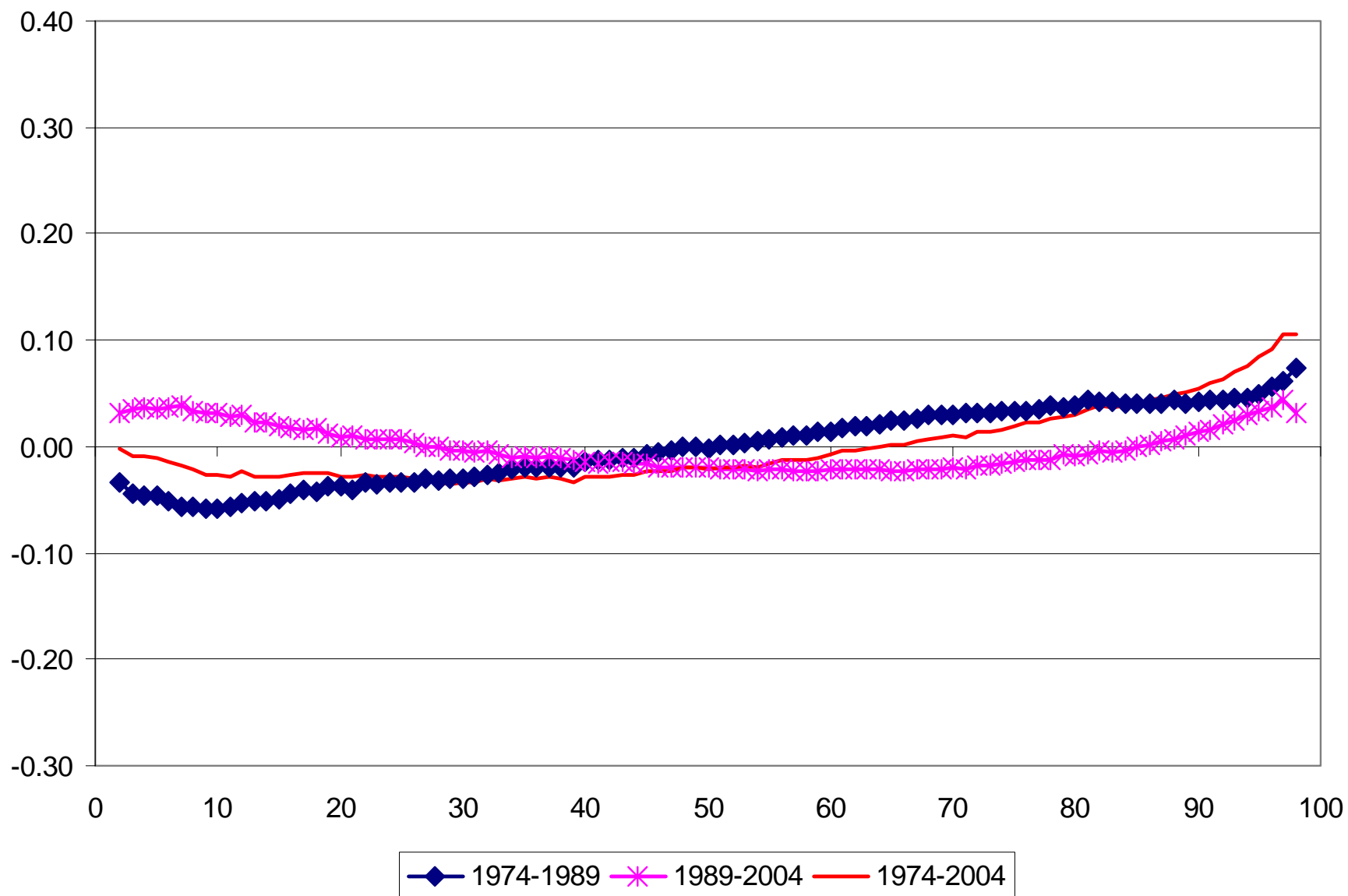


Figure 8b: Change in Wages Residuals by Percentile, Women

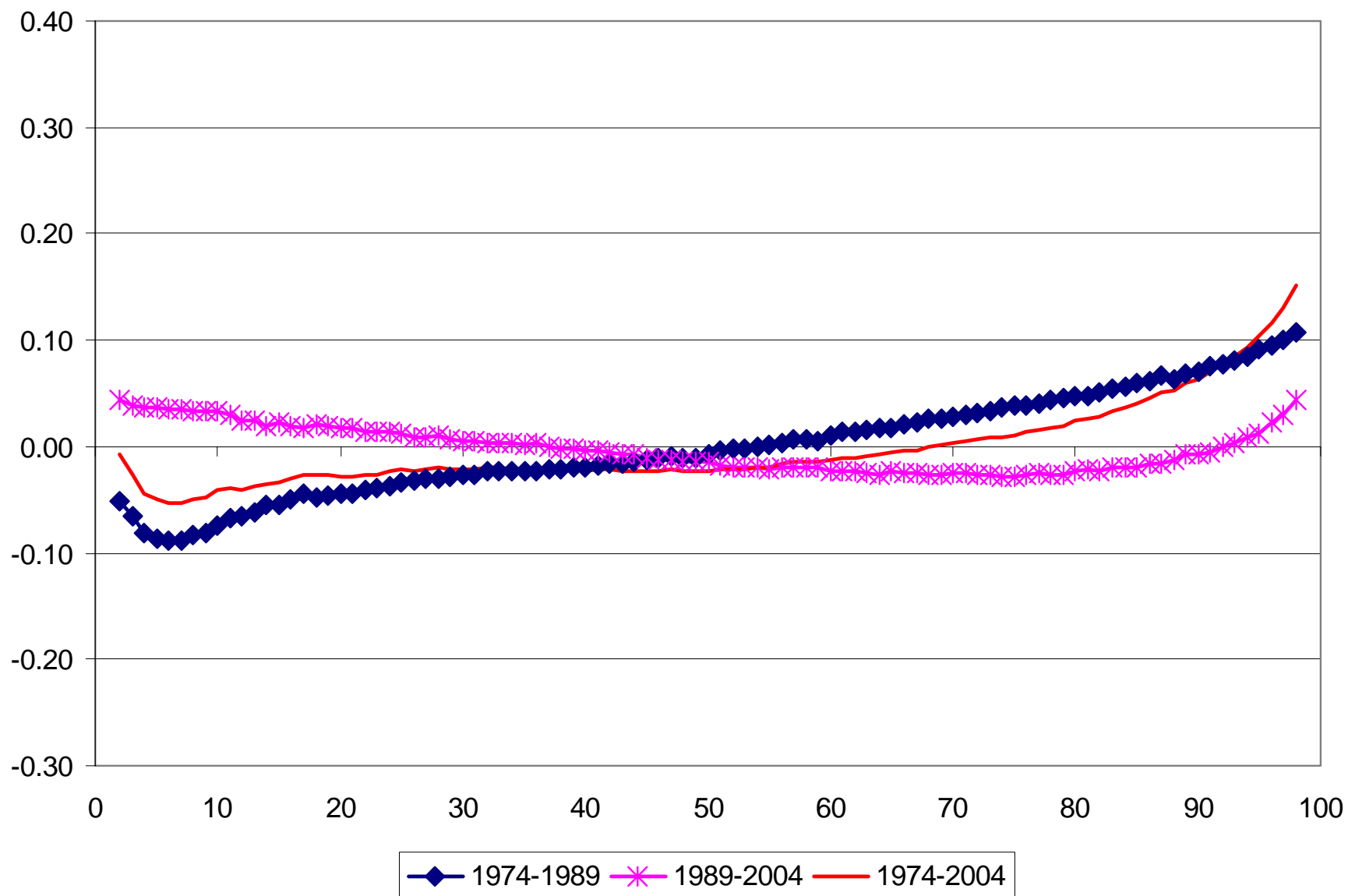


