

A MICROSIMULATION BASED METHOD TO NOWCAST HOUSEHOLD INCOME SURVEY DATA

WORK IN PROGRESS!

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RESEARCH

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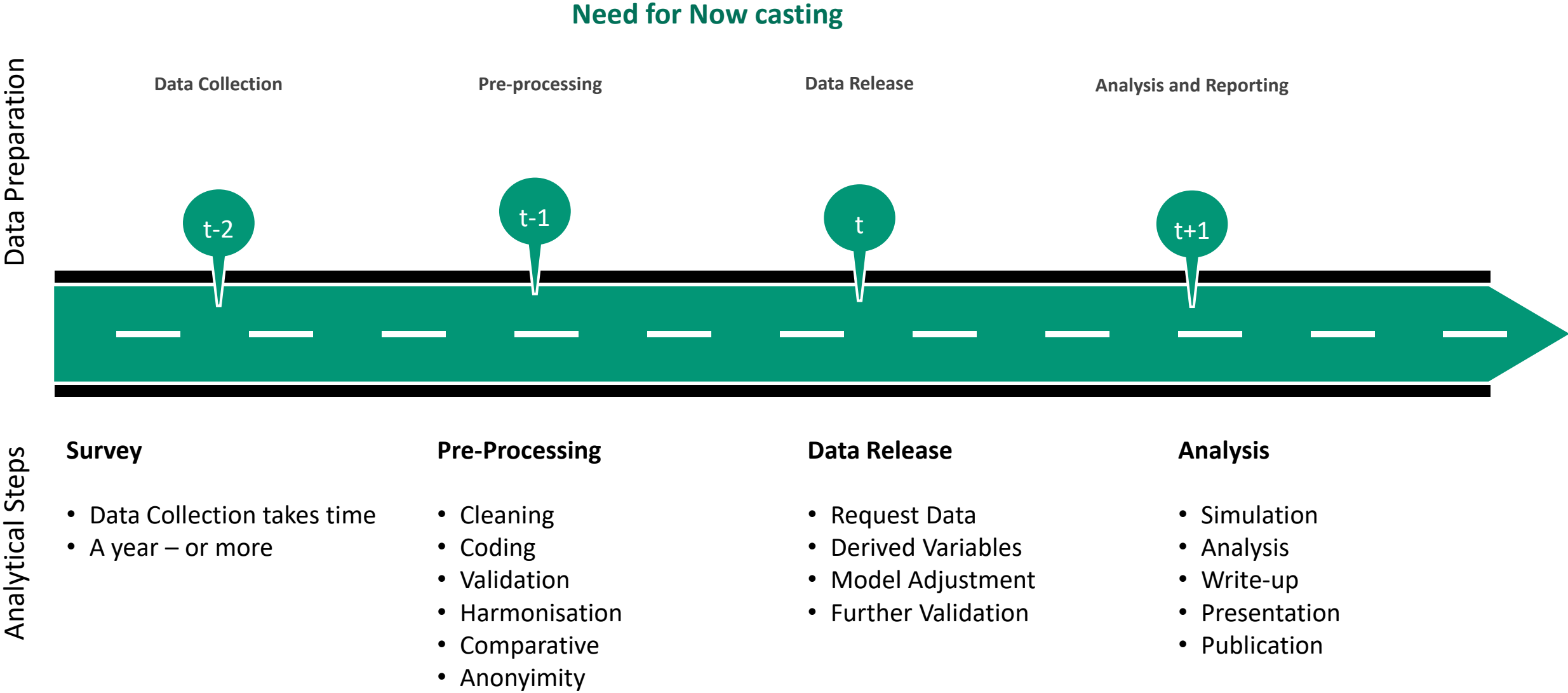


Centre for Microsimulation and Policy Analysis

— Home of Microsimulation —

Expert meeting on Nowcasting and mid-term
projections through microsimulation models

CONTEXT – SURVEY DATA COLLECTION



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NOWCASTING



- Macro-economic literature (Giannone et al., 2008)
- Directly modelling poverty incidence (Álvarez et al., 2014)
- O'Donoghue and Loughrey (2014) survey article
- EUROMOD Leventi et al., (2014); Navicke et al., (2014) applied Nowcasting to update poverty indicators, calibrating to LFS using transition probabilities of employment
 - Distributional impact of COVID crisis (Figari and Fiorio, 2020; Brewer and Tasseva, 2020; Brewer & Gardiner, 2020; Bronka et al., 2020).
- Addabbo et al., (2016) extend the parametric perspective of EUROMOD by modelling employment transitions using estimated logit equations.
- Carta (2019) Instead of taking labour force status from household surveys, imputes labour incomes into LFS.
- Li et al (2020) used semi-parametric perspective, a la DiNardo, Fortin and Lemieux (1996)

