

Using multiple longitudinal datasets to inform a micro-simulation model of the early life-course

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Whare Wānanga o Tāmaki Makaurau

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What are we doing – and why?



- To build a realistic simulation model of the early life course (0-13) for policy purposes, we are:
 1. Combining information across four longitudinal studies into a unified (more robust) data set.
 - To analyse to get rules for transitioning people from one state to the next
 2. Weighting the combined dataset by ethnicity
 - To analyse a sample that has a representative ethnic balance
 3. Preparing a synthetic birth cohort from 2006 Census
 - So that our simulation represents NZ today

- I will talk about 1 and 2 now, and 3 later

Four Studies



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- ❑ Christchurch Health & Development Study (CHDS)
 - 1265 children born in ChCh 1977. Followed since
- ❑ Dunedin Multidisciplinary Health & Development Study (DMHDS)
 - 1037 children born in Dunedin 1972/3. Followed since
- ❑ Pacific Islands Families Study (PIFS)
 - 1398 children born at Middlemore, 2000, with at least one parent of Pacific Islands ethnicity. Followed since
- ❑ Te Hoe Nuku Roa Study (THNR)
 - Longitudinal study of Māori households (beginning 1995)
 - Auckland, Wellington, Manawatu, Gisborne, Northland, Southland, Nelson
 - 568 children (0-12) assessed at least twice in four waves

1. Data integration



- ❑ Original model based on CHDS
 - ❑ Use data from DMHDS & PIFS on those constructs used in CHDS-based model; ignore other constructs
- ❑ Issues around
 - ❑ Different times
 - ❑ Same constructs measured differently
 - ❑ Missing data
 - ❑ Ensuring combined is representative of NZ
- ❑ Solutions

1. Data integration - Different times

- ▣ Associations between X & Y assessed using longitudinal GEE analyses
 - ▣ Utilises data from all the ages available from the three studies (THNR not used)

Age	Y _{CHDS}	Y _{DMHDS}	Y _{PIFS}	X _{CHDS}	X _{DMHDS}	X _{PIFS}
Birth	✓	✓	✓	✓	✓	✓
1	✓		✓	✓		✓
2	✓		✓	✓		✓
3	✓	✓		✓	✓	
4	✓		✓	✓		✓
5	✓	✓		✓	✓	
6	✓		✓	✓		✓
7	✓	✓		✓	✓	

1. Data integration - Construct measurement

- ❑ 30/36 constructs measured identically between CHDS & DMHDS; 24/26 between CHDS & PIFS
- ❑ 4 DMHDS constructs measured otherwise identically but cover different timeframe (e.g., past 2 years in DMHDS; past 1 year in CHDS)
 - Random imputation to subset to one year ($r \sim 0.65$)
- ❑ 2 DMHDS & 2 PIFS constructs measured using different scales
 - Conduct disorder, Harsh punishment
 - Align to same metric using min/max points

1. Data integration - Missing data



- ❑ 'Holes' in data in each study filled in
 - + 60% vars have <10% missing; 14% vars have 20-30%
 - + Model-based multiple imputation using within-study models, imputing vars with least error first (following SGP)
- ❑ Constructs in DMHDS/PIFS with missing ages
 - + 15% constructs
 - + Model-based multiple imputation using within-study models (or another study if time trends important)
- ❑ Missing constructs in DMHDS/PIFS
 - + 4/40 constructs in DMHDS, 14/40 constructs in PIFS
 - + Model-based multiple imputation using CHDS study models

2. Weighting by ethnicity



- Combined CHDS, DMHDS & PIFS not representative of NZ's ethnic distribution currently
 - Weight by ethnicity:

Ethnicity	DMHDS	CHDS	PIFS	Combined	Census	Weight
NZ European	90.1%	86.1%	2.8%	55.9%	58.2%	58.2/55.9 = 1.04
Maori	8.4%	10.7%	6.2%	8.4%	24.2%	24.2/8.4 = 2.88
Pacific	1.5%	3.2%	91.0%	35.7%	9.2%	9.2/35.7 = 0.26
Asian					8.5%	

2. Weighting by ethnicity - Cultural affiliation

- Likely that CHDS & DMHDS Māori not representative of Māori nationally
- Solution?
 - Use cultural affiliation as ‘representativeness’ indicator
 - Compare cultural affiliation between CHDS & DMHDS Māori and THNR Māori, and weight CHDS & DMHDS distributions to look like THNR
 - CHDS, DMHDS & THNR each have items on
 - Marae visit, Tangi attendance, involvement in Māori groups, language understanding, Māori language TV/radio
 - NB, No Māori cultural affiliation items in PIFS
 - Draw principal component from these items and compare CHDS & DMHDS against THNR quintiles

2. Weighting by ethnicity - Cultural affiliation distributions



Quintiles	THNR (%)	CHDS (%)	DMHDS (%)
1 - low	20.0	53.7	66.7
2	20.0	22.3	12.3
3	20.0	8.3	7.0
4	20.0	12.4	5.3
5 - high	20.0	3.3	8.8

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2. Weighting by ethnicity - Cultural affiliation weights



Quintiles	THNR (%)	CHDS (%)	Weight	DMHDS (%)	Weight
1 - low	20.0	53.7	20/53.7 =0.37	66.7	=20/66.7 =0.30
2	20.0	22.3	20/22.3 =0.90	12.3	=20/12.3 =1.63
3	20.0	8.3	20/8.3 =2.41	7.0	=20/7.0 =2.86
4	20.0	12.4	20/12.4 =1.61	5.3	=20/5.3 =3.77
5 - high	20.0	3.3	20/3.3 =6.06	8.8	=20/8.8 =2.27

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2. Weighting by ethnicity - Cultural affiliation assumptions

- A Māori sample representative on cultural affiliation will be a representative Māori sample
 - Perhaps. Geographic differences??
- THNR is a representative Māori sample
 - Probably for regions sampled.
 - Te Kupenga (Māori Social Survey) another option?
- Cultural affiliation is measured well by the items we used
 - Probably. Cultural affiliation items load on one factor.
- Cultural affiliation is stable across the life-course
 - Possibly. Items measured longitudinally (THNR) correlated moderately - strongly

Summary and Next Steps



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- ❑ Integration of data from datasets feasible
 - ❑ Bit of work, similarity of constructs has helped
- ❑ Method to make analysis sample ethnically representative
 - ❑ Weighting; including weighting to attempt to get a representative sample of Māori
- ❑ Analyses about to be undertaken
 - ❑ Can compare results from one vs. three studies
 - ❑ Can compare results for weighted vs. unweighted analyses