

Modelling the early lifecourse: A decision support tool for policy makers

International Methodology Symposium, Ottawa October 2014



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MINISTRY OF BUSINESS,

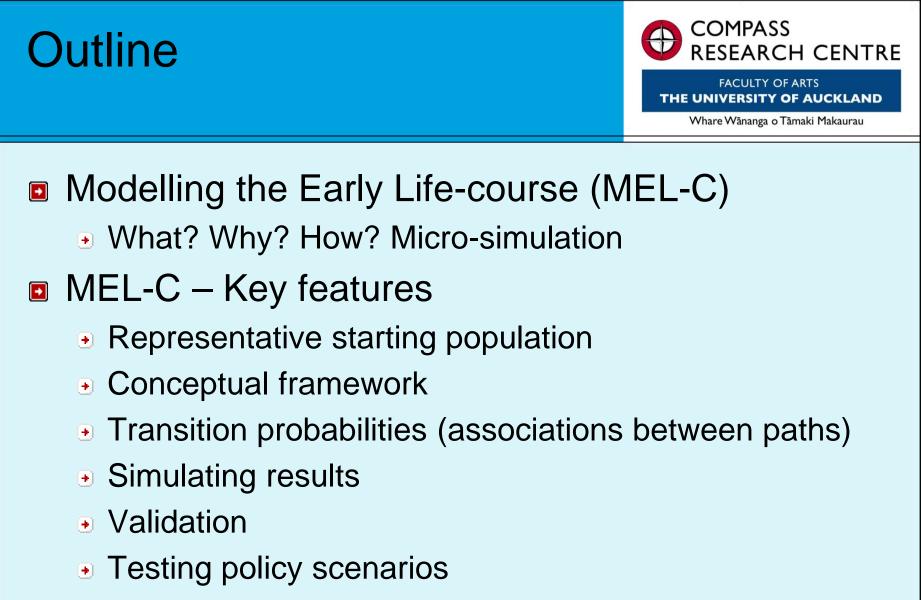
HIKINA WHAKATUTUKI

INNOVATION & EMPLOYMENT

Barry Milne

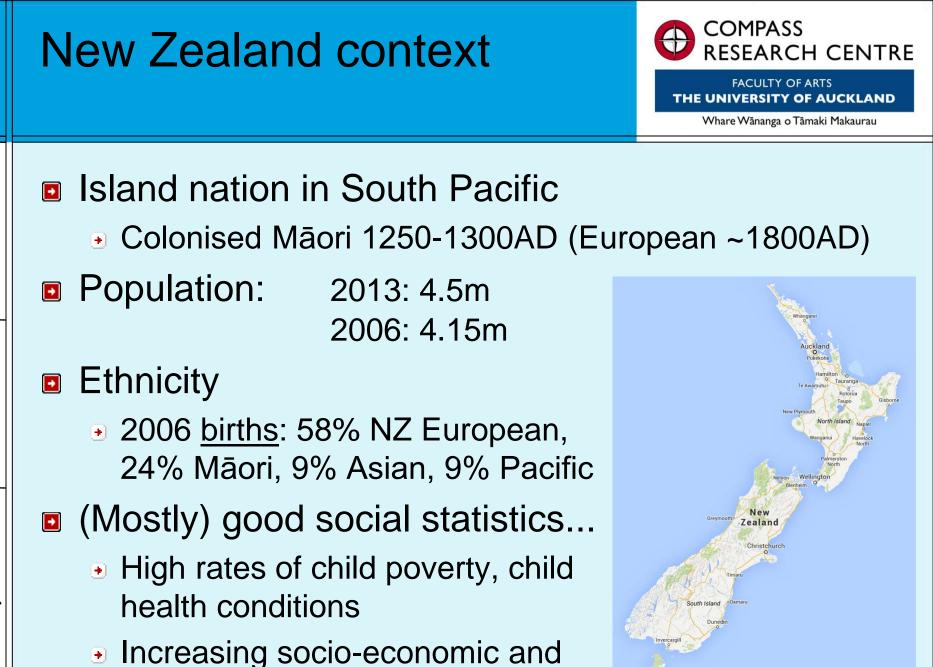
COMPASS Research Centre Faculty of Arts University of Auckland





- Some results
- Conclusions

New Zealand



skland New

Zealand

The University of Auckland

ethnic equality

MEL-C - What? Why? How?



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1. Goals ... what are we trying to do?