Figure 9a: Decomposition of Changes in 90-50 Gap, Men

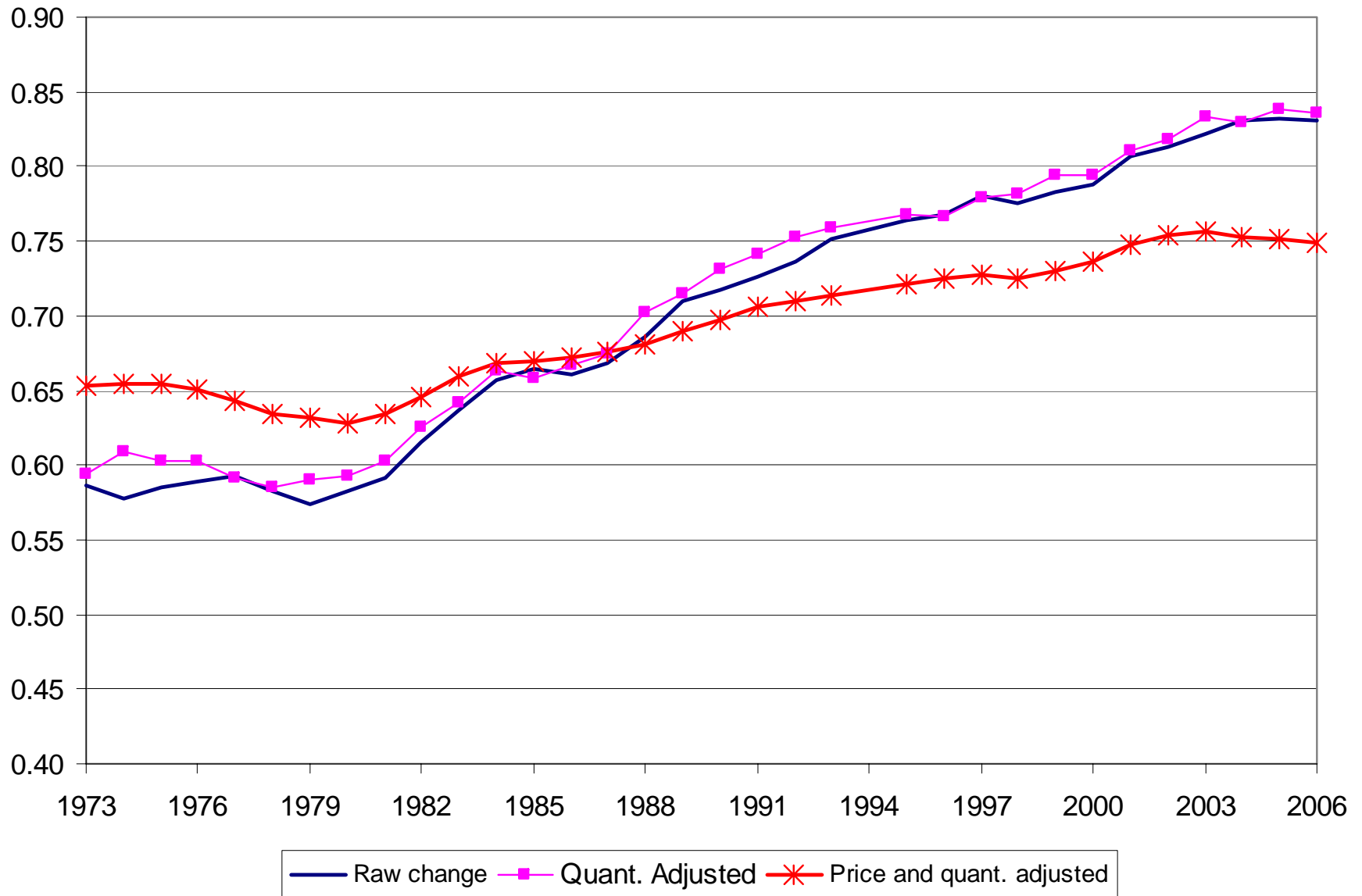


Figure 9b: Decomposition of Changes in the 50-10 Gap, Men

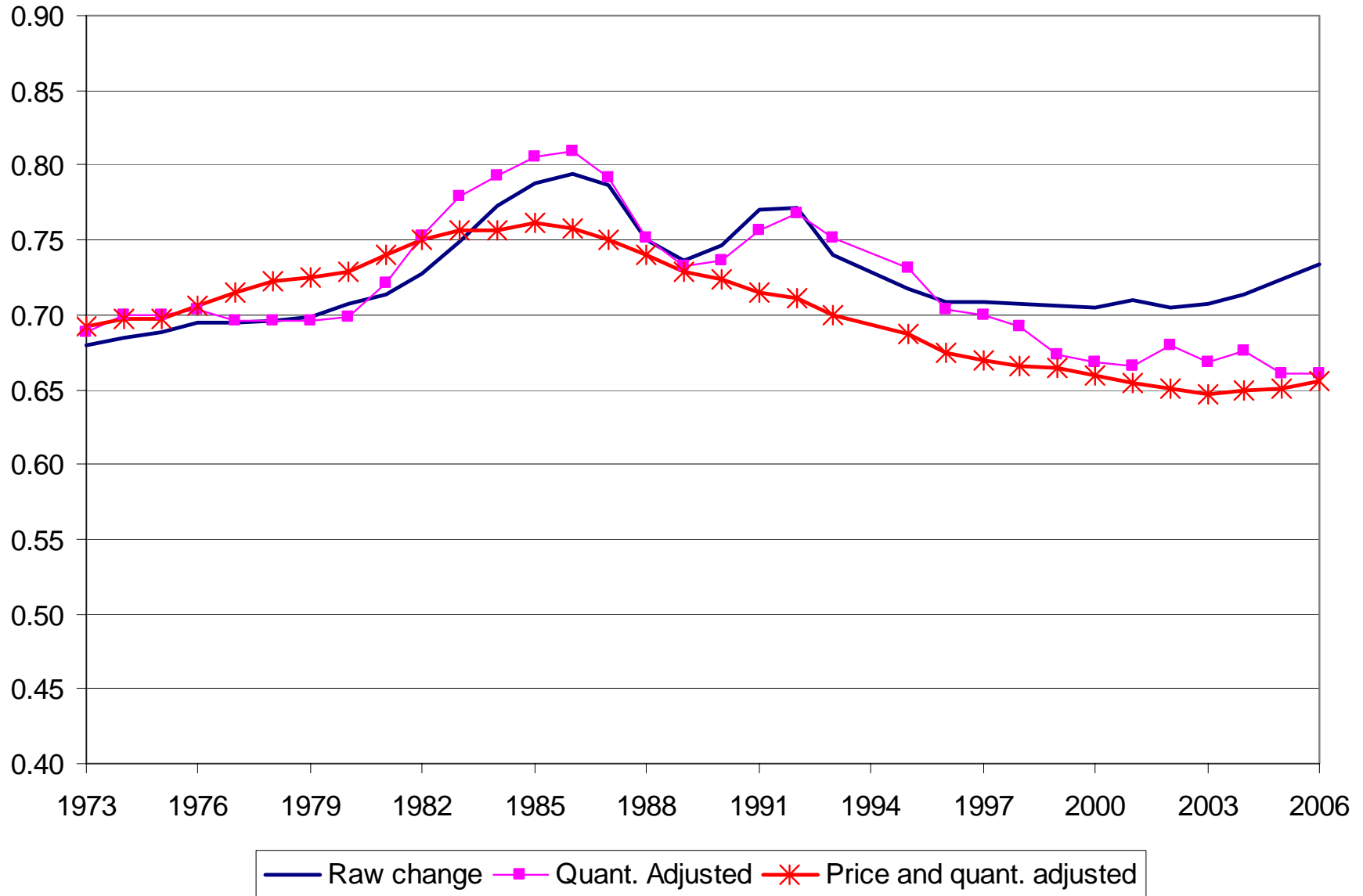


