

# Little School on the Prairie: A Push for Structural Transformation

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<sup>a</sup>Supported in part by funding from the Social Sciences and Humanities Research Council. All errors are my own.

# The One Room School

- One room schools loom large in the American mythos and in the development of education systems around the world.
  - Common in the United States, Canada, United Kingdom, Australia, and Switzerland.
  - Formed an input for the US becoming a world leader in educating its population (Goldin & Katz, 2008).
  - Rural, ungraded, between ages of 5-14, teach children the 3R's: Reading, Writing, and Arithmetic.
- This project: What was the impact of one room schools on children and their communities?
  - What is the role of changing labor supply in structural transformation out of agriculture?

## Research Question: What was the impact of one room schools?

- Rural one room schools were not known for high quality education:
  - Low educational quality, poor resources, difficulty recruiting teachers, and lower educational returns (Goldin & Katz, 2000; Lachanski, 2024).
- But early forms of education were important:
  - Decentralization allowed communities to custom tailor education (Dippel et al., 2020).
  - Widespread schooling raised intergenerational mobility, decreasing correlation with mother's human capital (Althoff et al., 2025).
  - Schooling as a push factor out of agriculture (Caselli & Coleman II, 2001; Porzio et al., 2022).

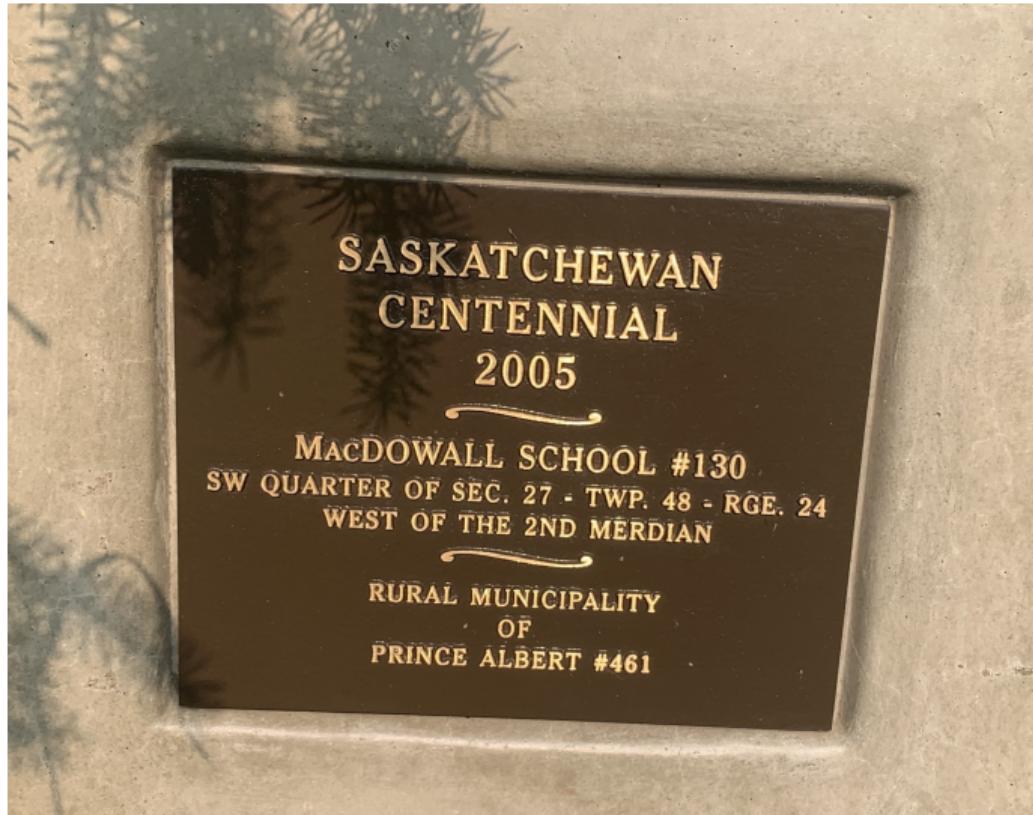
## Schools Push People out of Agriculture

- Utilize the rollout of rural schools in the western Canadian prairies at the start of the 20th century.
  - Combine census records with the near universe of school openings from 1871-1912.
  - Geolocate schools and households to exact land parcel (unique feature to Canada).
- Distance to school is likely endogenous, could reflect a variety of factors.
- Plausibly exogenous timing of school construction relative to a child's age.
  - Counterfactual is having a school versus not.

## School Example: MacDowall/Cecil School #130



## School Example: MacDowall/Cecil School #130



## School Example: MacDowall School #130



## Preview of Results

- Estimate a difference-in-differences finding that exposure to one room schools:
  - Decreases the likelihood of being a farmer 3.4pp (6%), increases incomes 12.3%, and increases distance from childhood home by 12.7%.
  - Increases likelihood of becoming teachers (1.1pp), managers (0.9pp), and agents (0.5pp), decreases likelihood of becoming retail clerks (-0.7pp).
  - Increases farm sizes by 4.8% in the areas of construction.
- Despite low returns to education from one room schools, they were an important institution to sort children to other occupations and encourage regional development. Two possible mechanisms:
  1. Higher returns to human capital outside of agriculture.
  2. One room schools reveal underlying ability in non-agricultural sectors.

## Related Literature

- **Early Education in United States & Canada:** Goldin and Katz (1998), Goldin (1998), Goldin and Katz (2000), MacKinnon and Minns (2009), Dippel et al. (2020), Schaede (2021), Card et al. (2022a), Card et al. (2022b), Lachanski (2024) Althoff et al. (2025).
  - **Contribution:** Plausibly exogenous availability of schooling.
- **Education & Structural Change:** Caselli and Coleman II (2001), Porzio et al. (2022), Budí-Ors (2023), Gauthier et al. (2025).
  - **Contribution:** Causal evidence that education reallocates labour away from agriculture.
- **Western Settlement and Development:** Mattheis and Raz (2019), Smith (2022), Leonard and Kogelmann (2022), French (2022), Nagy (2023), Bagagli (2023).
  - **Contribution:** First to geolocate entire rural populations.
- **Census Linking:** Abramitzky et al. (2012), Antonie et al. (2014), Abramitzky et al. (2014), Feigenbaum (2016), Antonie et al. (2020), Abramitzky et al. (2020), Bailey et al. (2020), Price et al. (2021) Helgertz et al. (2022), Feigenbaum et al. (2023), Buckles et al. (2023), Abramitzky et al. (2024).
  - **Contribution:** First to link Canadian census in 20th century, including waves of prairie census.

## Background & Data

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## Schools in the Prairies

- The early need for schools was filled by single one room schoolhouses which covered Alberta, Saskatchewan, and Manitoba.
- The rollout of these schools was not uniform and required a petition by the community to the provincial government. ▶ Details
  - Created a logistical problem in the 1910's as new schools were needed to literally fill in the gaps between existing schools.
- Schools would accommodate the local religion, with no difference in teacher certification by religion.
- Attendance by children was inconsistent.
  - *"Parents who keep children out of school because of distance or danger, appear to have no hesitation in sending the boys to market"* - A. Kennedy, Inspector of Schools, 1910
  - ▶ More

- Historical Canadian Census: 1901, 1906 (Prairies), 1911, 1916 (Prairies), 1921, 1926 (Prairies), 1931.
  - Primary focus: 1906, 1911, and 1931.
- Historical school locations and founding dates from 1871 to 1912. ▶ Details
- Historical railroad line locations.
- FAO estimates for agricultural yields.
- ▶ Other Data

# Geolocation

- The widespread use of the survey boundaries as addresses is unique to the Canadian prairies. [▶ Details](#)
- In other projects (UK, US) it's typical to be able to locate 20-30% of rural households, and then only to the nearest town.
- This project: Geolocate 72% of rural households to the section number (1 mile square) in 1906 census, 84% in 1921, and 71% in 1931. [▶ Household Rates](#)
- Schools are located in a similar fashion, with some geolocations being less precise.
  - I'm able to locate 91.8% of schools relative to the number listed as operating.[▶ School Counts](#)

# An Example Family

		NAME AND RESIDENCE.					
Dwelling house Family, household or institution	Number of each person in family, household or institution	NAME			PLACE OF ABODE.		
		of each person in family, household or institution.			(On rural localities give parish, section, township, range and section. In cities, towns and villages, give street and number of dwelling.)		
		Section	Township	Range	Mari-	Municipality	
		NOM ET RÉSIDENCE.					
Habitation Famille, ménage ou institution	Nombre de chaque personne dans la famille, le ménage ou l'institution.	NOM			DOMICILE.		
		de chaque personne dans la famille, le ménage ou l'institution.			(Dans les localités rurales, spécifier la paroisse, la section, le rang et la section. Dans les villes, les villes et les villages, spécifier la rue et le numéro de la maison.)		
1	2	3	4	4a	4b	4c	5
53185	Dzadysilho	5-4013					
	Zelka	" "	" "	" "	" "	" "	
	Nicla	" "	" "	" "	" "	" "	
	Mike	" "	" "	" "	" "	" "	
	John	" "	" "	" "	" "	" "	
	Dorates	" "	" "	" "	" "	" "	
	Mary	" "	" "	" "	" "	" "	
	Petra	" "	" "	" "	" "	" "	

# Family and Surrounding Township



## Canadian Census Linking

- Need to link census observations in 1906 with later waves in 1911 and 1931.
  - Unlike the US there are no standard set of links available for these census waves.
- Implement the MLP method used at IPUMS to link US census records.
  - Use US training data, training on the set of features that can be recreated with the Canadian data.
  - Match 26% of children from 1906-1911, and 16% from 1906-1931.
    - ▶ Rates
    - ▶ Methodology
  - Lower than MLP results for the United States (46.3%), similar to ABE (26.5%)
  - Expected as the Canadian immigration is much higher at this time.

## Bias From Linking - Makes Estimates More Conservative

- Earlier linking methods used on immutable characteristics.
- Newer methods like MLP and the newest ABE use mutable characteristics to differentiate observations that are otherwise observationally equivalent (Abramitzky et al., 2024).
  - John Smith in Ontario in 1911 is more likely to be in Ontario in 1921 than in Nova Scotia.
  - Get linking performance improvements for shorter distances [▶ MLP-ABE Distance](#).
  - Bias works against effects on occupation changes and migration.
- More likely to match people who are younger , men , wealthier , and native to Ontario or the United Kingdom .