Develop a software application as a decision-support tool for policy-making

2. Rationale ... why are we doing it?

To improve policymakers' ability to respond to issues concerning children and young people

3. Means ... how are we doing it?

By building a computer simulation model with data from existing longitudinal studies to quantify the underlying determinants of progress in the early life course

MEL-C - Micro-simulation



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Micro-simulation

- Use data from the real world to create an artificial one that mimics the original.
- Carry out 'virtual' experiments with artificial world (Gilbert and Troitzsch 2005).
- Simulate individual units and aggregate these to get population estimates
- Requirements
 - Starting population
 - Rules to transition individuals from state to state

MEL-C How?



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- We start with a sample of individuals
 - Synthetic (derived from Census)
- We derive statistical rules to create a 'virtual <u>cohort'</u> through to age 13
 - Analyse combined data from 3 longitudinal studies
 - A sample of children with typical biographies over the life-course, allowing for variation
- We then simulate what might happen if policy were to change, by altering parameters
 - Using software application

Key features - Starting population



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- Subset NZ 2006 Census to include just new-borns (0-year olds) and their parents
 - Randomly select 10,000 (Processing speed)
- Calculate distance (Euclidean) between each of the 10,000, based on 52 Census characteristics.
- Choose the closest 2 ranks to form 10,000 clusters of 3 individuals
- Randomly choose which indiv's characteristics are used for each synthetic individual
 - Characteristic by characteristic

Key features - Starting population



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	<u>Clu</u>	ster of 3 Child	ren		
Characteristic	Child 1	Child 2	Child 3	Random Draw {1,2,3}	Synthetic child
Child sex	Male	Female	Female	2	Female
Mother age	29	41	31	1	29
Father age	32	40	38	1	32
Home ownership	Owned	Owned	Rented	3	Rented
Deprivation score (1-10)	9	7	8	3	8

Key features - Starting population

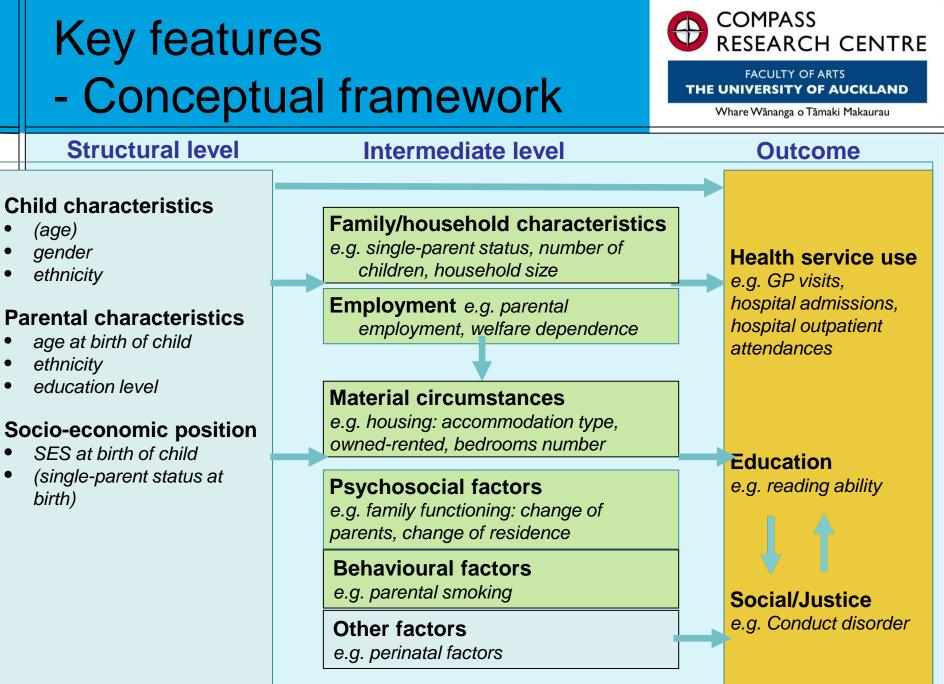


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- Voilà! A synthetic basefile of 10,000 composite individuals
- Match Census distributions and interrelations



Key features - Transition probabilities



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Combined data from 3 NZ longitudinal studies