NOWCASTING – CHOICE OF METHOD



- Use of Monte-Carlo simulations based upon cell-specific probabilities may ignore some of the important heterogeneity exhibited in a crisis. For example, family status may be an important driver.
 - Partners within a family may be correlated as identified in Carta (2019).
 - Carta (2019) avoids issues associated with intra-household variation or sectoral biases by using recent labour force survey data.
 - Ignoring the impact of other non-labour force characteristics and the impact of public policy as a insulating mechanism is an issue.
 - Question → is it harder to simulate labour market changes in a model that contains the full range of incomes and policies or vice versa.
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NOWCASTING – CHOICE OF METHOD

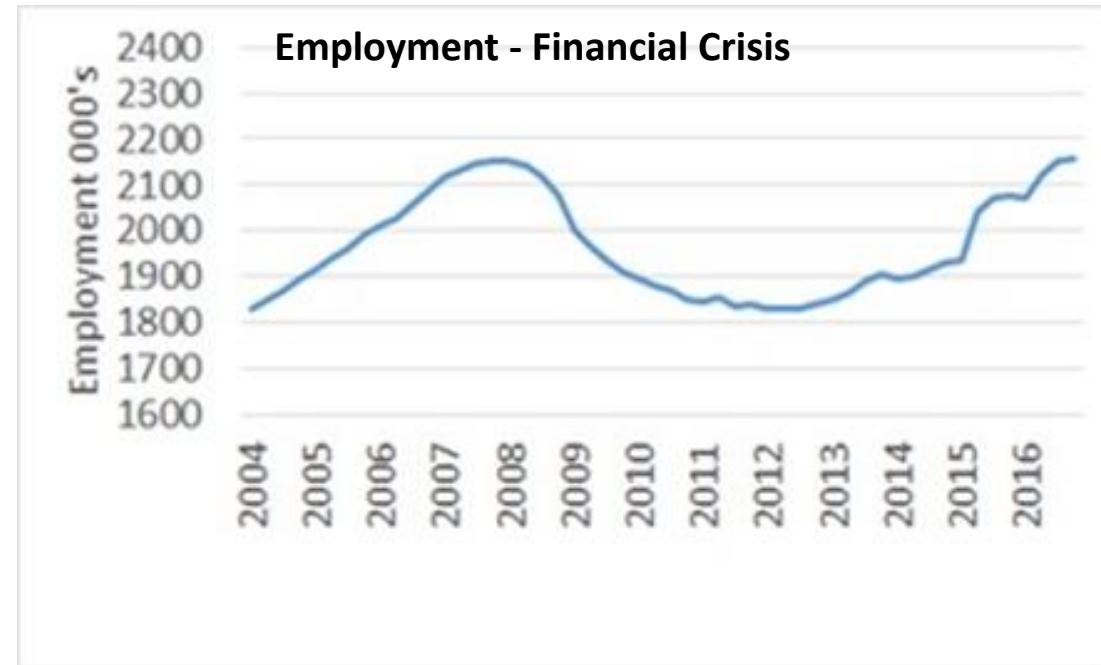


- Reweighting or semi-parametric approaches are strong in the sense that they avoid distributional assumptions.
 - However like other static-reweighting procedures they rely on sufficient sample sizes and face the risk with very many dimensions of relying for cell weights on small numbers. (Klevmarken 1997).
- In the case where reweighting or semi-parametric approaches prove unfeasible, a parametric approach may be a more pragmatic approach.
- In this paper we consider in more detail a parametric approach akin to the Addabbo et al., (2016) methodology, but in addition drawing upon the alignment aspects of the dynamic microsimulation literature.



THE FINANCIAL CRISIS

- Financial Crisis 2008-2012
- One of the deepest and longest lasting Financial Crises (FC)
 - Worst year 2009. Lowest point 2012
 - Young and Male
 - Big tax loss



THEORETICAL FRAMEWORK — INCOME GENERATING PROCESS

- Estimate system of equations representing
 - Z Demographic and Data Sampling Error
 - $I_i()$ Presence of Income Source i
 - $Y_i()$ Level of Income Source i

$$Y_{it}^H = \sum_F \sum_I \sum_S \frac{Y(a + B_{st}X_{it} + v_{si} + u_{ist})}{I(c + G_{st}Z_{it} + w_{si} + e_{sit})}$$