## Motivating Facts

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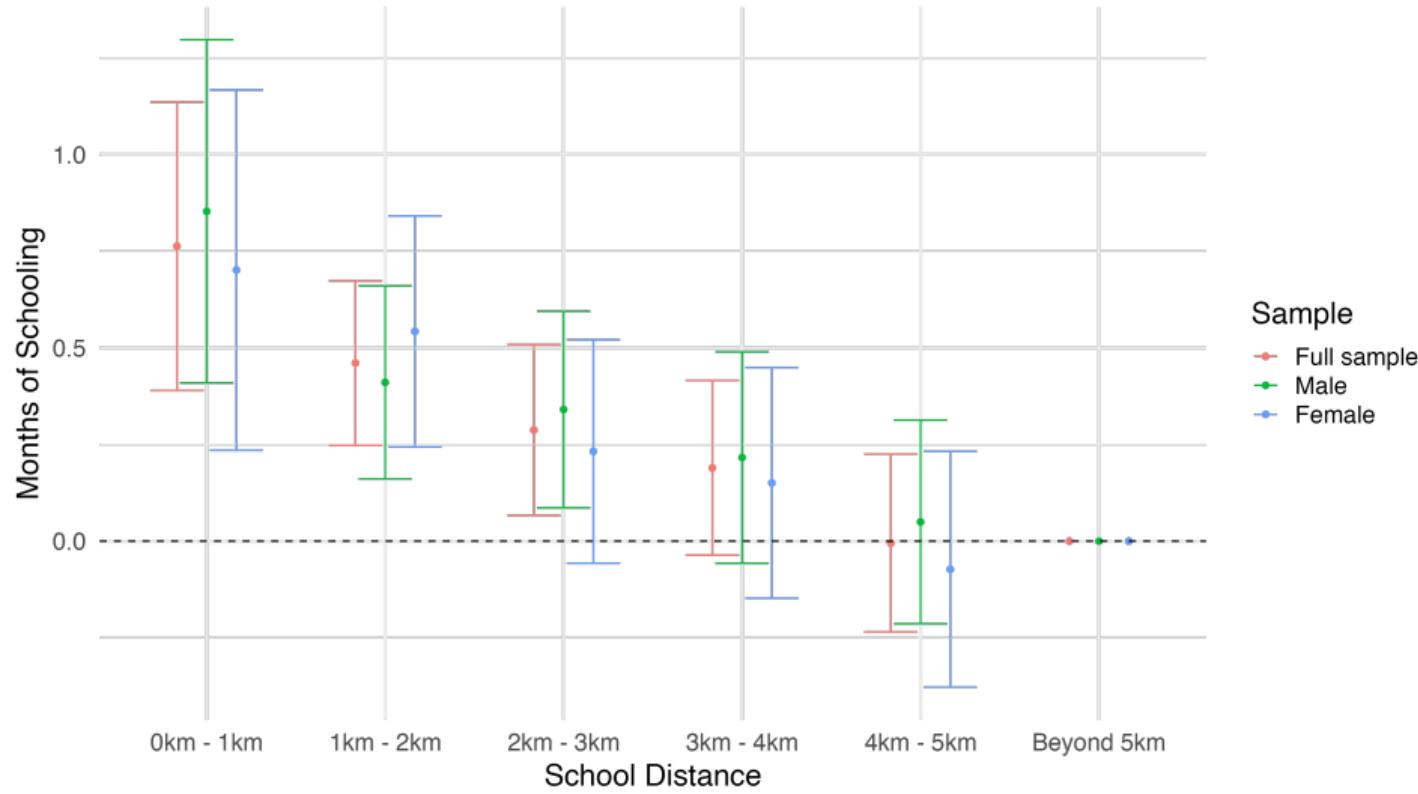
## How does distance affect schooling?

- Distance is mentioned repeatedly as a barrier to school attendance, and as one of the arguments for one room schools.
  - School locations don't overlap with railways, towns, post offices, or other amenities.  
▶ Map
- Link 1906 (have locations) with 1911 (have school attendance), restrict sample to households that didn't move.
- Estimate effect of distance on school attendance in 1km bins:

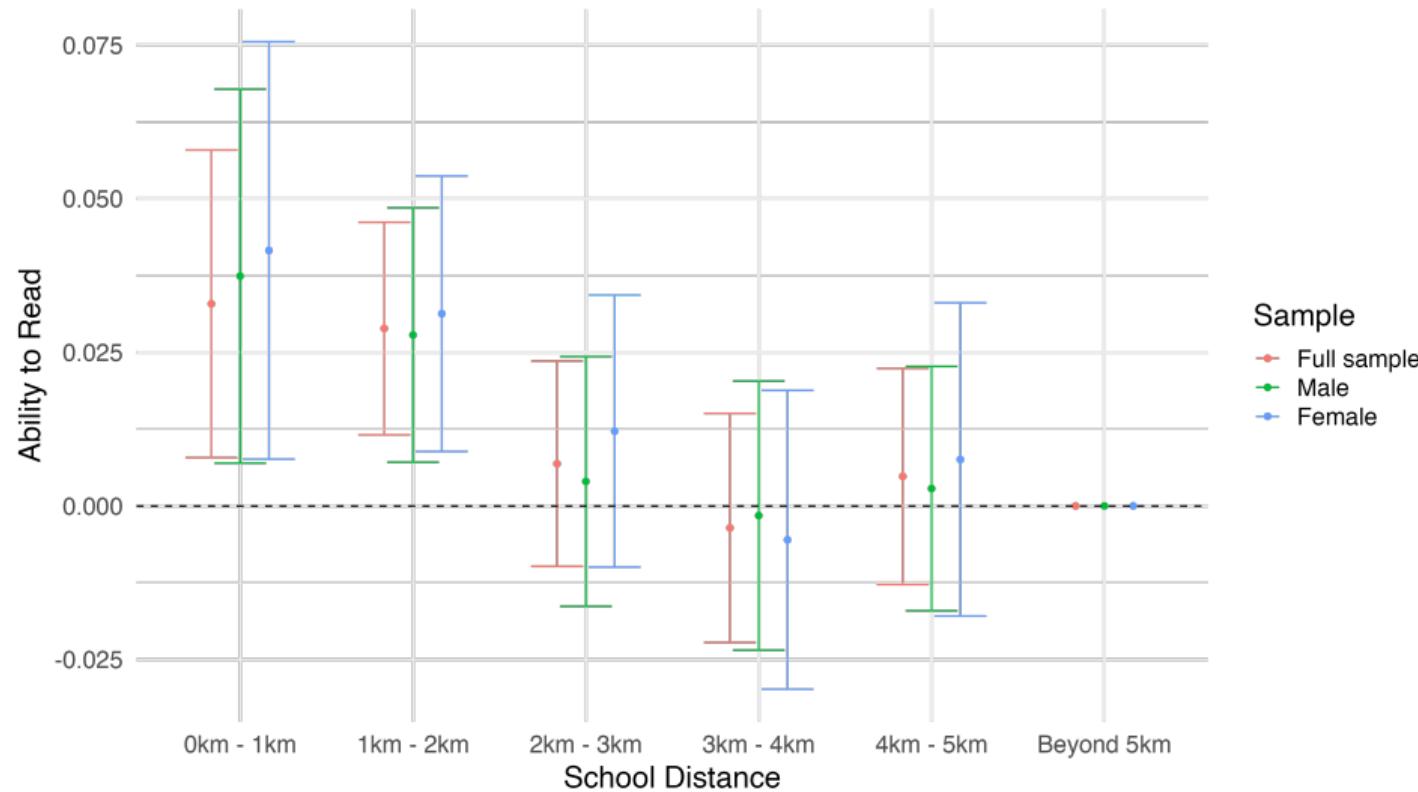
$$y_i = \sum_{k=1}^5 \beta_k \mathbb{1}\{d \in k_{1\text{km}}\} + X_i + p_i + \epsilon_i \quad (1)$$

- Include sex by age by birth order, birthplace, and enumeration subdistrict fixed effects, family economic and railway distance controls.

# Schooling vs Distance



# Ability to Read vs Distance



## Distance Decreases Schooling

- Similar effects on both intensive and extensive margins  .
- In regressions, find an elasticity of  $-0.039$  between log months of school and log distance to school, and coefficient of  $-0.027$  between likelihood of attending school and distance to school.
- Note that the effects dissipate to zero around 5km mark.
  - Policy was to make school districts no larger than 20 square miles, or  $\approx 4\text{km}$  in radius.

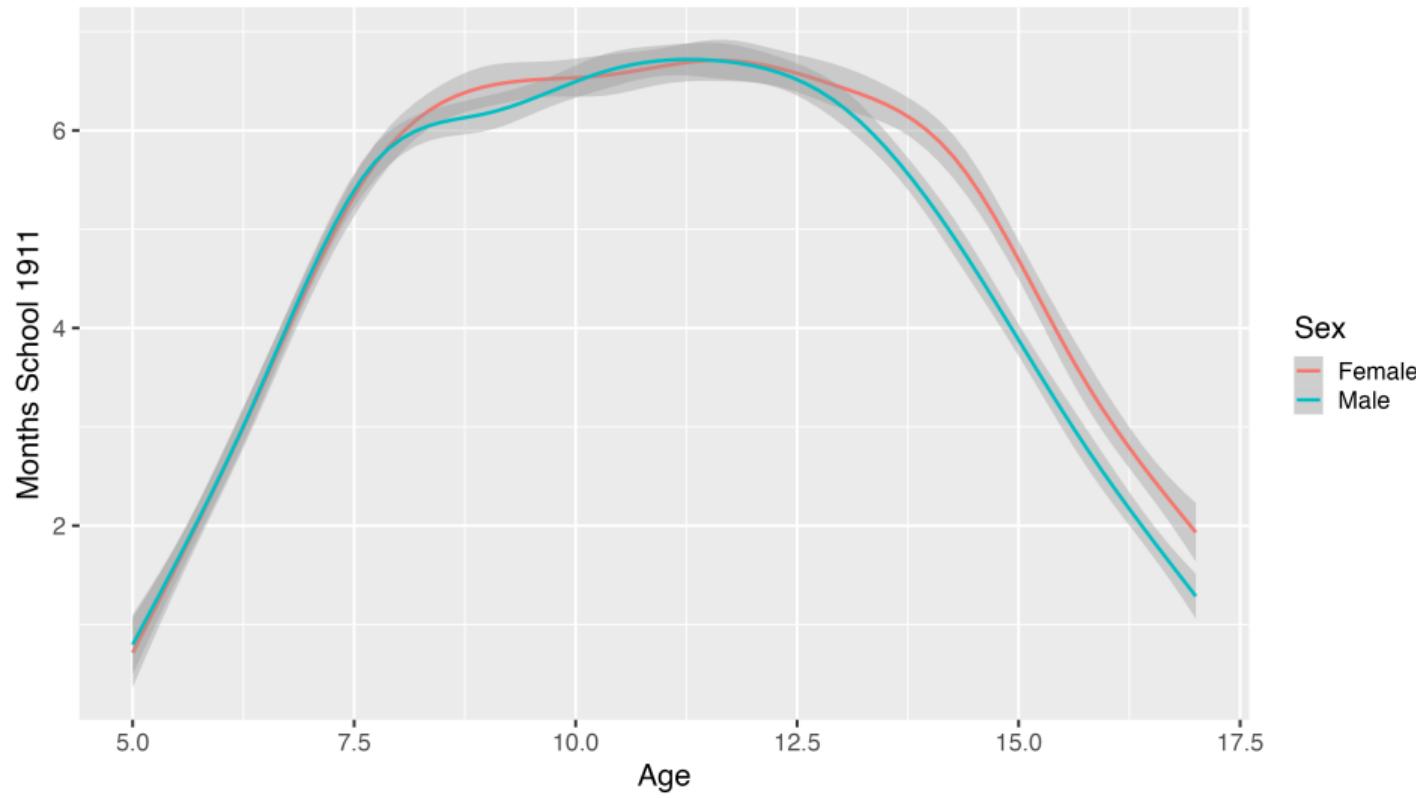
## **Impact of School Construction on Later Life**