- Christchurch Health & Development Study (CHDS)
 - 1265 children born in Christchurch, 1977. Followed since.
- Dunedin Multidisciplinary Health & Development Study (DMHDS)
 - 1037 children born in Dunedin, 1972/73. Followed since.
- Pacific Islands Families Study (PIFS)
 - 1398 children born at Middlemore, 2000, with at least one parent of Pacific Islands ethnicity. Followed since.
- Regression analysis of each path in conceptual framework
 - Data weighted to represent NZ current ethnic distribution
 - Māori data weighted to represent NZ Māori national cultural affiliation distribution (Te Hoe Nuku Roa Study)

Key features - Simulating results



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Simulating Reading score: Simplified rule from statistical model:

E[reading score] = 13.00 + .91*reading.score.previous + .07*months.breast.fed + 1.04*father.tertairy.qualification + .87*father.secondary.qualification

	Child A
Characteristics	
Reading score at age 8	40
Number of months breast fed	12
Father's Education	Secondary Apply Rule
Predicted reading score at age 9	13.00 + .91*40 + .07*12 + .87 = 50.58 Expected value
Random draw from a normal distribution	50.23 Stochastic
Reading score assigned at age 9	50 component

Key features - Validation



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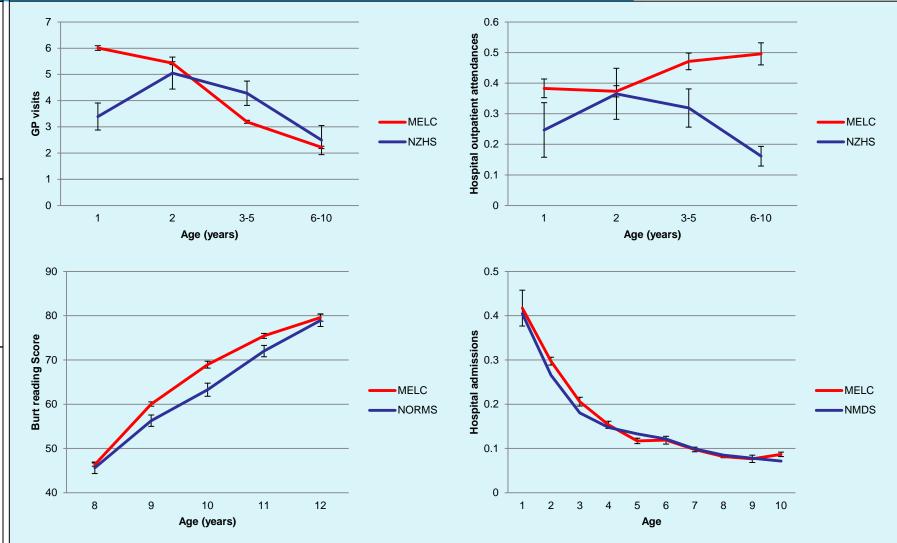
- Models need to simulate real world accurately
- Given we start with a representative NZ population (synthetic base file), we should simulate NZ national rates
- Compare simulated rates with published rates / available datasets
 - NZ Heath Survey (GP visits, outpatient attendances)
 - NMDS (hospital admissions)
 - Burt Reading norms (reading)
 - Nothing for conduct only national rates available are based on the longitudinal studies we use in our models!

Key features - Validation



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New Zealand

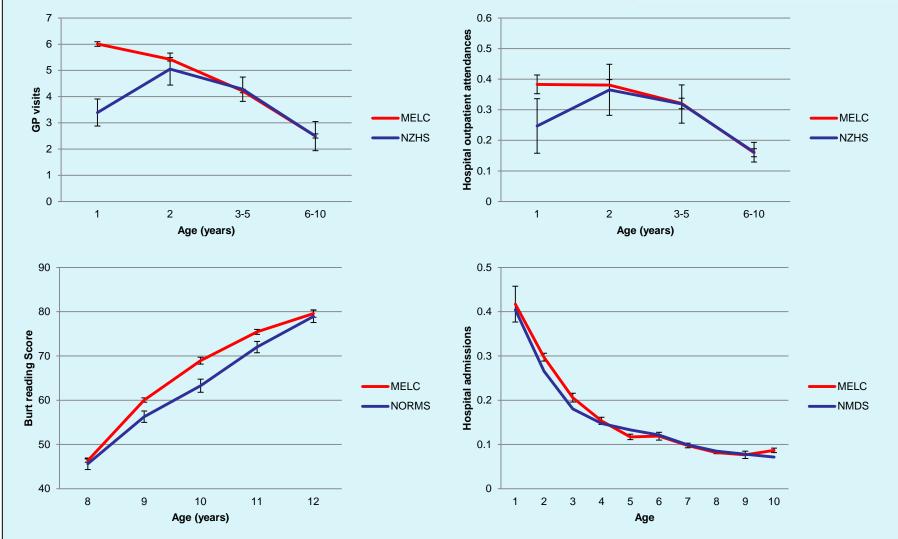
Key featuresValidation (realigned)