Correlation between income sources

$$\text{Corr}(u_{is_1t}, u_{is_2t}) = \rho_{s_1s_2}$$

Correlation over time

$$\text{Corr}(u_{its_2}, u_{ist_2}) = \rho_{t_1t_2}$$

Correlation between people in household

$$\text{Corr}(u_{i_1st}, u_{ist}) = \rho_{i_1i_2}$$

We assume they remain the same – in reality they change

THEORETICAL FRAMEWORK — ASSUMPTIONS ABOUT PROCESS

$$Y_{it}^H = \sum_F \sum_I \sum_S \frac{Y(a + B_{st}X_{it} + v_{si} + u_{ist})}{I(c + G_{st}Z_{it} + w_{si} + e_{sit})}$$

Assume

$$\text{Var}(u_{ist,}) = \text{constant} \quad \text{Var}(e_{ist,}) = \text{constant}$$

$$B_{st}=B_{st+1}$$

$$G_{st}=G_{st+1}$$

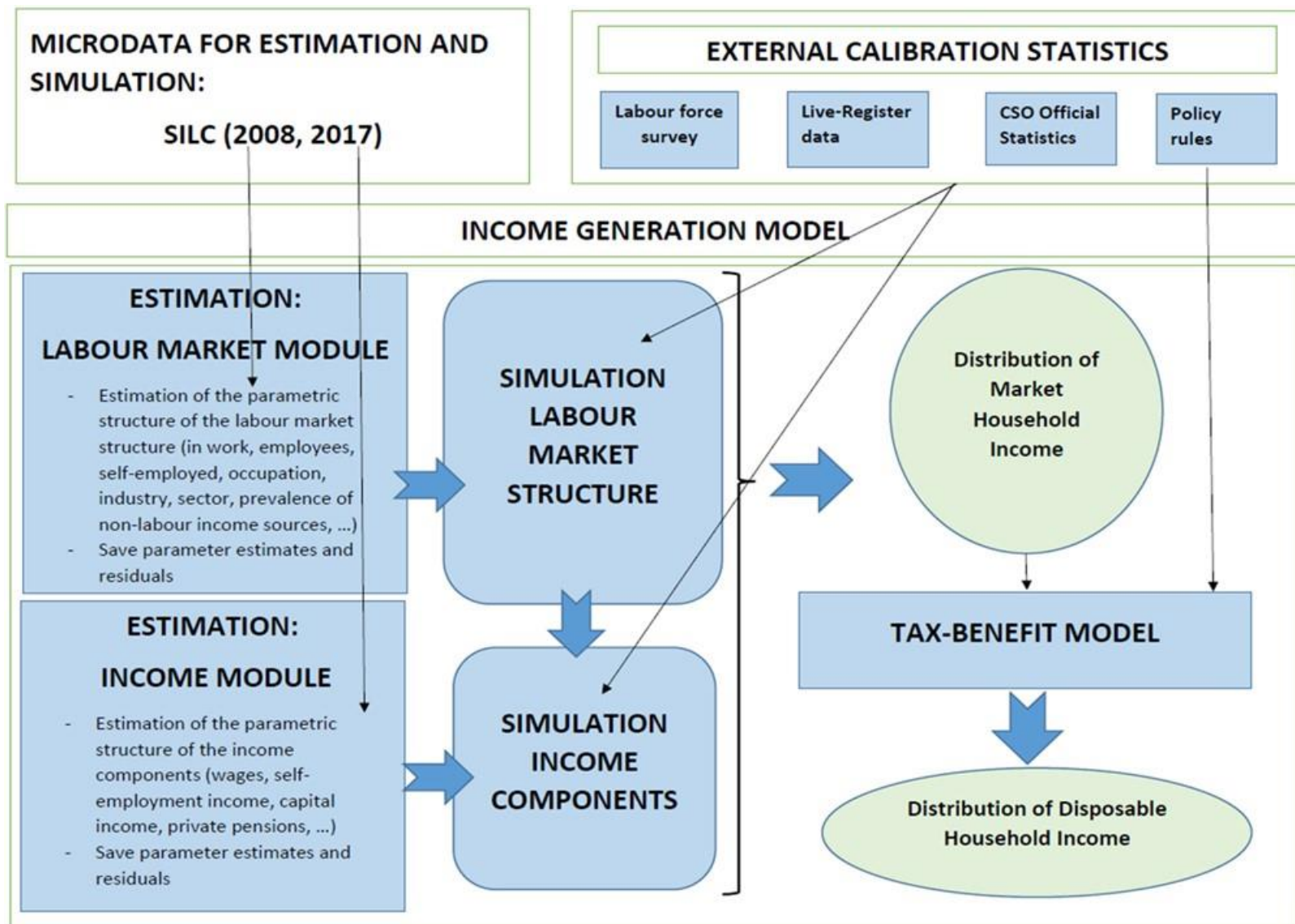
$$X_{st}=X_{st+1}$$

$$Z_{st}=Z_{st+1}$$

- Calibration means adjusting the constant or at most adjusting the constant for separate population groups for which we have control total
- In addition to the assumptions about the data generating process that are made
- There is sampling variation → so other variation