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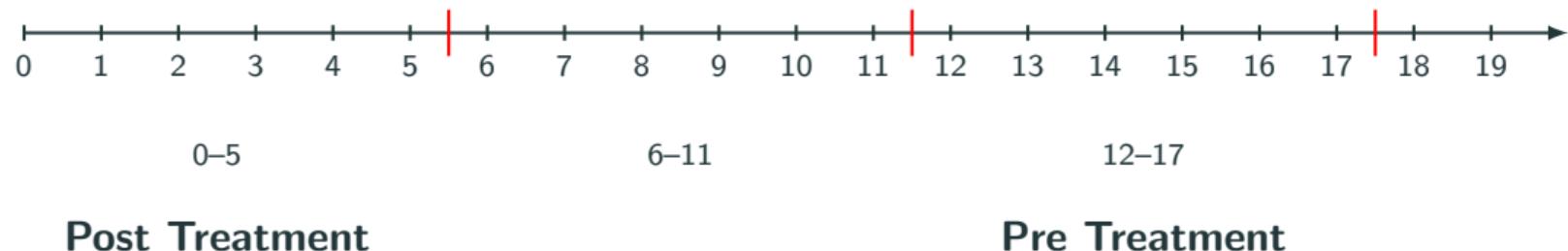
## Difference in Differences

- School location is likely endogenous with a variety of factors.
- Solution: Exploit the timing of schools, comparing children who had schools constructed nearby (treated) to those who didn't (control) in younger versus older cohorts as in Duflo (2001).
  - Difference: treatment is defined at the individual instead of region.
- School construction was rapid, with a new school district being established every day by 1905 [▶ Graph](#).
- Assumptions:
  1. Children do not benefit from school construction after a certain age.
  2. Parents cannot manipulate when the school is constructed relative to the age of their children.
- Link 1906 census (boys with school distances) to 1931 (adults).

## Schooling vs Age (Raw Data)

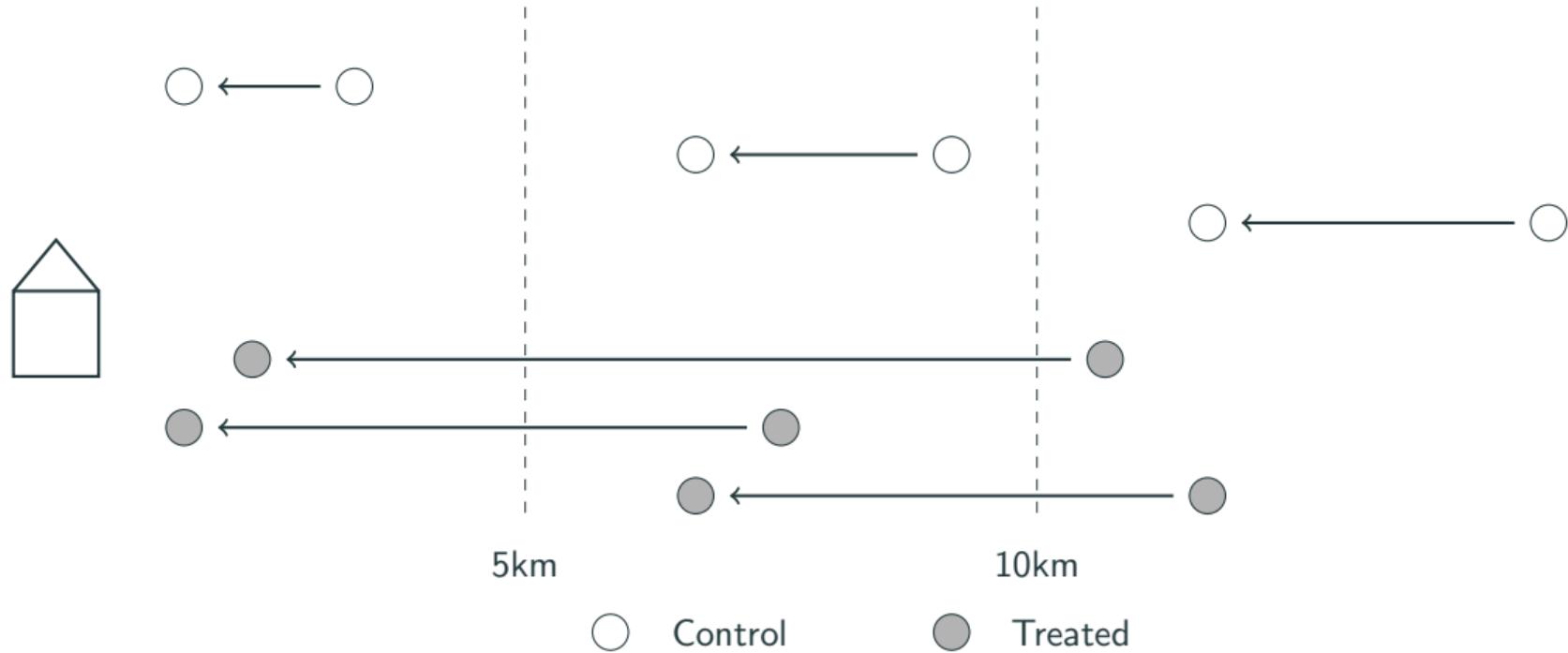


# Difference in Differences



- Comparison between those in the 0-5 cohort ( $t_c = \mathbb{1}\{\text{age}_{1906} \leq 5\}$ ) in 1906, and those who are 12-17 in 1906.
- Considered treated if the distance to school decreased by 5km or more between 1900 to 1905, and the final distance is within 10km ( $s_i = \mathbb{1}\{\Delta dist_{i,1900,1905} > 5\text{km} | dist_{i,1905} < 10\text{km}\}$ ).
- Include geographic fixed effects  $\alpha_d$ , probability of link  $p_i$  and weight by observations by  $p_i$ .

## Treated vs Control



Dependent Variable:	Family in Ag				
Model:	(1)	(2)	(3)	(4)	(5)
School Constr. × Cohort	-0.0409*** (0.0151)	-0.0317** (0.0154)	-0.0331** (0.0154)	-0.0342** (0.0153)	-0.0346** (0.0153)
School Constr.	0.0112 (0.0162)	0.0057 (0.0164)	0.0060 (0.0164)	0.0047 (0.0165)	0.0049 (0.0163)
Cohort	0.0394*** (0.0091)	0.0452*** (0.0095)	0.0626*** (0.0093)	0.0626*** (0.0092)	0.0618*** (0.0092)
Log RR Distance (1906)					0.0210*** (0.0045)
Link Prob.	0.4399*** (0.0245)	0.4261*** (0.0245)	0.4149*** (0.0244)	0.4146*** (0.0245)	0.4156*** (0.0244)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes	Yes
Fam. Bpl.		Yes	Yes	Yes	Yes
Homestead			Yes	Yes	Yes
Livestock			Yes	Yes	Yes
Ag. Productive.				Yes	Yes
Observations	14,944	14,944	14,944	14,944	14,944
R <sup>2</sup>	0.07873	0.08514	0.09582	0.09796	0.09986
Dependent variable mean	0.69279	0.69279	0.69279	0.69279	0.69279

Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses  
Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

Dependent Variable:	Log Income				
Model:	(1)	(2)	(3)	(4)	(5)
School Constr. × Cohort	0.1012* (0.0566)	0.1108* (0.0581)	0.1160** (0.0587)	0.1213** (0.0590)	0.1226** (0.0591)
School Constr.	0.0130 (0.0541)	0.0083 (0.0552)	0.0052 (0.0558)	0.0058 (0.0549)	0.0050 (0.0546)
Cohort	-0.3057*** (0.0276)	-0.2693*** (0.0294)	-0.2639*** (0.0306)	-0.2594*** (0.0307)	-0.2585*** (0.0308)
Log RR Distance (1906)					-0.0318** (0.0123)
Link Prob.	-0.2310*** (0.0620)	-0.1887*** (0.0620)	-0.1915*** (0.0618)	-0.1966*** (0.0620)	-0.2018*** (0.0621)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes	Yes
Fam. Bpl.		Yes	Yes	Yes	Yes
Homestead			Yes	Yes	Yes
Livestock			Yes	Yes	Yes
Ag. Productive.				Yes	Yes
Observations	5,750	5,750	5,750	5,750	5,750
R <sup>2</sup>	0.15606	0.16376	0.16727	0.17174	0.17288
Dependent variable mean	6.5231	6.5231	6.5231	6.5231	6.5231

*Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses*  
*Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1*

## Migration Out of Agriculture

- Treated cohorts are also likely to live further away from home ➔.
- Some evidence that they are more likely to be employees ➔.
- These results hold while varying the treatment cutoff ➔, how treatment is defined ➔, and the cohorts used comparison ➔.
- Results are concentrated in those that are furthest from the railway. ➔ Robustness
- If people are migrating out of agriculture, what are they doing?

