xtoaxaca - Extending the Kitagawa-Oaxaca-Blinder Decomposition Approach to longitudinal data analyses

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Who would benefit from using xtoaxaca

- You have at least two time points
- You want a flexible way to decompose the level over time
- You want a (counterfactual) decomposition of change over time

Prior approaches

- Based on different research questions very many different decompositions can be chosen
- At the moment we cannot make a systematic comparison
- We present our preferred solution

$$\blacksquare R = E(Y_A) - E(Y_B)$$

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$$R = E + C + I$$

$$R = [E(X_B) - E(X_A)]\beta_A + E(X_A)(\beta_B - \beta_A) + [E(X_B) - E(X_A)](\beta_B - \beta_A)$$

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Research questions for the parts of the decomposition

■ **E**: How much smaller/bigger would the gap be, if the endowments of group A were the same as for group B?

Research questions for the parts of the decomposition

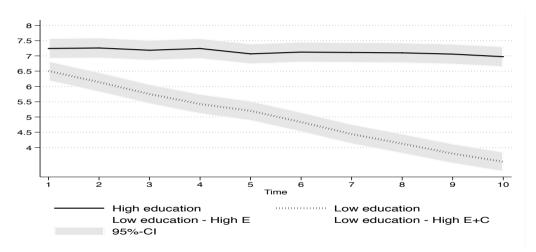
- E: How much smaller/bigger would the gap be, if the endowments of group A were the same as for group B?
- **C**: How much smaller/bigger would the gap be, if the effect of the explanatory variables of group A were the same as the effects for group B?

GC prediction approach

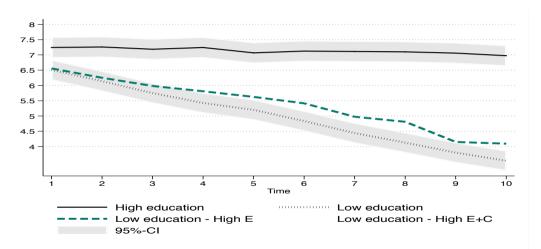
$$R(t) = E(Y_B(t) - Y_A(t))$$

- Latent growth curve model for change in happiness
- parametric, semi-parametric, non-parametric

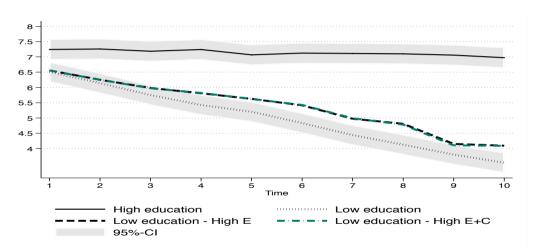
Predictions of happiness



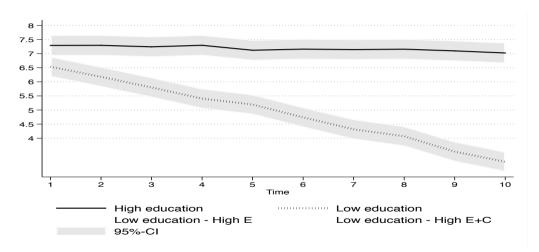
Counterfactual predictions of happiness



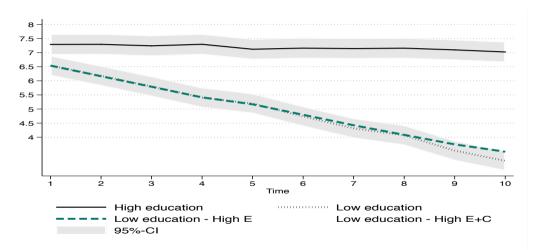
Counterfactual predictions of happiness



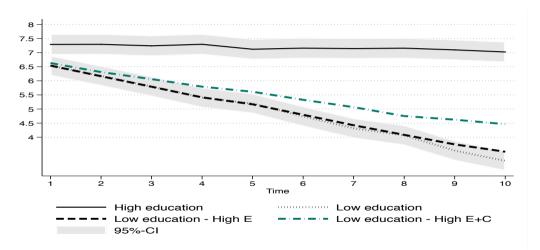
Predictions of happiness



Counterfactual predictions of happiness



Counterfactual predictions of happiness



•
$$R(t) = E(t) + C(t) + I(t)$$

Decomposition is time dependent

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Decomposition is time dependent

$$R = [E(X_{ht}) - E(X_{lt})]\beta_{lt} + E(X_{lt})(\beta_{ht} - \beta_{lt}) + [E(X_{ht}) - E(X_{lt})](\beta_{ht} - \beta_{lt})$$