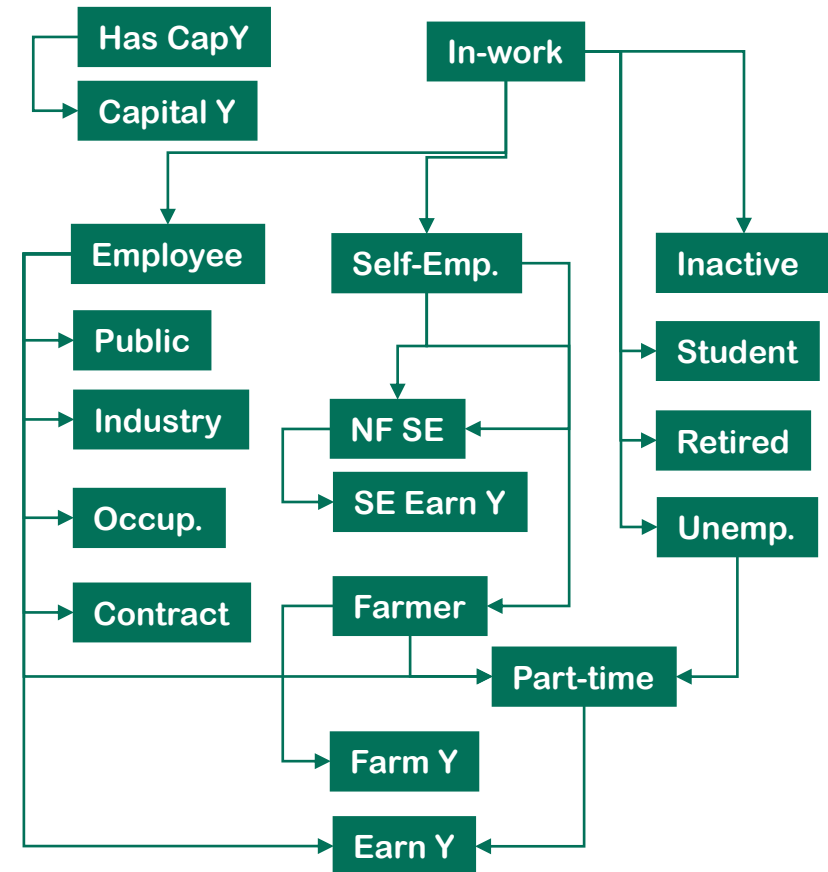
We assume they remain the same – in reality they change

METHODOLOGY



METHODOLOGY – SYSTEM OF EQUATIONS

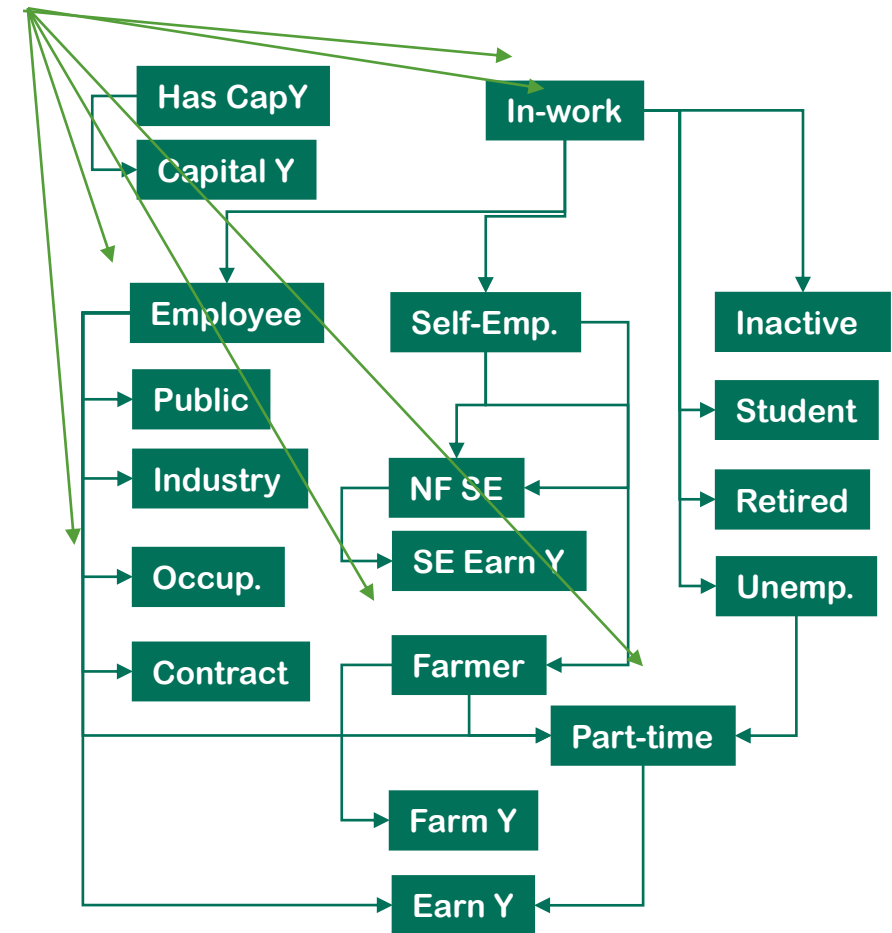
- relies on a system of hierarchically structured, multiple equation models for detailed income sources, combining:
 - a set of personal (individual and household) characteristics, X ,
 - parameters describing how employment, the receipt and level of income sources vary with personal characteristics, B ,
 - residuals linking model predictions to observed income sources, e ;
 - a tax-benefit simulator for converting market incomes into disposable incomes
 - parametric structure of the IGM
- $Y = m(X, e, l(B), r(B), tb)$
- l =labour market models, r =returns models, tb =tax-benefit system