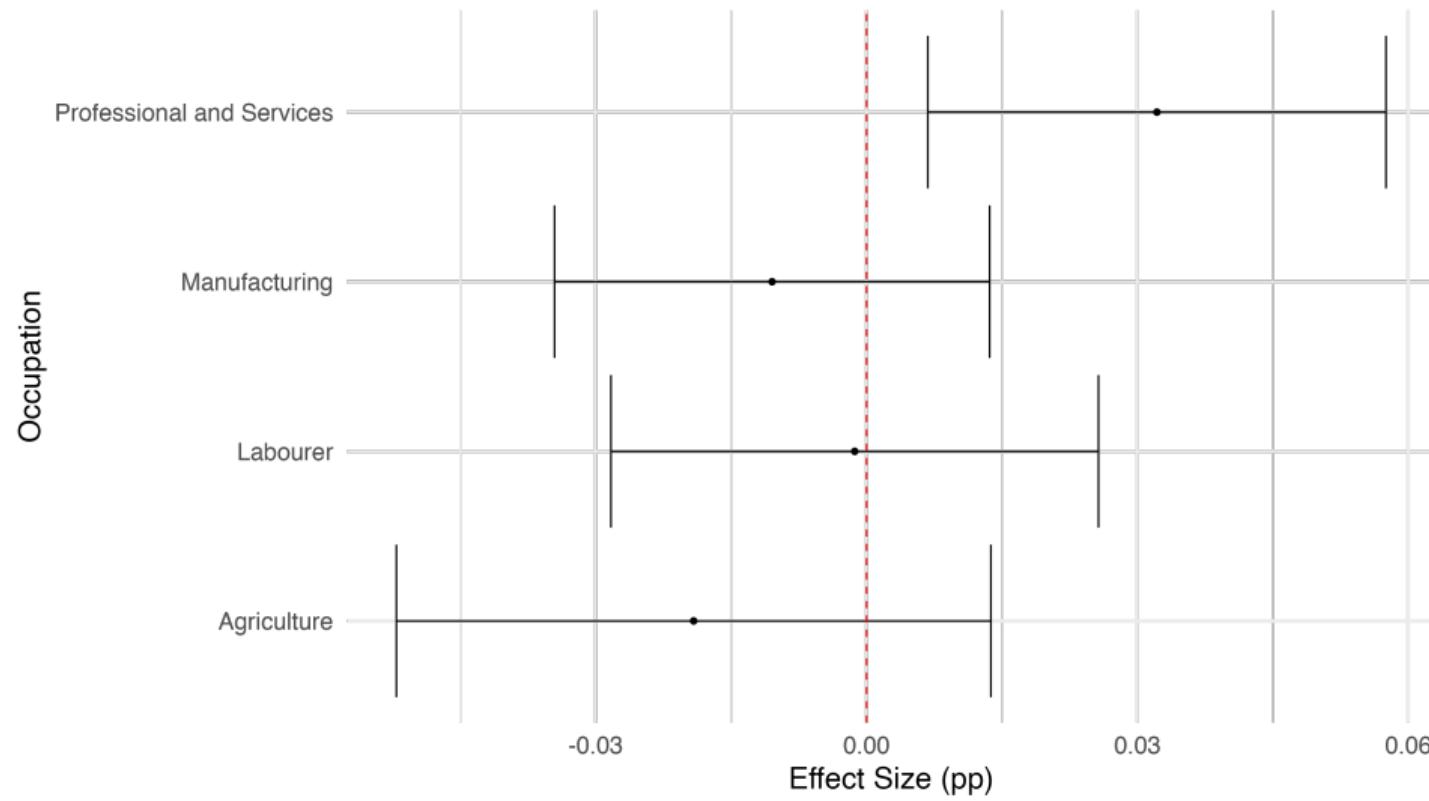
## Changes in Occupations

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## Switch to Services

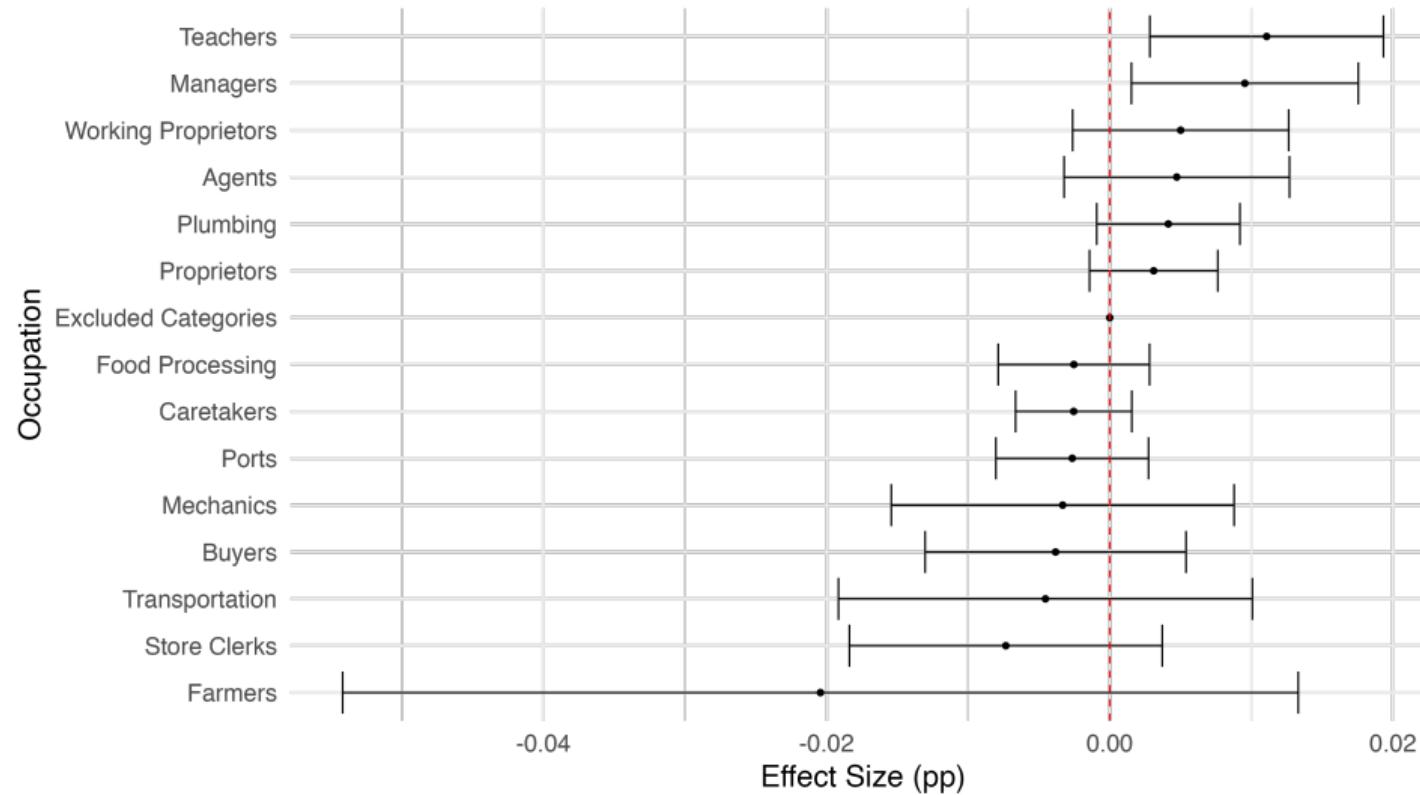
- Can use occupation codes to determine what types of occupations people are switching into.
- Start by looking at four primary industries: Agriculture, Manufacturing, Services, and General Labour.
- Then further divide into 2 digit OCCHISCO codes.

# Primary Industries: Bypassing Manufacturing



# Switching to High Skill Services

All



## But in Different Industries than Urbanites

Agents who grew up in Rural Areas

Industry	n
Unknown	131
Grain Elevator	56
Insurance	47
General	12
Real Estate	9
Steam Railway	9
General Farm	6
Lumber Mill	5
Retail Hardware	5
Steam Railroad	5
Other	28

Agents who grew up in Urban Areas

Industry	n
Unknown	97
Insurance	29
Real Estate	10
Grain Elevator	7
Steam Railway	5
Mill Farm	4
Wholesale Grocery	3
Financial	2
Automobile	1
Barber Shop	1
Other	15

## **Land Consolidation**

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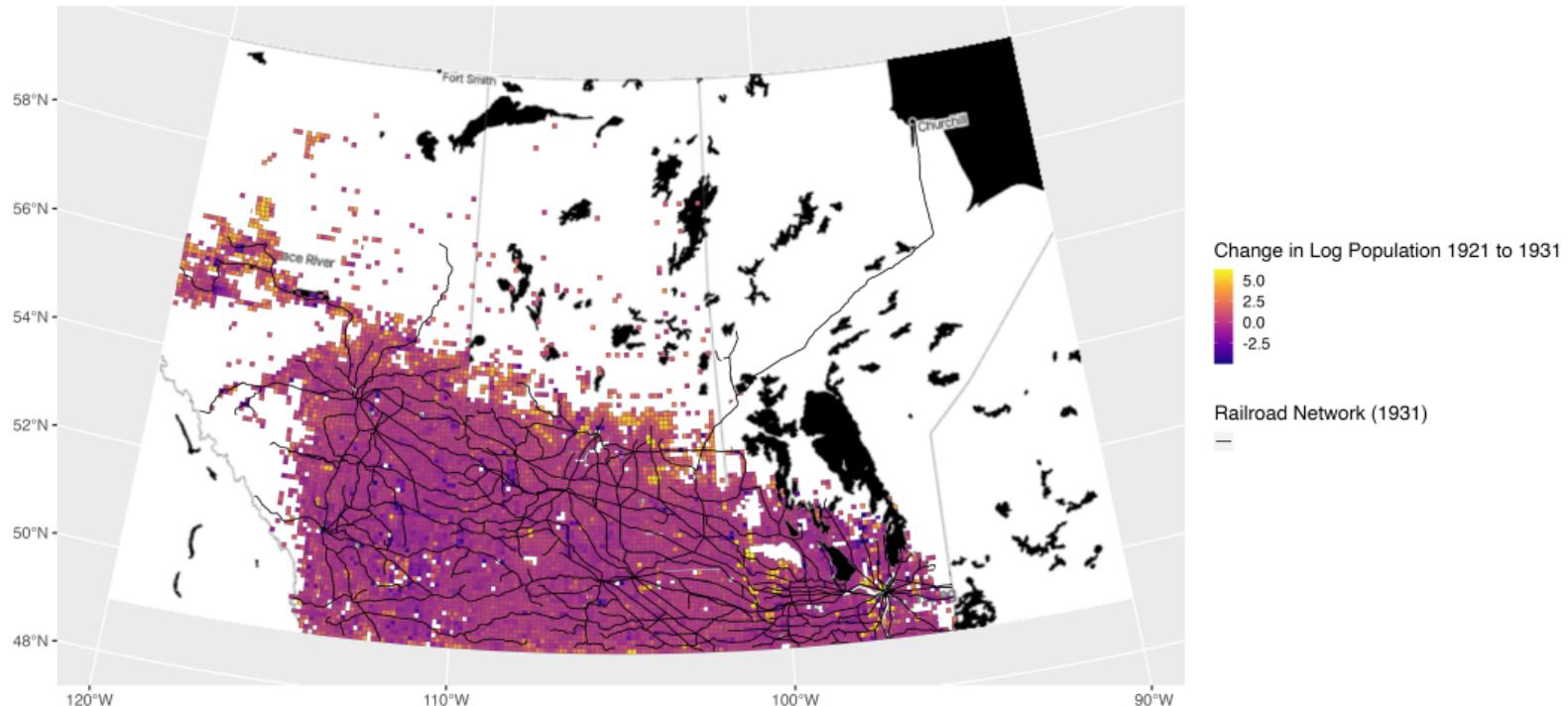
## Land Consolidation

- If people are leaving agriculture and farms are primarily owner operated, then we would expect farm sizes to increase.
- Don't observe farm sizes directly, only the location of the farm.
- Can infer farm sizes by looking at the density of neighbors.
  - A farmer with fewer neighbors is likely to have a larger farm.
- Imperfect measure with a lot of noise.
  - Validate by looking at correlations with other farm characteristics and persistence in farm size.