Figure 9c: Decomposition of Changes in 90-50 Gap, Women

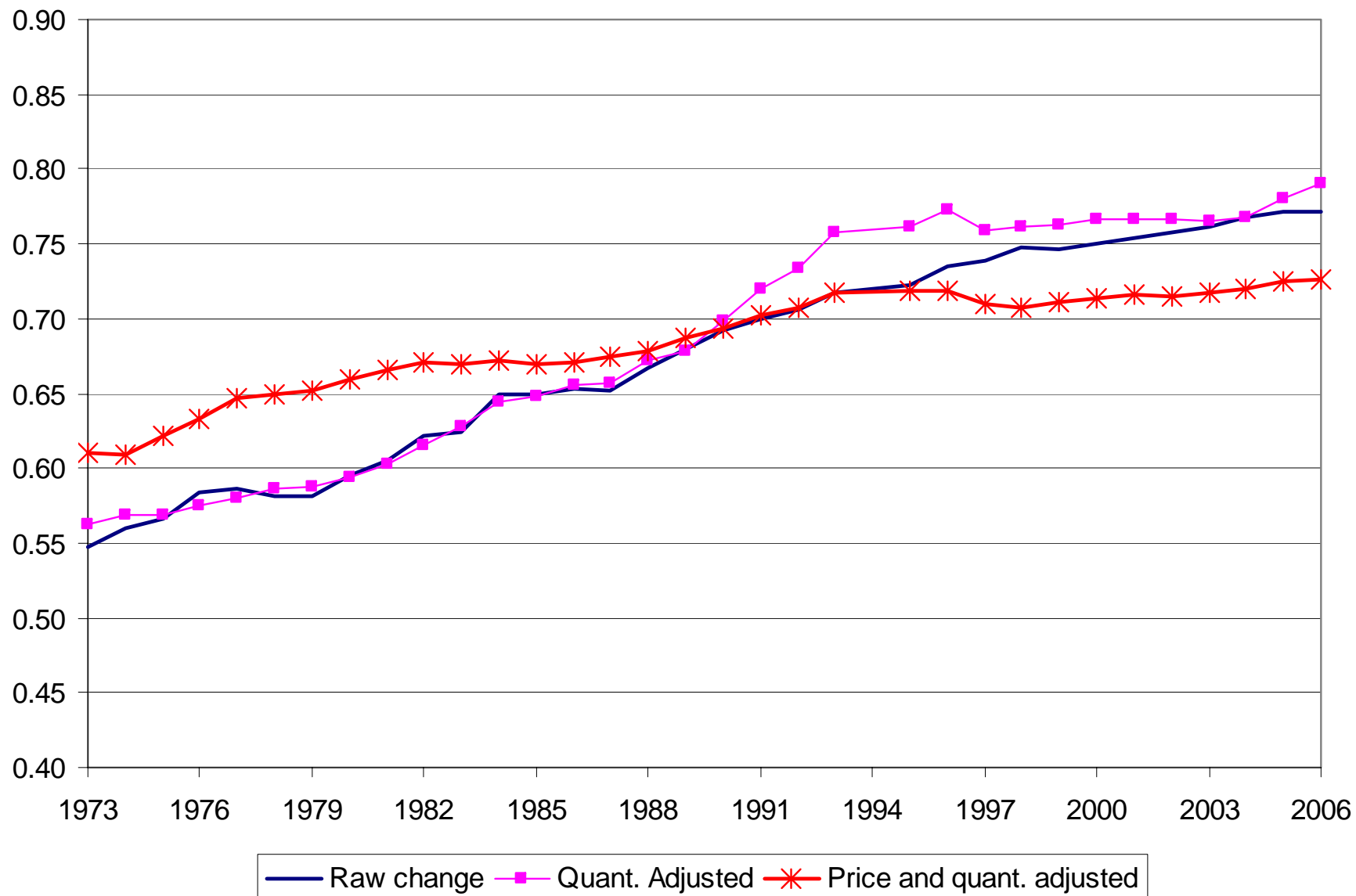
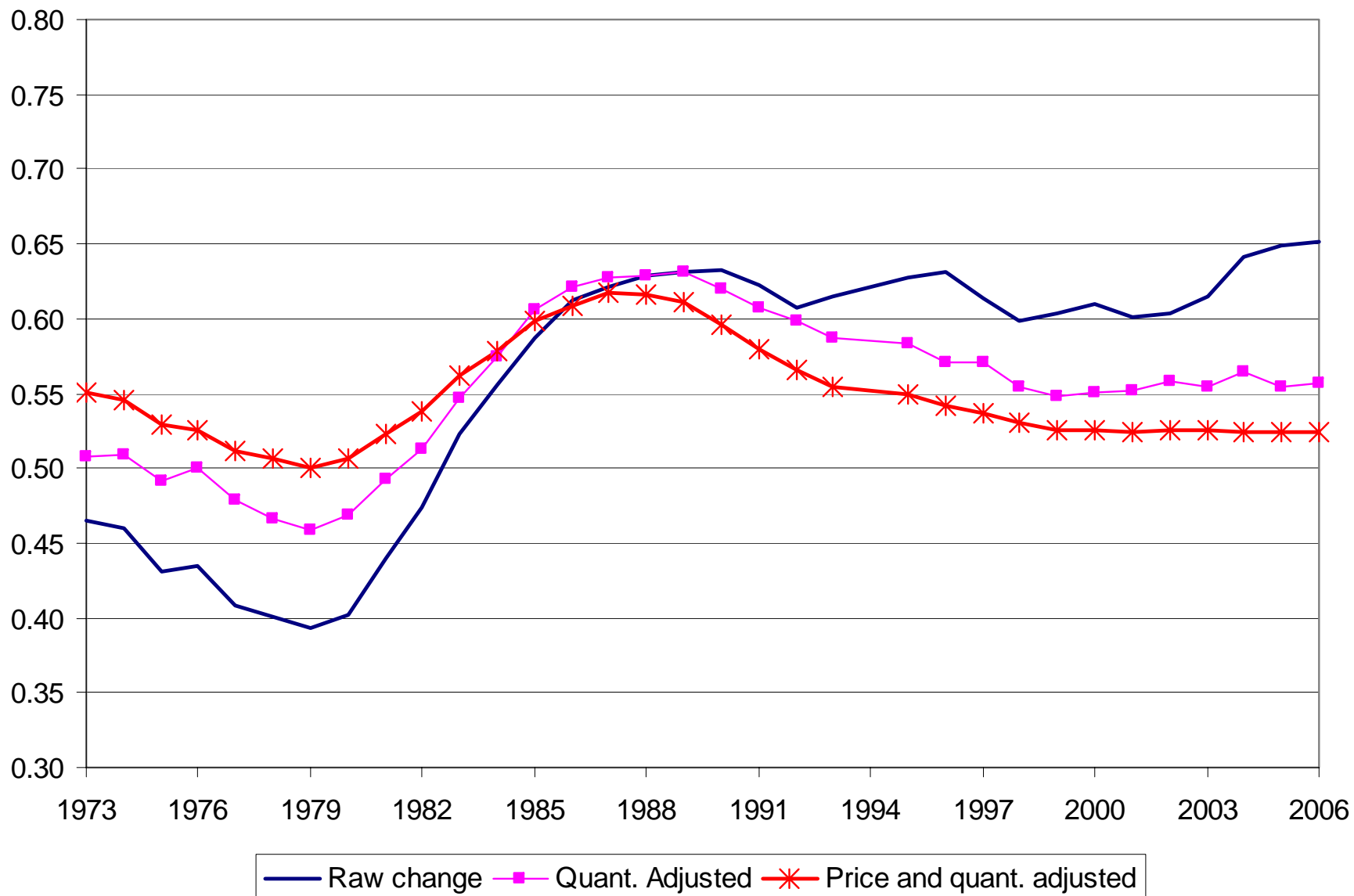