METHODOLOGY – STEPS

$l^*(B_s)$

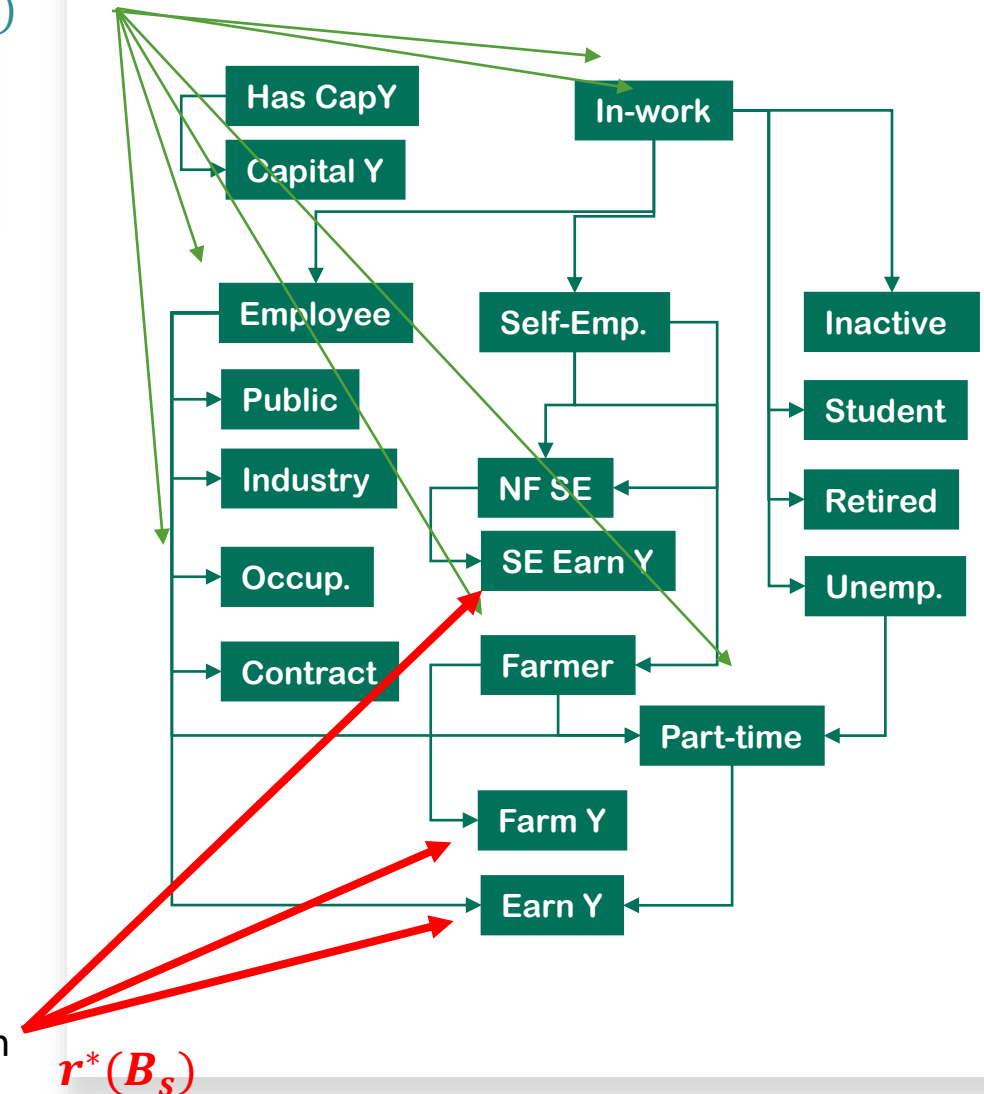
- Step 1: Estimation
 - we use the latest available survey year (s) to estimate the parameters of the IGM
- $Y_s = m(X_s, e_s, l(B_s), r(B_s), tb_s)$
- Step 2: Nowcasting to period t
 - X_s (exogenous characteristic), e_s (error structure), B_s (IGM parameters) are kept the same
 - $l^*(B_s)$ – labour market simulations are calibrated to external controls (go to picture)
 - $r^*(B_s)$
 - simulate all income sources based on the new labour market structure $l^*(B_s)$ (new inwork, new occupation, new industry structure → wage structure)
 - uprate pre-fiscal monetary variables to align them with the policy parameters
- tb_t policy parameters of period t