▶ Characteristics

▶ Persistance

# Population Change from 1921 to 1931



## Land Consolidation - Triple Differences

- Can't use linked census waves because cohort timing doesn't apply to land.
- Instead, look at the entire set of farms observable in 1906, 1921, and 1931, and use the set of schools open in 1912, with opening dates between 1895 and 1912.
- Define a close school if it was constructed within 5km of the parcel ( $p_i = \mathbb{1}\{dist_i < 5\text{km}\}$ ).
- Define early school construction as areas that got a school before 1906 ( $s_j = \mathbb{1}\{date_j < 1906\}$ ).
- Use 1906 as the reference period for 1921 and 1931 ( $T_t$ ).
- Include school fixed effects  $\alpha_j$

$$y_{ijt} = \beta p_i s_j T_t + \text{Interactions}_{i,j,t} + \alpha_j + \epsilon_{i,j,t} \quad (3)$$

Dependent Variable:	Ln Farm Area				
Model:	(1)	(2)	(3)	(4)	(5)
Early $\times$ Close $\times$ Year = 1931	0.0612*	0.0415	0.0453*	0.0477*	0.0484*
	(0.0325)	(0.0260)	(0.0258)	(0.0260)	(0.0260)
Early $\times$ Close $\times$ Year = 1921	0.0223	0.0104	0.0117	0.0104	0.0106
	(0.0312)	(0.0258)	(0.0256)	(0.0257)	(0.0257)
Early $\times$ Close	-0.0446*	-0.0298	-0.0343	-0.0342	-0.0340
	(0.0265)	(0.0220)	(0.0219)	(0.0221)	(0.0221)
Homestead			-0.2121***	-0.2135***	-0.2135***
			(0.0050)	(0.0050)	(0.0050)
RR Dist in Year				0.0407***	0.0407***
				(0.0035)	(0.0035)
Family Size					-0.0006
					(0.0005)
Fam. Adult Males					0.0045***
					(0.0016)
Other Terms	Yes	Yes	Yes	Yes	Yes
School District		Yes	Yes	Yes	Yes
Observations	439,027	439,027	439,027	439,027	439,027
R <sup>2</sup>	0.00614	0.13406	0.15113	0.15336	0.15384

Clustered (Enum. Sub.) standard-errors in parentheses

Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

## Land Consolidation - Not Driven by Settlement

- This result is robust to changing the cutoff boundary and early versus late year selection. ▶ Boundary ▶ Year
- Importantly find the same result if looking only at farms in 1921 and 1931, after the majority of settlement occurs. ▶ Year
  - Possible concern is that 1906 is in the middle of the settlement period so could be picking up general geographic expansion.

## Conclusion

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## Conclusion

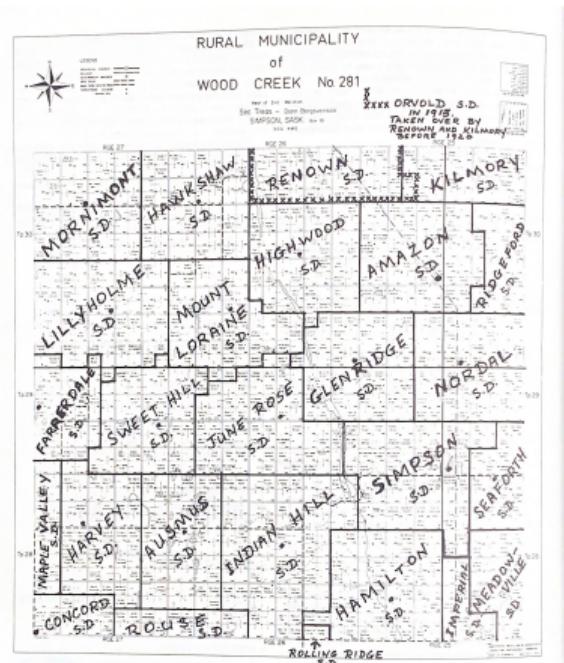
- This project has focused on the role that one room schools play in the development of rural areas.
- Exposure to these schools lead to agricultural outmigration, and switching towards high skill services.
- This encouraged the consolidation of land into larger farms.
- Want to think about mechanisms next. Two possibilities:
  1. Higher returns to human capital outside of agriculture.
  2. One room schools reveal underlying ability in non-agricultural sectors.

**Thank you!**

- Establishing a school district required petitioning the provincial government and a ratepayer vote.
  - Sometimes, votes would fail due to fear of higher taxes.
- If a school district was formed, the location of the school would be located in the middle of the district. ▶ Example
  - Some minor exceptions if land could not be acquired.
  - Settlers could not anticipate the location of the school when settling.
- Distance was a key concern for school attendance.
  - Students were exempt from mandatory schooling in Alberta (1910) if they lived further than 2 miles from the nearest school.
  - Same in Saskatchewan after 1917, with a distance of 2.5 miles.

# School Board Layout

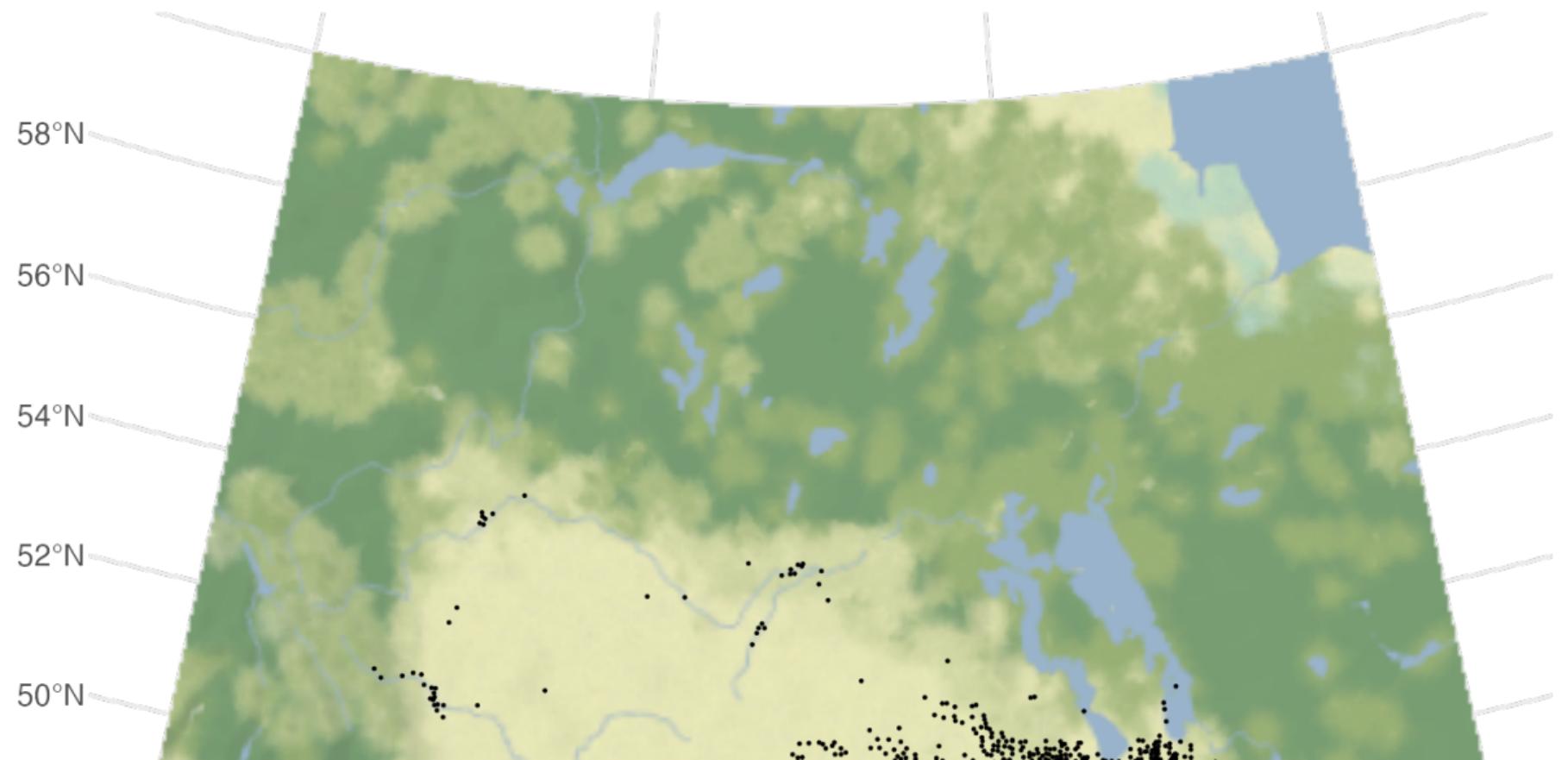
▶ Back



- "*The attendance in rural districts is not so regular as one would expect or could hope for. Undoubtedly, it is in part due to the scarcity of farm help and the long distances some children have to go; but I fear it is also caused by a lack of interest on the part of many parents*", H.H. Smith, Inspector of Schools, Saskatoon, December 31, 1909
- "*When a boy is old enough to work an outfit on a farm, his school days are over*", John S. Huff, Davidson, Saskatchewan., March 1st, 1912

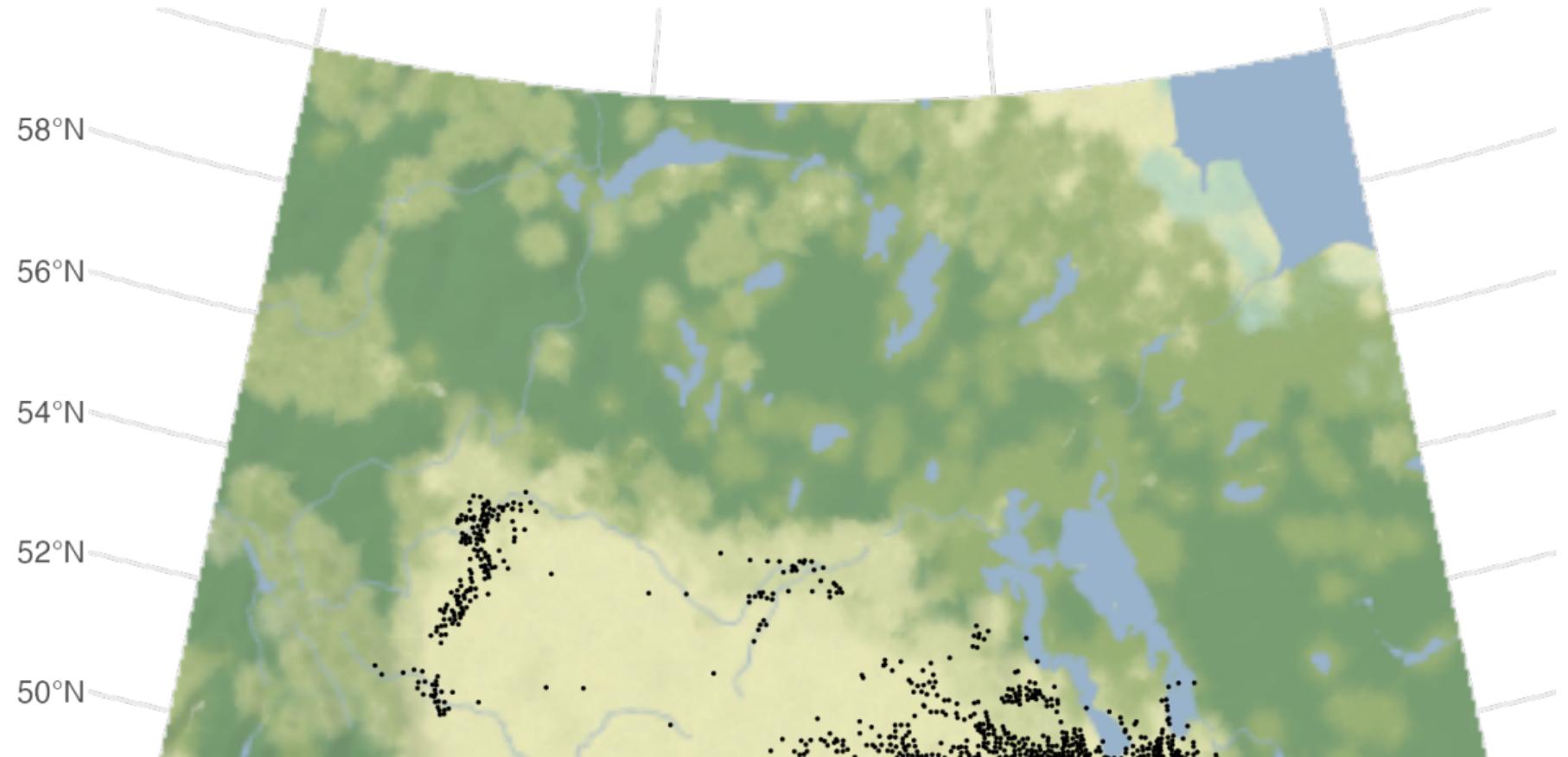
- Panel of school openings from 1871 to 1912 covering Alberta, Saskatchewan and Manitoba.
- Merge four independent sources of data for schools in the prairies:
  - Alberta: Glenbow Museum Archives
  - Saskatchewan: One Room School Project, Annual reports of the Department of Education
  - Manitoba: Manitoba Historical Society
- Saskatchewan data doesn't include the opening date for each school, but can infer date using Alberta school opening dates.
  - Alberta and Saskatchewan were both part of the Northwest Territories until 1905, so share a school numbering system until that time.