Figure 9d: Decomposition of Changes in 50-10 Gap, Women



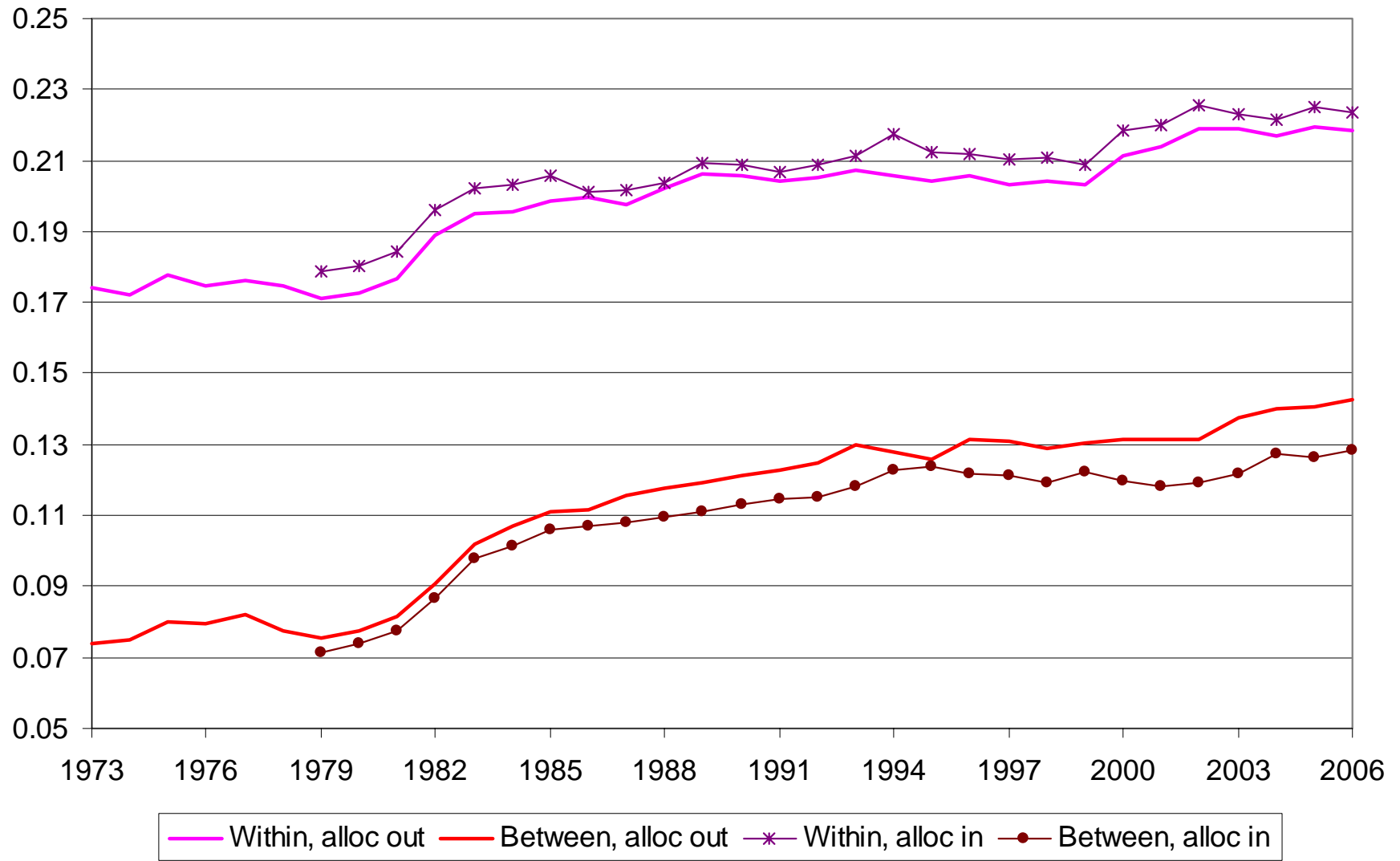
Main findings

- Within does not play much of a role in the growth in overall dispersion, especially after controlling for composition effects
 - But top and low end of residual distribution moving in opposite directions
 - Most of the growth in inequality linked to between-group component
 - Men: relative wages of college and post-graduates
 - Women: same plus experience gap
 - Most of the growth in the 1980s, though continuing growth at top end in the 1990s/2000s
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Robustness to measurement issues

- Allocators or no allocators?
- Top-coding
- March or May-ORG?