METHODOLOGY – STEPS

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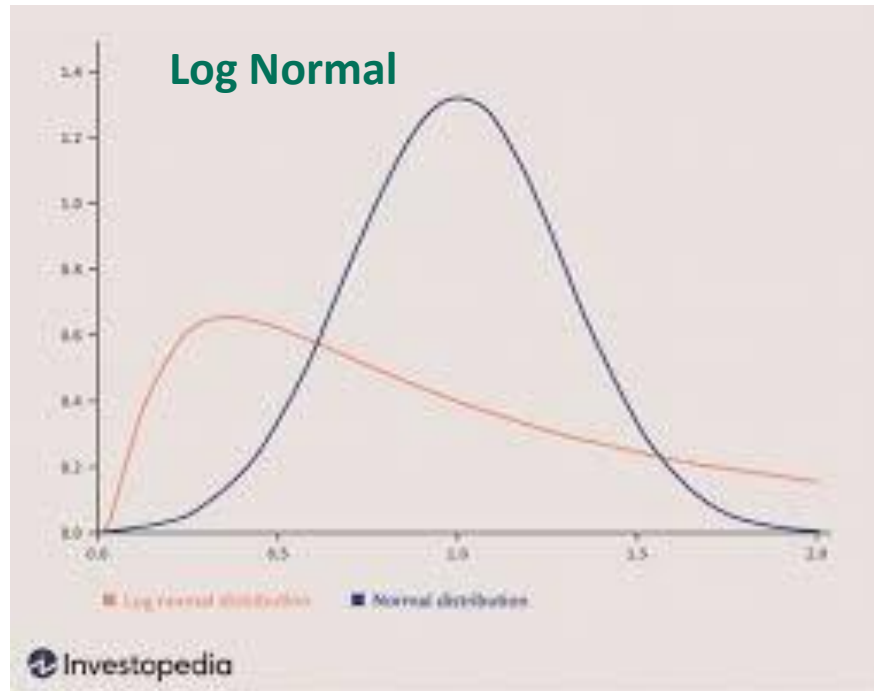