# Schools 1890



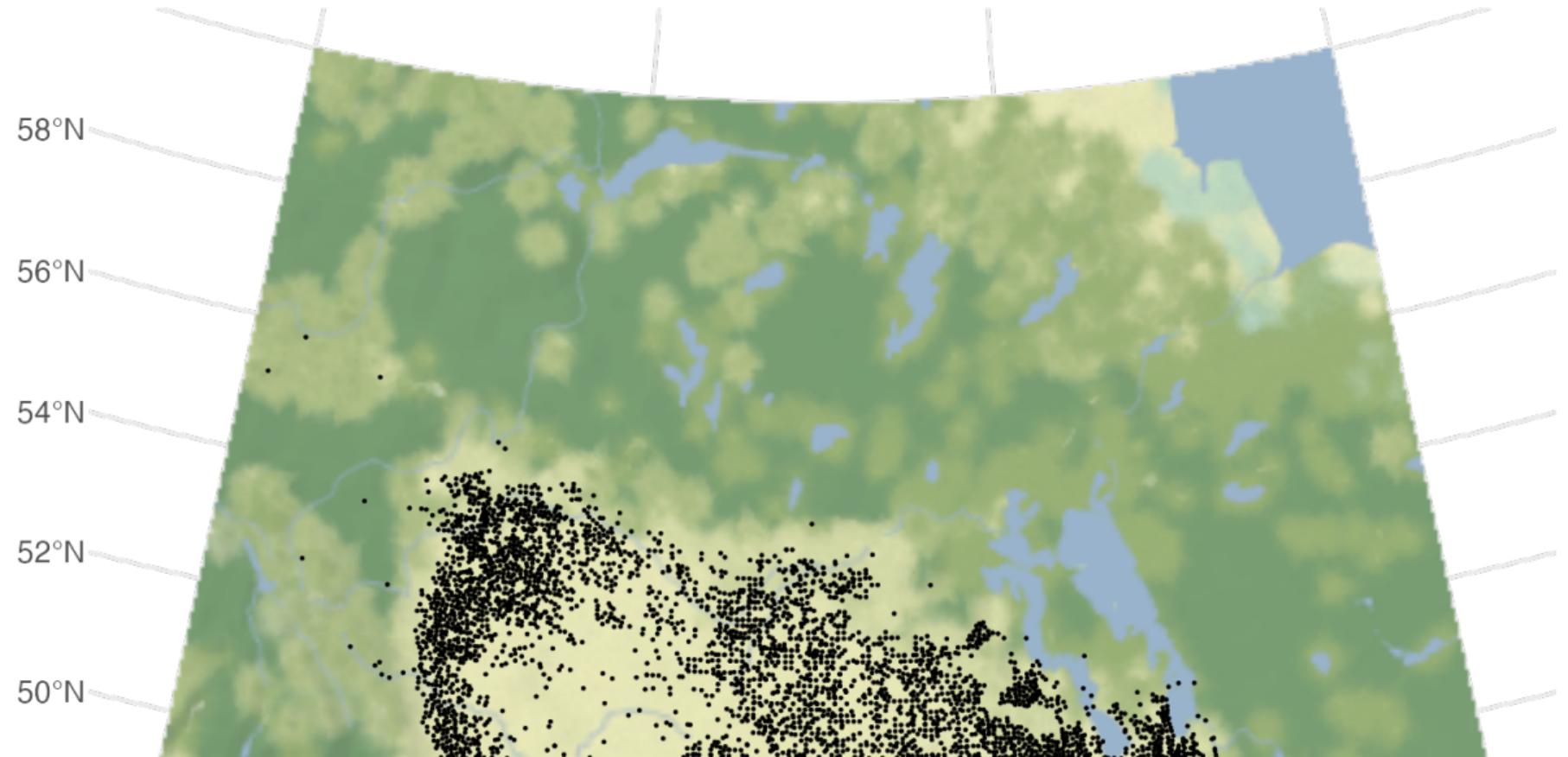
# Schools 1900

[Back to Data](#)



# Schools 1910

[Back to Data](#)



- Cummins Maps. [Example](#)
  - Alberta, Saskatchewan, and Manitoba property ownership maps (approximately 1915 - 1930).
  - Some have been digitized, currently negotiating to get labelled versions for Manitoba and Saskatchewan.
- Land Grants of Western Canada, 1870-1930, CPR Land Sales Records, Hudsons Bay Company Land Records.
- Massey-Harris dealer records.

# Household Geolocation Rates

[Back to Geolocation](#)

**Table 1:** Geolocation rates by census year for Alberta, Saskatchewan, and Manitoba. Exact geolocations are at the section level or in CSDs that are less than 1.5 miles in area. General geolocations are those that are at the township level or finer. CSD geolocations are at the CSD level. Splitting the geolocation rate by urban and rural households reveals that the majority of geolocations in both census waves are for the rural households.

	Total	Exact	General	CSD	Exact (Rural)	Exact (Urban)
<b>1906</b>	100.00	55.57	84.41	100.00	71.94	0.00
<b>1921</b>	100.00	65.97	69.61	100.00	83.89	36.35
<b>1931</b>	100.00	42.76	86.57	100.00	70.81	0.00

# School Geolocation Rates

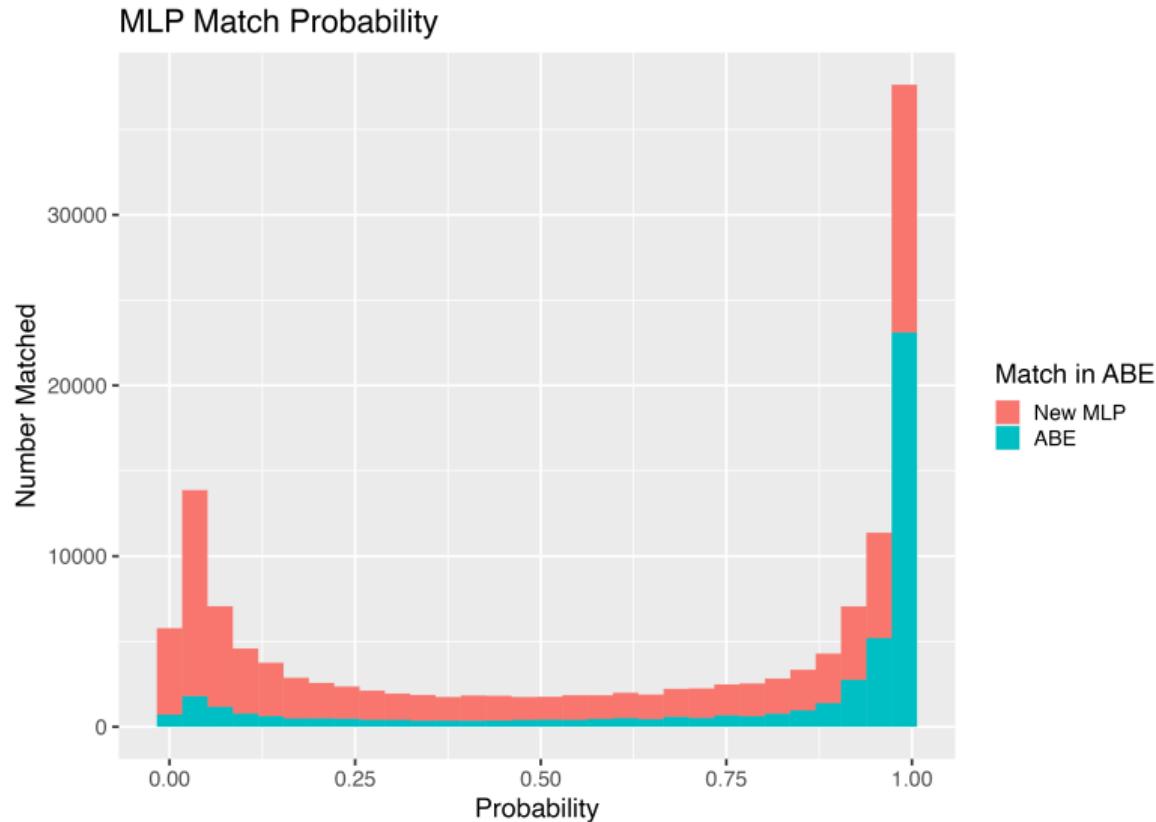
[Back to Geolocation](#)

**Table 2:** Summary statistics by province for the number of schools in 1911. The first row lists the number of schools reports in the 1911 report from the department of education for Alberta (Government of Alberta, 1912), Saskatchewan (Saskatchewan Department of Education, 1906, 1907, 1908, 1910, 1911, 1912, 1914), and Manitoba (Government of Manitoba, 1911, 1912). This figure includes both public and separate schools in Alberta and Saskatchewan while the data contain only public schools, so is a slight overestimate. The second row, "Observed" lists the number of schools I find in the data in 1911, which requires that they have an opening date before 1911. Any schools with an unknown opening date are dropped. "Geolocated" reports the number of schools, of those which are observed, which I am able to geolocate. "Geolocated (Section)" reports the number of schools where the geolocation is at the section level (1 mile square) or finer.