Figure 10a: Variance with and without Allocated Wages, Men



**Figure 10b: Variance with and without Allocated Wages,
Women**

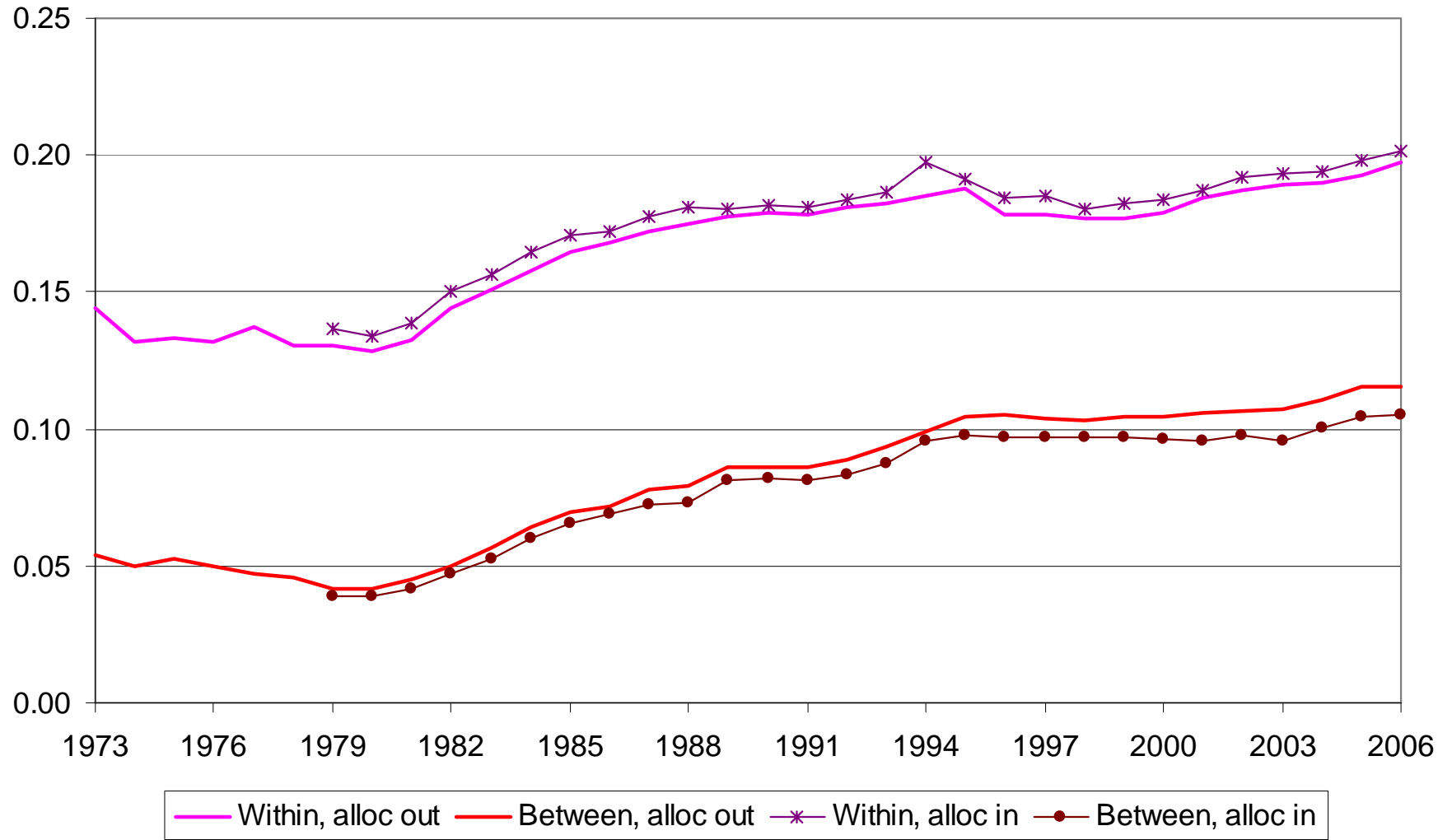


Figure 11a: Effect of Top-coding Adjustment, Men

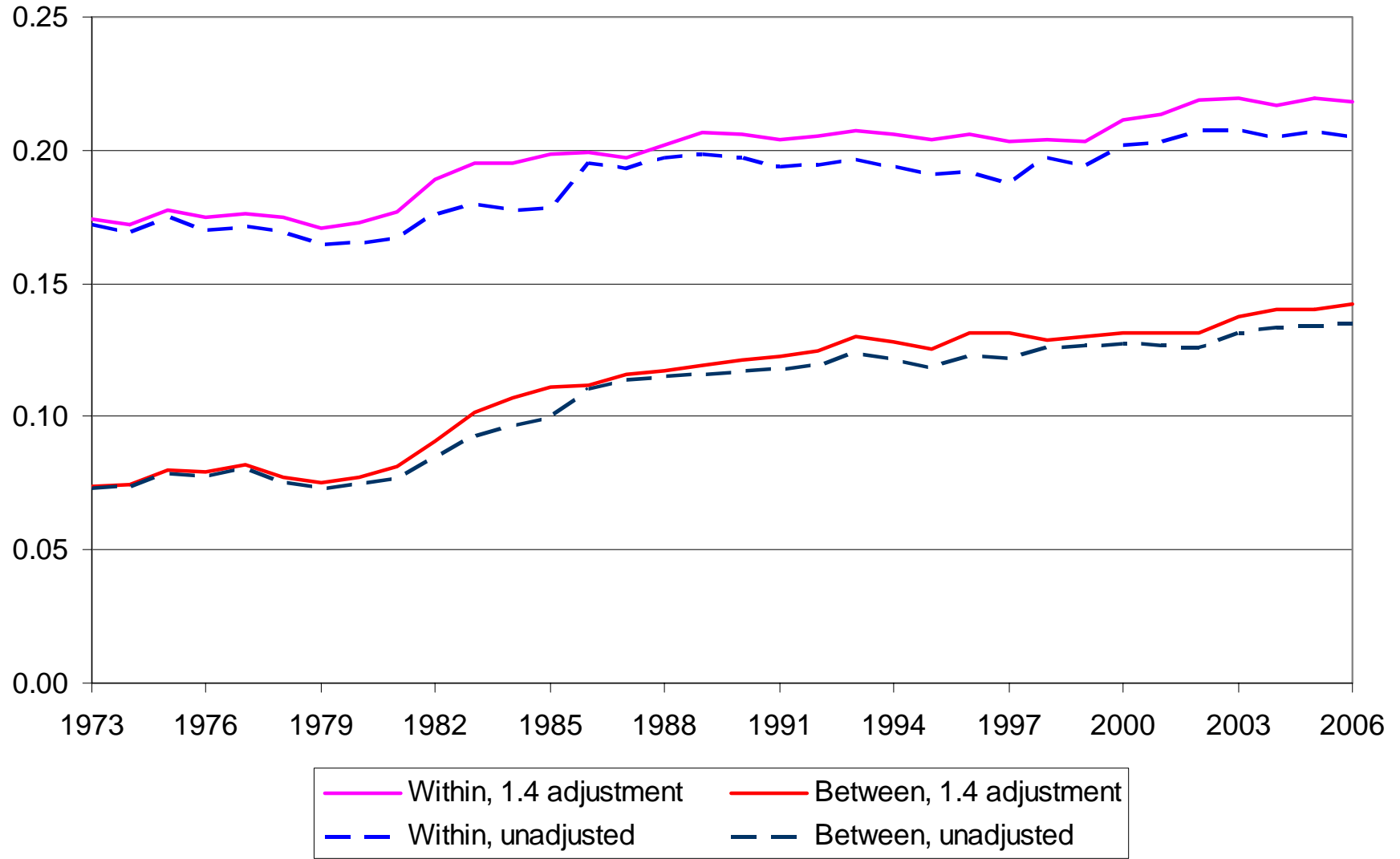


Figure 11b: Effect of Top-coding Adjustment, Women

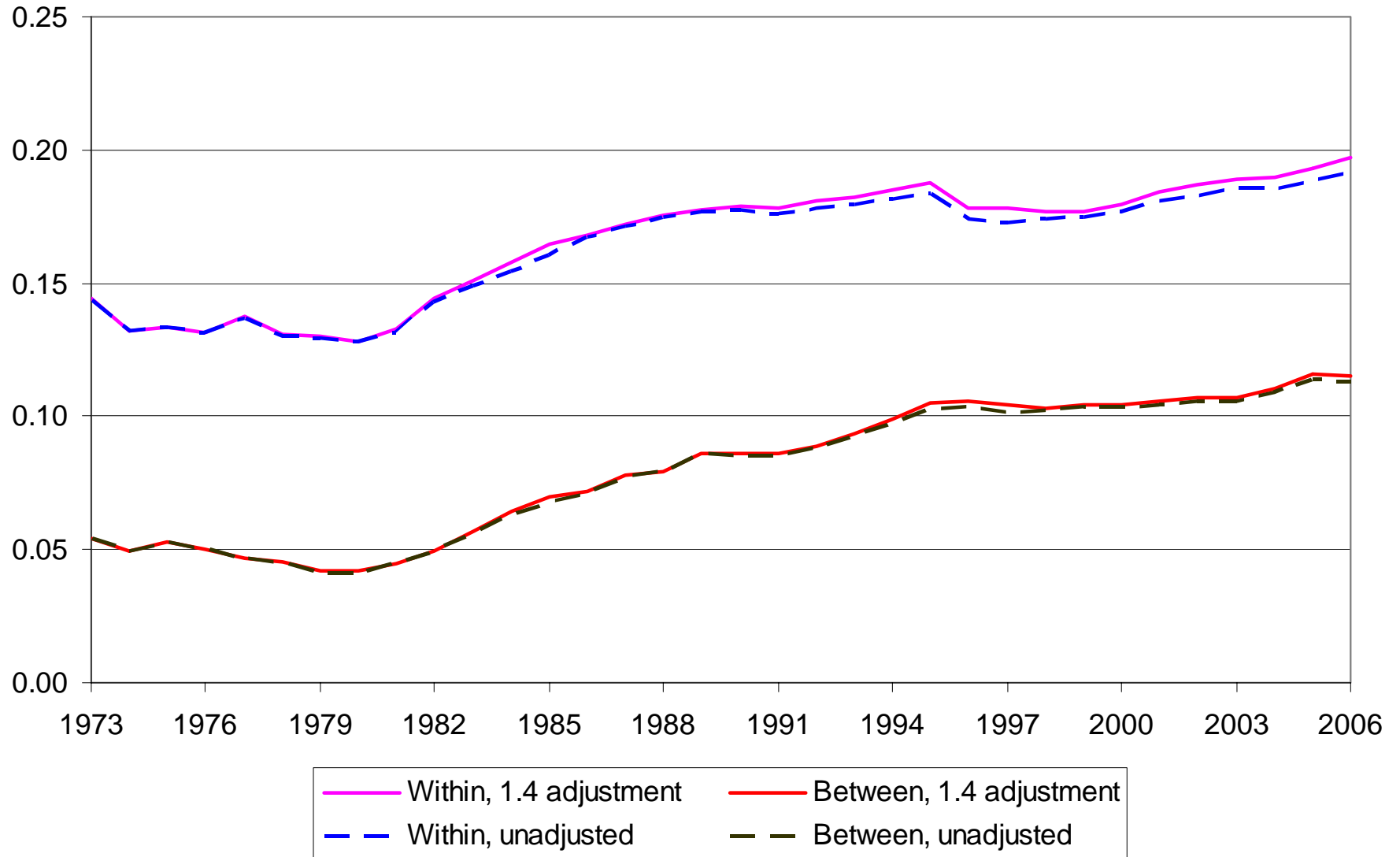


Figure 12a: Variance in May-ORG vs. March CPS, Men

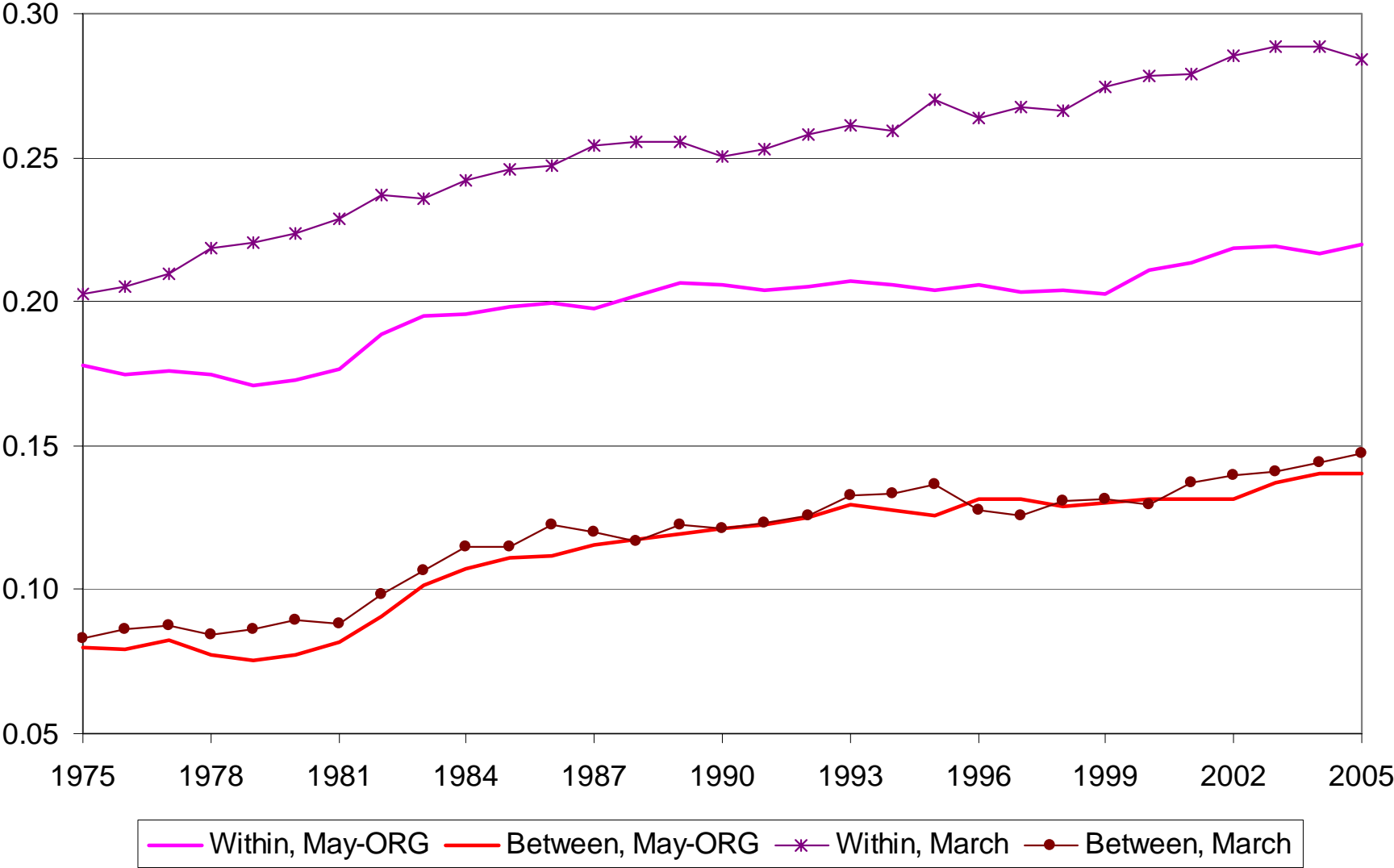


Figure 12b: Variance in May-ORG vs. March CPS, Women

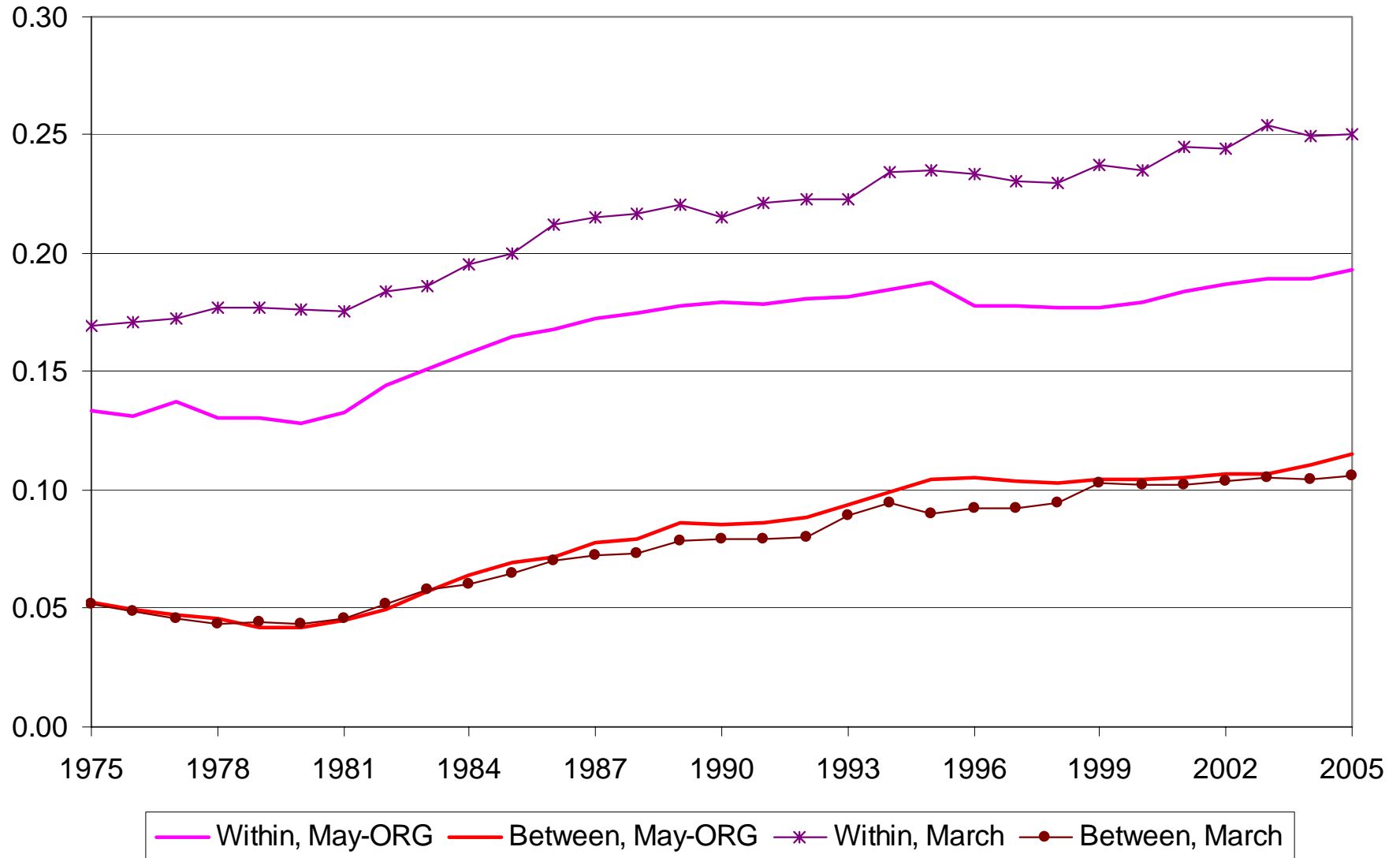
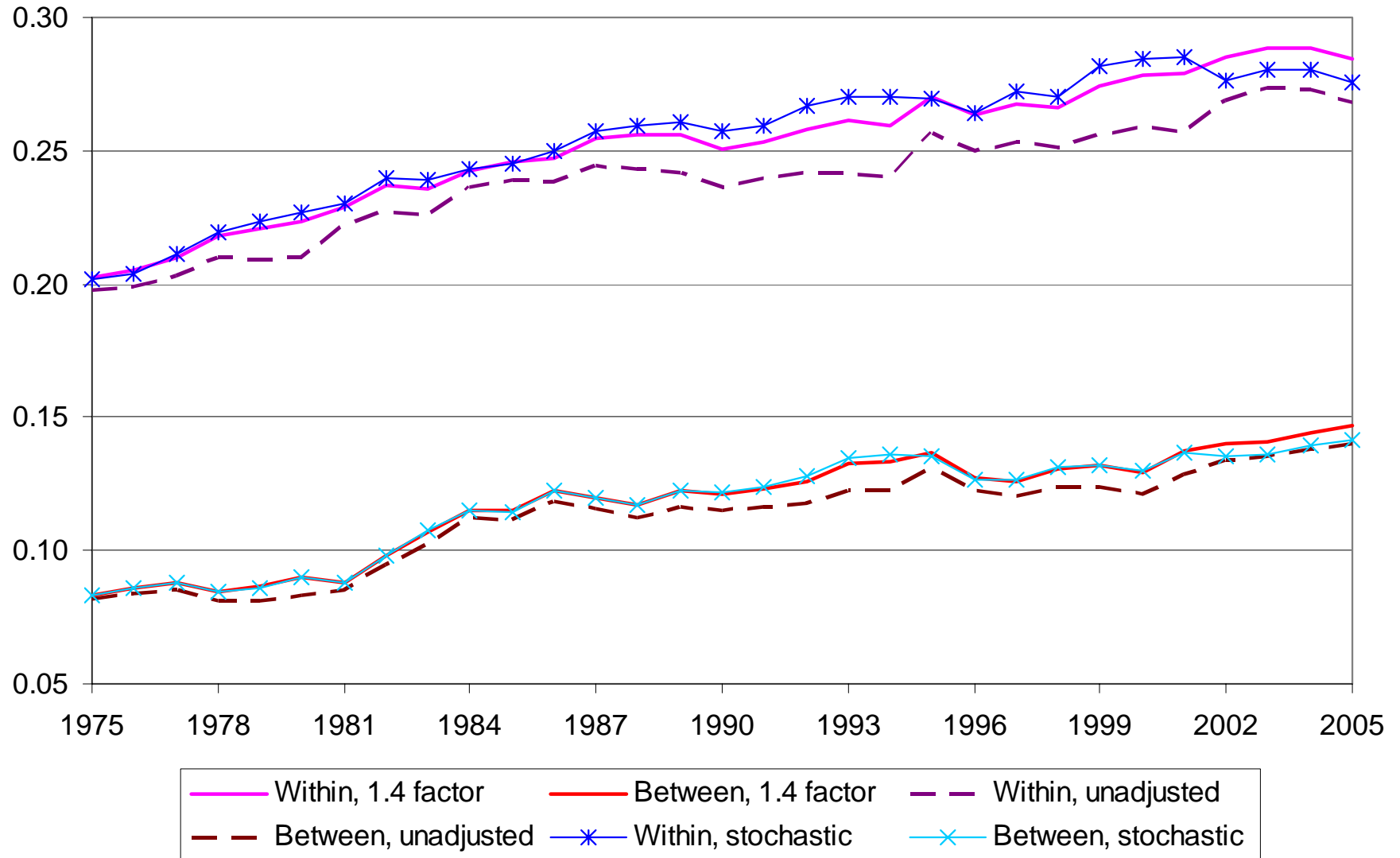


Figure 13a: Top-coding Adjustments in March CPS, Men



Main findings

- Evidence for between-group variance/wage differentials extremely robust to these measurement issues
 - Main difficulty is the difference in the level and trend in within-group inequality between May-ORG and March CPS