New Zealand

The University of Auckland



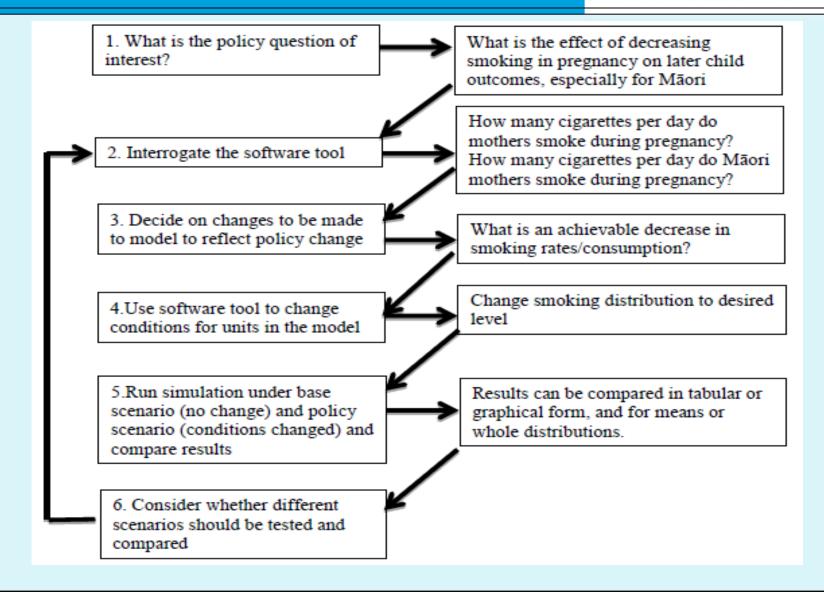
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Key features - Testing policy scenarios



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Modelling software



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Cigarettes smo	oked per day during	pregnancy						
Adjust Proporti	ons for your Scena	гіо						
	0 (%)	1-5 (%)	6-10 (%)	11-15	5 (%)	16-20 (%)	21	1+ (%)
Cigarettes	0 (%)	1-5 (%)	6-10 (%)	11-15	5 (%)	16-20 (%)	2	1+ (%)
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Modelling software



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Modelling software



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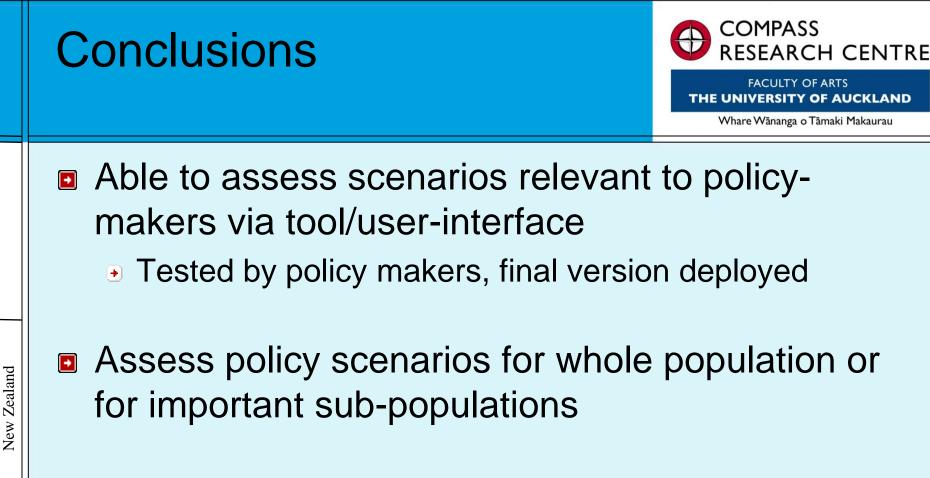
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Select Options				≤	≥	7	8	9
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Cigarettes	78.30	7.80	4.80		5.20	1.5	50	2.40
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- Use of existing data for simulation rules and for simulation starting population
 - Extend to use published 'best' estimates for rules
- Several scenarios can be assessed

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QUESTIONS?

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