Type	Total	Alberta	Saskatchewan	Manitoba
Actual	5955	1784	2573	1598
Observed	5618	1750	2370	1498
Geolocated	5472	1750	2227	1495
Geolocated (Section)	4141	1619	1027	1495

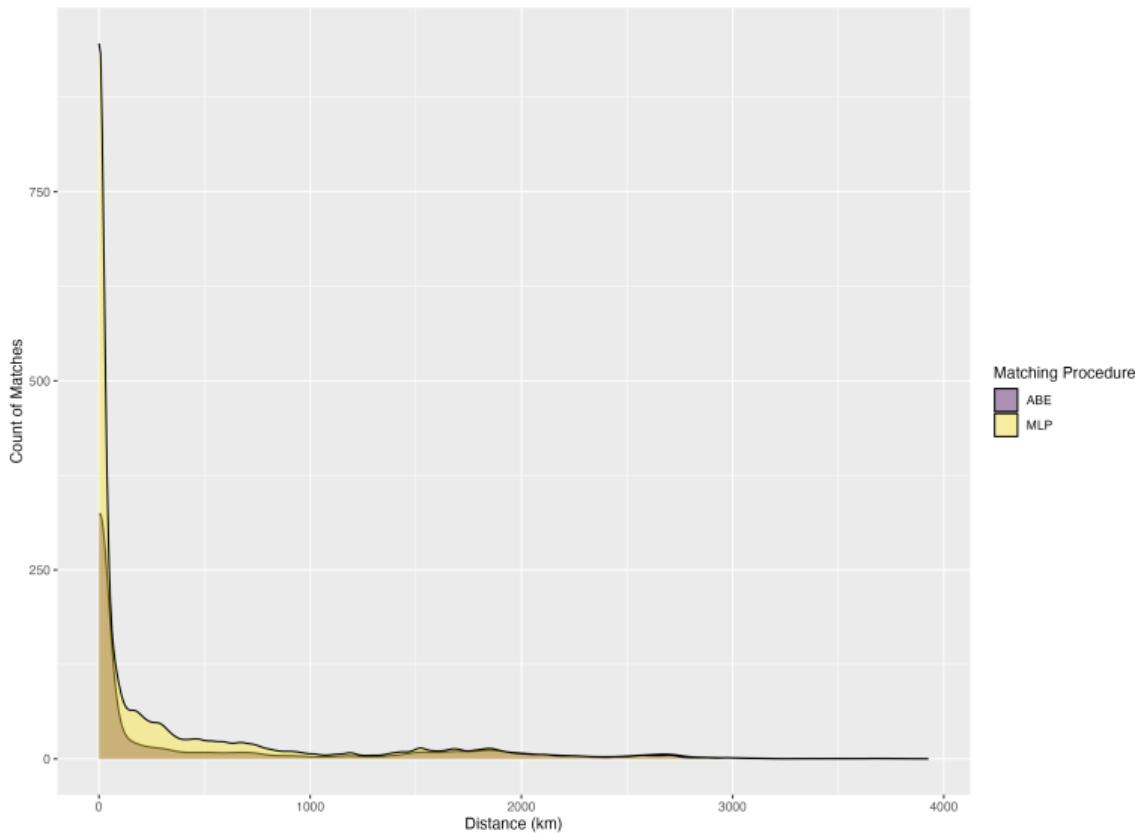
# Link Probability MLP 1906-1921

[Back to Linking](#)



# Distance from 1906-1911 by Linking Method

[Back to Linking](#)



# Matching Rates

[Back to Linking](#)[By Province](#)[Linking Counts](#)

	Total	Matched	Children	Adults	Male	Female
<b>1906-1911</b>	100.00	22.80	26.69	21.43	26.75	17.76
<b>1906-1921</b>	100.00	19.73	20.59	20.12	25.62	12.02
<b>1906-1931</b>	100.00	16.20	16.27	16.91	23.02	7.18
<b>1911-1921</b>	100.00	19.95	23.06	18.25	24.69	14.67
<b>1911-1931</b>	100.00	13.37	13.92	13.19	18.86	7.21

- MLP works in two stages:
  1. Generate all possible matches between two years. Use logit/machine learning to score matches. Take matches above a minimum threshold, and above a threshold for the next best match.
  2. Use these matches to identify families, then do a second stage of matching within families.
- Currently using the training data from the MLP replication package to train a model for the Canadian data.

# Matching Counts

[Back to Linking Rates](#)[Back to Linking](#)

	Total	Matched	Children	Adults	Male	Female
<b>1906-1911</b>	802940	183049	80170	102879	122854	60112
<b>1906-1921</b>	802940	158416	61840	96576	117681	40694
<b>1906-1931</b>	802940	130060	48864	81196	105740	24303
<b>1911-1921</b>	7197003	1436085	638288	797797	940674	495213
<b>1911-1931</b>	7197003	961982	385183	576799	718432	243467

# Match Rate by Province

[Back to Linking Rates](#)[Back to Linking](#)

	Canada	AB	SK	MB	ON	QC
<b>1906-1911</b>	22.80	22.95	21.36	23.75		
<b>1906-1921</b>	19.73	18.98	19.17	20.51		
<b>1906-1931</b>	16.20	16.01	16.05	16.40		
<b>1911-1921</b>	19.95	18.22	18.41	19.92	21.21	18.19
<b>1911-1931</b>	13.37	14.40	14.28	14.74	13.60	12.21

# Match Rate by Age

[Back to Linking Bias](#)

Model:	Dependent Variable: Matched				
	(1) 1906-1911	(2) 1906-1921	(3) 1906-1931	(4) 1911-1921	(5) 1911-1931
Age = 1	0.0730***	0.0786***	0.0673***	0.0686***	0.0367***
Age = 2	0.0074*	0.0155***	0.0583***	-0.0178***	0.0132***
Age = 3	-0.0355***	0.0100***	0.0521***	-0.0302***	0.0088***
Age = 4	-0.0083**	0.0209***	0.0443***	0.0052***	0.0201***
Age = 6	-0.0268***	-0.0569***	-0.0872***	-0.0325***	-0.0582***
Age = 7	-0.0458***	-0.1299***	-0.1473***	-0.0806***	-0.1143***
Age = 8	-0.0642***	-0.1663***	-0.1502***	-0.1197***	-0.1272***
Age = 9	-0.1267***	-0.1751***	-0.1359***	-0.1434***	-0.1211***
Age = 10	-0.1999***	-0.1869***	-0.1474***	-0.1499***	-0.1134***
Age = 11				-0.1073***	-0.0706***
Age = 12				-0.1558***	-0.1033***
Sex	0.1307***	0.1465***	0.1467***	0.1197***	0.1084***
Family Size	0.0041***	0.0010***	0.0001	$-5.8 \times 10^{-5}$	$-7.95 \times 10^{-5}***$
Birthplace	Yes	Yes	Yes	Yes	Yes
Relation	Yes	Yes	Yes	Yes	Yes
Enum. Subdist.	Yes	Yes	Yes	Yes	Yes
Homestead	Yes	Yes	Yes		
Milk Cows	Yes	Yes	Yes		
Horses	Yes	Yes	Yes		
Wheat Prod.	Yes	Yes	Yes		
Religion				Yes	Yes

# Match Rate by Relation

[Back to Linking Bias](#)

Dependent Variable:		Matched				
Model:		(1) 1906-1911	(2) 1906-1921	(3) 1906-1931	(4) 1911-1921	(5) 1911-1931
Family Size		0.0043***	0.0013***	0.0004	$-4.14 \times 10^{-5}$	$-6.45 \times 10^{-5}***$
Relation = Boarder		-0.1487***	-0.1326***	-0.1166***	-0.1034***	-0.0951***
Relation = Daughter		-0.1340***	-0.2163***	-0.2192***	-0.1301***	-0.1627***
Relation = Lodger		-0.1398***	-0.1533***	-0.1187***	-0.1049***	-0.0944***
Relation = Other		-0.1284***	-0.1156***	-0.0941***	-0.0972***	-0.0842***
Relation = Sister		-0.1942***	-0.2063***	-0.1794***	-0.1756***	-0.1637***
Relation = Son		0.0110***	0.0118***	0.0425***	0.0319***	0.0435***
Relation = Wife		-0.0379***	-0.0605***	-0.0819***	-0.0285***	-0.0458***
Age	Yes	Yes	Yes	Yes	Yes	Yes
Birthplace	Yes	Yes	Yes	Yes	Yes	Yes
Enum. Subdist.	Yes	Yes	Yes	Yes	Yes	Yes
Homestead	Yes	Yes	Yes			
Milk Cows	Yes	Yes	Yes			
Horses	Yes	Yes	Yes			
Wheat Prod.	Yes	Yes	Yes			
Religion				Yes	Yes	
Observations	430,101	430,101	430,101	1,191,247	1,191,247	
R <sup>2</sup>	0.08044	0.08238	0.09728	0.05538	0.06190	

Clustered (Enum. Subdist.) standard-errors in parentheses

Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

# Match Rate by Livestock

[Back to Linking Bias](#)

Model:	Dependent Variable:		
	(1) 1906-1911	(2) 1906-1921	(3) 1906-1931
Sex	0.1313***	0.1467***	0.1466***
Family Size	0.0041***	0.0011***	$9.13 \times 10^{-5}$
Milk Cows = 1-4	0.0266***	0.0266***	0.0199***
Milk Cows = 5-9	0.0537***	0.0475***	0.0372***
Milk Cows = 10-49	0.0618***	0.0495***	0.0403***
Milk Cows = 50+	0.0173	0.0080	-0.0086
Horses = 1-4	0.0311***	0.0280***	0.0205***
Horses = 5-9	0.0547***	0.0499***	0.0385***
Horses = 10-49	0.0634***	0.0559***	0.0436***
Horses = 50+	0.0406**	0.0231*	0.0205**
Age	Yes	Yes	Yes
Birthplace	Yes	Yes	Yes
Relation	Yes	Yes	Yes
Province	Yes	Yes	Yes
Homestead	Yes	Yes	Yes
Wheat Prod.	Yes	Yes	Yes
Observations	430,101	430,101	430,101
R <sup>2</sup>	0.05674	0.06955	0.08882

Clustered (Enum. Subdist.) standard-errors in parentheses

Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

# Match Rate by Homestead

[Back to Linking Bias](#)

Model:	Dependent Variable:		
	Matched		
	(1) 1906-1911	(2) 1906-1921	(3) 1906-1931
Sex	0.1313***	0.1467***	0.1466***
Family Size	0.0041***	0.0011***	$9.13 \times 10^{-5}$
Homestead = HBC	0.0059	-0.0019	-0.0015
Homestead = Railway	0.0004	0.0018	0.0012
Homestead = School	0.0079	-0.0006	0.0009
Age	Yes	Yes	Yes
Birthplace	Yes	Yes	Yes
Relation	Yes	Yes	Yes
Province	Yes	Yes	Yes
Milk Cows	Yes	Yes	Yes
Horses	Yes	Yes	Yes
Wheat Prod.	Yes	Yes	Yes
Observations	430,101	430,101	430,101
R <sup>2</sup>	0.05674	0.06955	0.08882

*Clustered (Enum. Subdist.) standard-errors in parentheses*

*Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1*

# Match Rate by Province

[Back to Linking Bias](#)

Model:	Dependent Variable: Matched				
	(1) 1906-1911	(2) 1906-1921	(3) 1906-1931	(4) 1911-1921	(5) 1911-1931
Sex	0.1313***	0.1467***	0.1466***	0.1196***	0.1083***
Family Size	0.0041***	0.0011***	$9.13 \times 10^{-5}$	-0.0001***	-0.0001***
Province = AB	0.0034	-0.0093	-0.0044	-0.0145***	-0.0003
Province = SK	-0.0114	-0.0077	-0.0072	-0.0099**	-0.0011
Age	Yes	Yes	Yes	Yes	Yes
Birthplace	Yes	Yes	Yes	Yes	Yes
Relation	Yes	Yes	Yes	Yes	Yes
Homestead	Yes	Yes	Yes		
Milk Cows	Yes	Yes	Yes		
Horses	Yes	Yes	Yes		
Wheat Prod.	Yes	Yes	Yes		
Religion				Yes	Yes
Observations	430,101	430,101	430,101	1,191,247	1,191,247
R <sup>2</sup>	0.05674	0.06955	0.08882	0.04816	0.05726