 - Level differences well explained by noisy measurement of hourly wage for workers paid by the hour in the March CPS
 - Trend differences a bit of a mystery
 - Should not based too much of proposed explanations on the within component
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What is to be explained

- No longer clear that within-group inequality grew in the 1970s: Problem for JMP story?
 - Major slowdown in inequality growth in the 1990s/2000s, but inequality still growing at the top-end
 - What is so special about post-secondary education?
 - Why such an explosion in earnings at the very top-end?
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Within-inequality in the 1970s: Problem for JMP?

- Not really: if unobservable ability and education are close substitutes the various “prices” should move in tandem
 - Offsetting factors at play: e.g. growing minimum wage for women
 - Does not really cause problems for the basic supply and demand story
 - But makes the “episodic” nature of the growth in inequality (1980s) more salient
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Top-end vs low-end inequality 1

- Autor and co-authors: computerization mostly negatively affects “routine” but often “skilled” tasks in the middle of the wage distribution
 - Plausible but still needs further probing
 - Most of inequality growth concentrated in 1980s: do you need institutional factors (unions and minimum wages) to get this?
 - Direct evidence of wages impacts not quite there yet
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Top-end vs. low-end inequality 2

- De-unionization does this too...
 - Unions mostly move (for men) workers from lower middle to middle: 50-10 up, 90-50 down
 - De-unionization can account for about 25% of the decline in 50-10 and growth in 90-50 since the late 1980s (Firpo, Fortin, Lemieux, 2007)
 - Other changes in wage setting institutions (performance pay) also help explain some of the growth at the top-end (Lemieux, MacLeod, and Parent, 2007)
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What is so special about post-secondary education?

- Provides you with cognitive/non-routine skills?
 - Stagnation/low growth in supply for men?
 - Elasticity of substitution low are higher level, high at lower levels? Plausible for highly specialized post-graduate degrees?
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Stunning growth at the real top end

- Unfortunately CPS/Census not too useful because of top-coding
 - PSID not top-coded but too small
 - Tax data is great but little demographics, in particular education
 - March CPS now collects non-topcoded earnings? If so progress could be made using master data.
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