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Decomposition is time dependent

$$\blacksquare R = [E(X_{ht}) - E(X_{lt})]\beta_{lt} + \frac{E(X_{lt})(\beta_{ht} - \beta_{lt})}{\beta_{lt} + \beta_{lt}} + [E(X_{ht}) - E(X_{lt})](\beta_{ht} - \beta_{lt})$$

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Decomposition is time dependent

$$\blacksquare R = [E(X_{ht}) - E(X_{lt})]\beta_{lt} + E(X_{lt})(\beta_{ht} - \beta_{lt}) + [E(X_{ht}) - E(X_{lt})](\beta_{ht} - \beta_{lt})$$

Decomposition of change

- Decomposing the change in happiness
- it always needs two time points to compare

Decomposition of change

$$\Delta Y^{l} = Y_{t}^{l} - Y_{s}^{l}$$
$$= \bar{X}_{t}^{l} \beta_{t}^{l} - \bar{X}_{s}^{l} \beta_{s}^{l}$$

Decomposition of change

$$\Delta Y^{l} = Y_{t}^{l} - Y_{s}^{l}$$
$$= \bar{X}_{t}^{l} \beta_{t}^{l} - \bar{X}_{s}^{l} \beta_{s}^{l}$$

and the change of the group difference over time then is

$$\begin{split} \Delta Y &= \Delta Y^A - \Delta Y^B \\ &= (\bar{X}_t^A \beta_t^A - \bar{X}_s^A \beta_s^A) - (\bar{X}_t^B \beta_t^B - \bar{X}_s^B \beta_s^B) \\ &= \bar{X}_t^A \beta_t^A - \bar{X}_s^A \beta_s^A - \bar{X}_t^B \beta_t^B + \bar{X}_s^B \beta_s^B \end{split}$$

Change due to endowment

$$\Delta Y_E = (\bar{X}_t^A - \bar{X}_s^A)\beta_s^A - (\bar{X}_t^B - \bar{X}_s^B)\beta_s^B$$

Change due to endowment

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Change due to coefficients

$$\Delta Y_C = (\beta_t^A - \beta_s^A)\bar{X}_s^A - (\beta_t^B - \beta_s^B)\bar{X}_s^B$$

Change due to coefficients

$$\Delta Y_C = (\beta_t^A - \beta_s^A) \bar{X}_s^A - (\beta_t^B - \beta_s^B) \bar{X}_s^B$$

Research questions for the parts of the decomposition

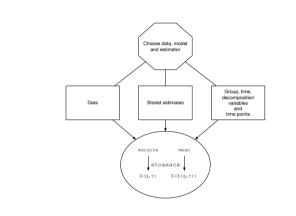
■ **dE**: How much smaller/bigger would the **change** in the gap be, if the endowments of group A had **changed** in the same way as for group B (and the difference in coefficients had stayed the same)?

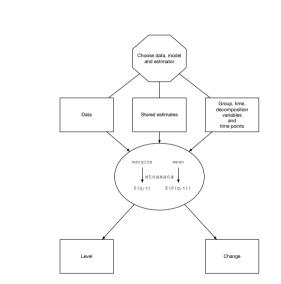
Research questions for the parts of the decomposition

- **dE**: How much smaller/bigger would the **change** in the gap be, if the endowments of group A had **changed** in the same way as for group B (and the difference in coefficients had stayed the same)?
- dC: How much smaller/bigger would the change in the gap be, if the coefficients of group A had changed in the same way as for group B (and the difference in endowments had stayed the same)?

Choose data, model and estimator







Quick note on the use of margins

- all decompositions are done at means or other specified values
- calculation of average over the population not necessary
- margins conducted only on one observation
- speed of the calculation of margins (not mean) independent of sample size

```
xtoaxaca exp, groupvar(group2) groupcat(1 2) timevar(time)
times(1 3 5 ) /// model1(base) model2(control) ///
timebandwidth(1) basecontrols(edu) change timeref(1)
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Stata example

General problems with decomposition

- Reference group
- Reference group of decomposition variables

Extensions?

- Different forms of change decomposition (e.g. Kim, Makepiece)
- Different counterfactual scenarios

- Reduce to one model estimation
- Extension for SEM