- tb_t policy parameters of period t

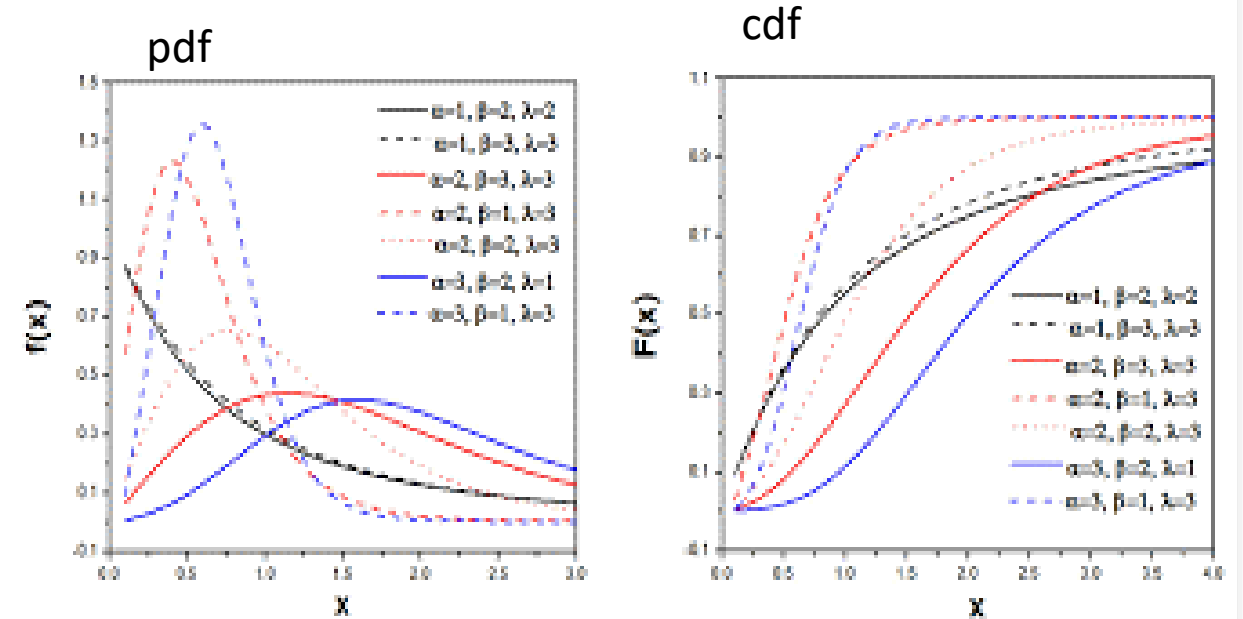
MAIN DIFFERENCES BETWEEN THE TWO NOWCASTING APPROACHES

IGM	Transition Matrix/Monte Carlo
Labour market component - Hierarchical system assures the sequential calibration of in-work, employment, occupational, industrial structure) - We allow transitions in-work and out-of-work	Labour market component - Applies employment probabilities to transition people out-of-work
Income component - Simulate all income components, taking into account the changes in the labour market structure (e.g. those that are classified in a different occupation in the calibration, have their wage updated)	Income component - Simulate incomes of those out of work

MODELLING ISSUE – EARNINGS DISTRIBUTIONAL ASSUMPTIONS



Singh Maddala Distribution – more control in parameterisation of curve



DATA ISSUES

- Time lag between collection and release for research and analysis.
 - The most recent analysis of SILC undertaken is for 2007 – compared with 2008 and 2009.
 - In normal times a lot happens in a two year period,
 - In a crisis the changes are so significant that such a lag can mean the data is relatively meaningless.
- There are more recent datasets available
 - LFS → available on a quarterly basis at a six week lag
 - the Live Register data and Price data → monthly basis on a short lag.
 - Pandemic Benefit Impact → weekly
- However these datasets do not contain income information.



ALIGNMENT IN IGM

- Simulate $GZ + v \rightarrow$ Logit Model
- Calibration Total by Sex and Gender
- Rank within Group
- Select the highest $GZ + v$ according to the external calibration totals



MODELLING CHOICES TO BE TESTED

- Do Models change over time?
- Is it important to model in more specific detail than in-work (ie structural changes)?
- How should we model income equations (Log-norm vs Singh Maddala Distributional Regression)?
- Comparing IGM with transition matrix/Monte Carlo
- Looking at historical crisis 2007-2009
- Use Simple Model then Full Model



DO MODELS CHANGE OVER TIME – CHOW TEST

Sex	Male		Male	
	Beta	S.E.	Beta	S.E.
University	1.2856***		0.10511.2856***	0.1051
Upper Secondary Education	1.1107***		0.08781.1107***	0.0878
Numer of Children Aged 0-3	0.1086		0.12350.1086	0.1235
Number of Children aged 4-11	-0.1594**		0.0626-0.1594**	0.0626
Number of Children aged 12-15	-0.2285***		0.0779-0.2285***	0.0779
Married	0.653***		0.09590.653***	0.0959
Age	0.3168***		0.01030.3168***	0.0103
Age Squared	-0.0038***		0.0001-0.0038***	0.0001
Per-Urban	-0.0884		0.0918-0.0884	0.0918
Rural	0.4898***		0.08900.4898***	0.0890
Year == 2009			-0.5203	0.3162
Year == 2009 x University			0.0949	0.1430
Year == 2009 x Upper Secondary Education			-0.2146*	0.1241
Year == 2009 x Numer of Children Aged 0-3			0.1195	0.1728
Year == 2009 x Number of Children aged 4-11			0.0092	0.0859
Year == 2009 x Number of Children aged 12-15			0.0013	0.1127
Year == 2009 x Married			-0.052	0.1342
Year == 2009 x Age			0.0107	0.0150
Year == 2009 x Age Squared			-0.0001	0.0002
Year == 2009 x Per-Urban			-0.1125	0.1292
Year == 2009 x Rural			-0.1392	0.1247
Constant	-5.7354***		0.2159-5.7354***	0.2159



**Chow Test – Models
Different over time**



IS THE INDUSTRY DISTRIBUTION DIFFERENT?

- No difference in industry distribution when not simulated for Males, but difference for females
- IGM captures changes

Chi Square Test of Industry Distribution		
	Monte Carlo	IGM
Male	0.014	0.000
Female	0.116	0.000

ARE SIMULATED MEAN EARNINGS DIFFERENT? RATIOS

- Log Normal creates means that are different for both Transition Matrix/Monte Carlo and IGM
- Singh Maddalla very close

Ratio of Means		
	Male	Female
Monte Carlo	1.06	1.19
IGM (OLS)	1.07	1.19
IGM (SMD)	1.02	1.02

ARE SIMULATED MEAN EARNINGS DIFFERENT? STATISTICAL DIFFERENCE

- Means different for both Transition Matrix/Monte Carlo
- No difference for SMD

Statistically Significant Difference in Means		
	Male	Female
Monte Carlo	1	1
IGM (OLS)	1	1
IGM (SMD)	0	0

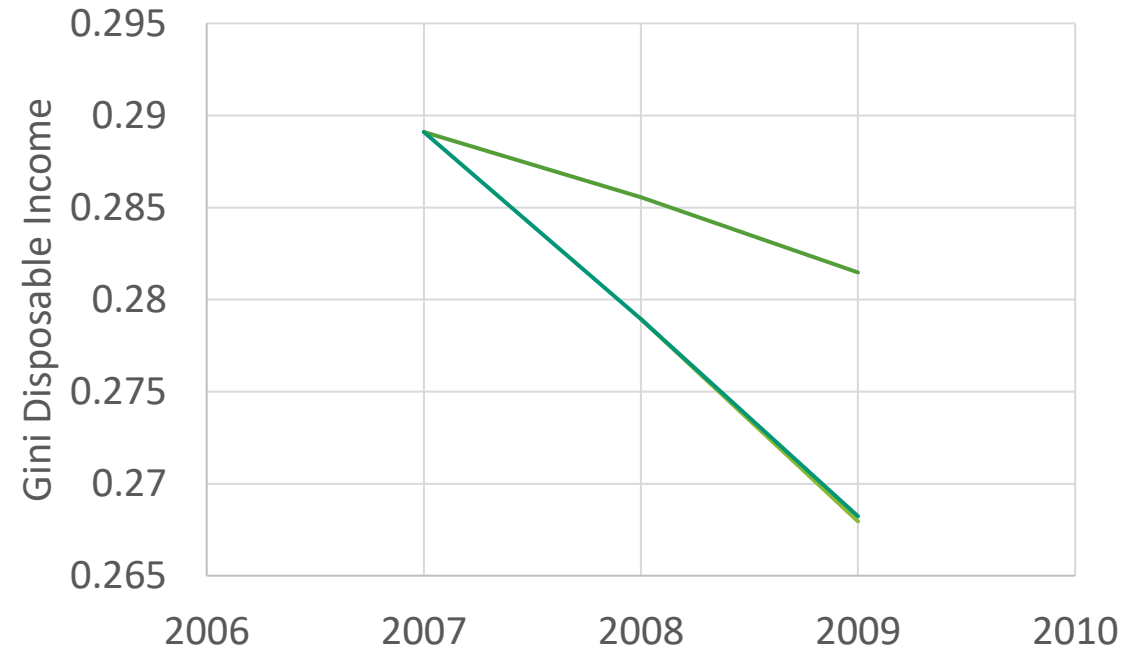
ARE SIMULATED MEAN EARNINGS DIFFERENT? KOLGOMOROV-SMIRNOV TEST OF DISTRIBUTIONS (P. VALUE OF DIFFERENCE)

- Transition probability/Monte Carlo Different for all
- Distributional Issues using log-normal
- No difference with actual for SMD for Males
- Difference of females due to changes in residuals not

Kolgomorov-Smirnov Test of Distributions (p. value of difference)		
	Male	Female
Relative to Actual 2009		
Monte Carlo	0.006	0.000
IGM (OLS)	0.006	0.000
IGM (SMD)	0.133	0.001
IGM (SMD - No Industry)	0.185	0.001

NOWCASTING — EVALUATION

- Evaluation
 - Looking back at simulation properties of a now cast during the financial crisis
 - Comparing simulated with actual, we find a good fit.
 - However performance weakens the further the now cast
 - Over longer term, method picks up on turning points and trend, but weaker on level



SUMMARY

- Nowcasting a useful methodology
- Not much difference between Transition Matrix and IGM for In-work
- Significant difference if other structural changes
- Problems in using Log-Normal distribution
- Model replicates changes in Income Distribution over 1 year
- Over longer term captures the trend, but weakens the further we get from estimation year
- What is the success bar?
- Future work – trends in B and Sigma's

Thank you

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