Clustered (Enum. Subdist.) standard-errors in parentheses

Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

# Match Rate by Birthplace

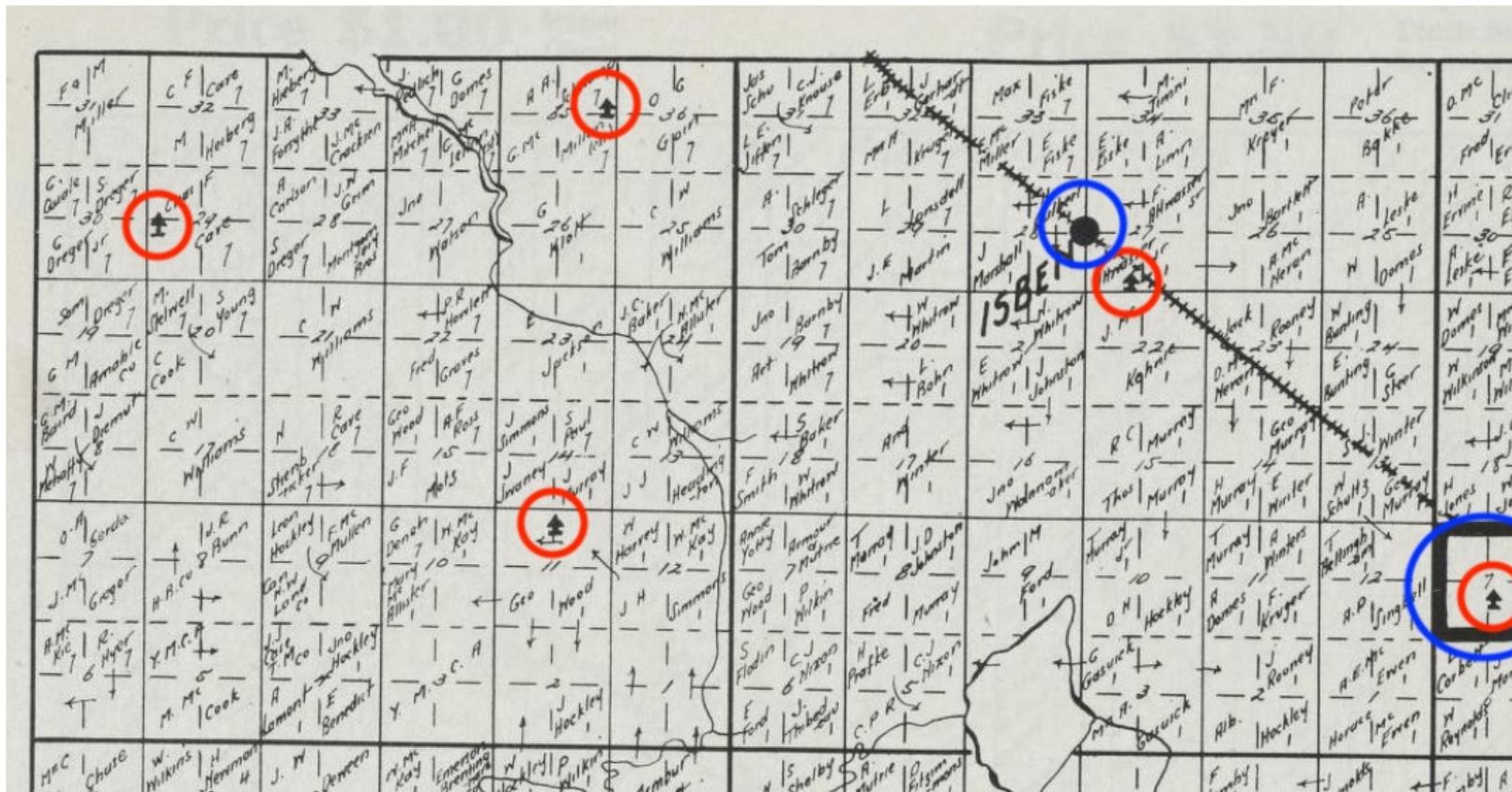
[Back to Linking Bias](#)

Model:	Matched				
	(1) 1906-1911	(2) 1906-1921	(3) 1906-1931	(4) 1911-1921	(5) 1911-1931
Sex	0.1307***	0.1465***	0.1467***	0.1197***	0.1084***
Family Size	0.0041***	0.0010***	0.0001	$-5.8 \times 10^{-5}$	$-7.95 \times 10^{-5} ***$
Birthplace = UnitedStates	-0.0347***	-0.0533***	-0.0462***	-0.0260***	-0.0283***
Birthplace = Canada+BC	-0.0491***	-0.0425***	-0.0391***	-0.0131***	-0.0101***
Birthplace = MB	-0.0378***	-0.0389***	-0.0417***	0.0113***	0.0062***
Birthplace = Maritimes	-0.0322***	-0.0213***	-0.0155**	-0.0111***	-0.0132***
Birthplace = QC	-0.0563***	-0.0268***	-0.0199***	0.0067	0.0032
Birthplace = Caribbean	-0.1479***	-0.0886***	-0.1132***	-0.0441**	-0.0571***
Birthplace = Nordic	-0.0760***	-0.0547***	-0.0334***	-0.0327***	-0.0133***
Birthplace = UnitedKingdom	0.0319***	0.0322***	0.0358***	0.0003	0.0090***
Birthplace = Ireland	-0.0157*	-0.0178**	-0.0030	-0.0009	0.0065*
Birthplace = RestofEurope	-0.1094***	-0.0683***	-0.0469***	-0.0306***	-0.0115***
Birthplace = ROW	-0.0722***	-0.1175***	-0.0977***	-0.0556***	-0.0498***
Age	Yes	Yes	Yes	Yes	Yes
Relation	Yes	Yes	Yes	Yes	Yes
Enum. Subdist.	Yes	Yes	Yes	Yes	Yes
Homestead	Yes	Yes	Yes		
Milk Cows	Yes	Yes	Yes		
Horses	Yes	Yes	Yes		
Wheat Prod.	Yes	Yes	Yes		
Religion				Yes	Yes

## Cummins Map Example

▶ Back to Data

→ Back to Schooling Choice



## Townships and Sections

[Back to Geolocations](#)

- We can geolocate households throughout the prairies because of how land was surveyed.
- Canadian prairies were surveyed in a grid, with 6 mile by 6 mile Townships, and 1 mile by 1 mile sections.
- In census records, administrative documents, and colloquial use, addresses were expressed using the survey boundaries of land.
  - These correspond to the modern day land divisions, allowing for precise geolocations.

# Townships

[Back to Geolocations](#)

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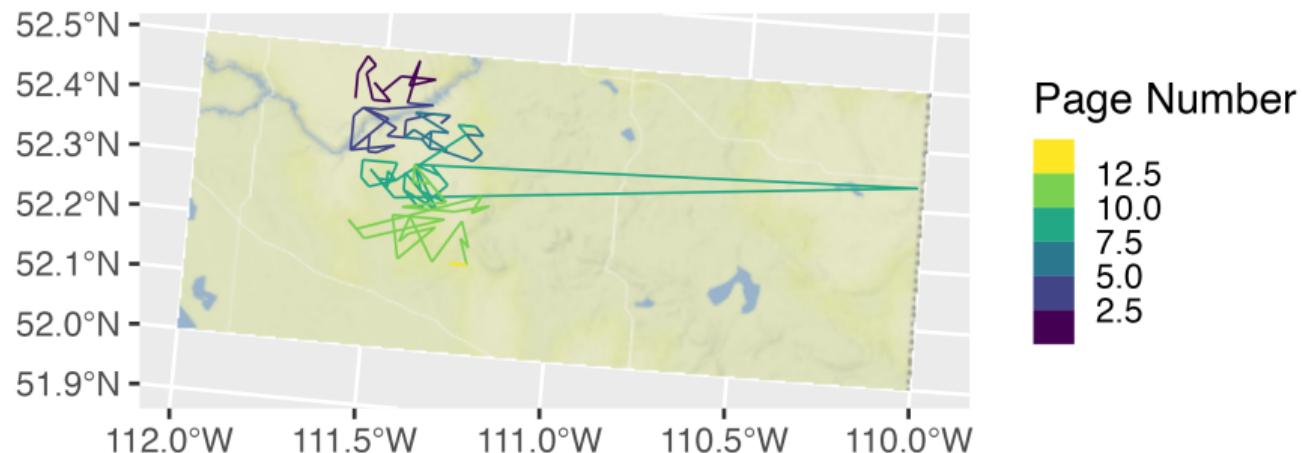
18 IX

## INDEX TO TOWNSHIPS IN MANITOBA, SASKATCHEWAN, ALBERTA, AND BRITISH COLUMBIA

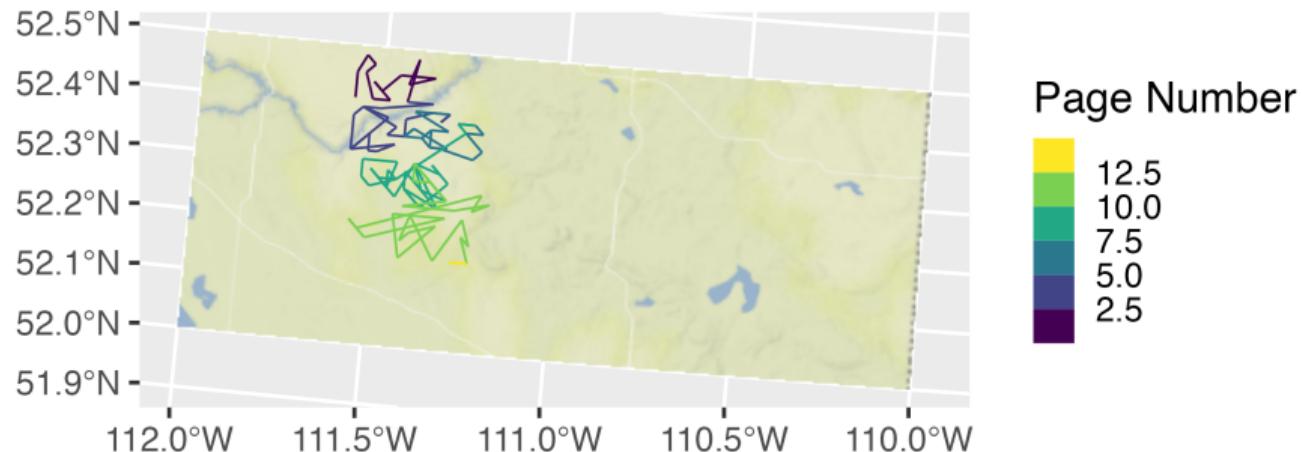
Showing the Townships for which Official and Preliminary Plans have been issued up to January 1st, 1929.



## Raw Path - District 21, Subdistrict 7, 1906

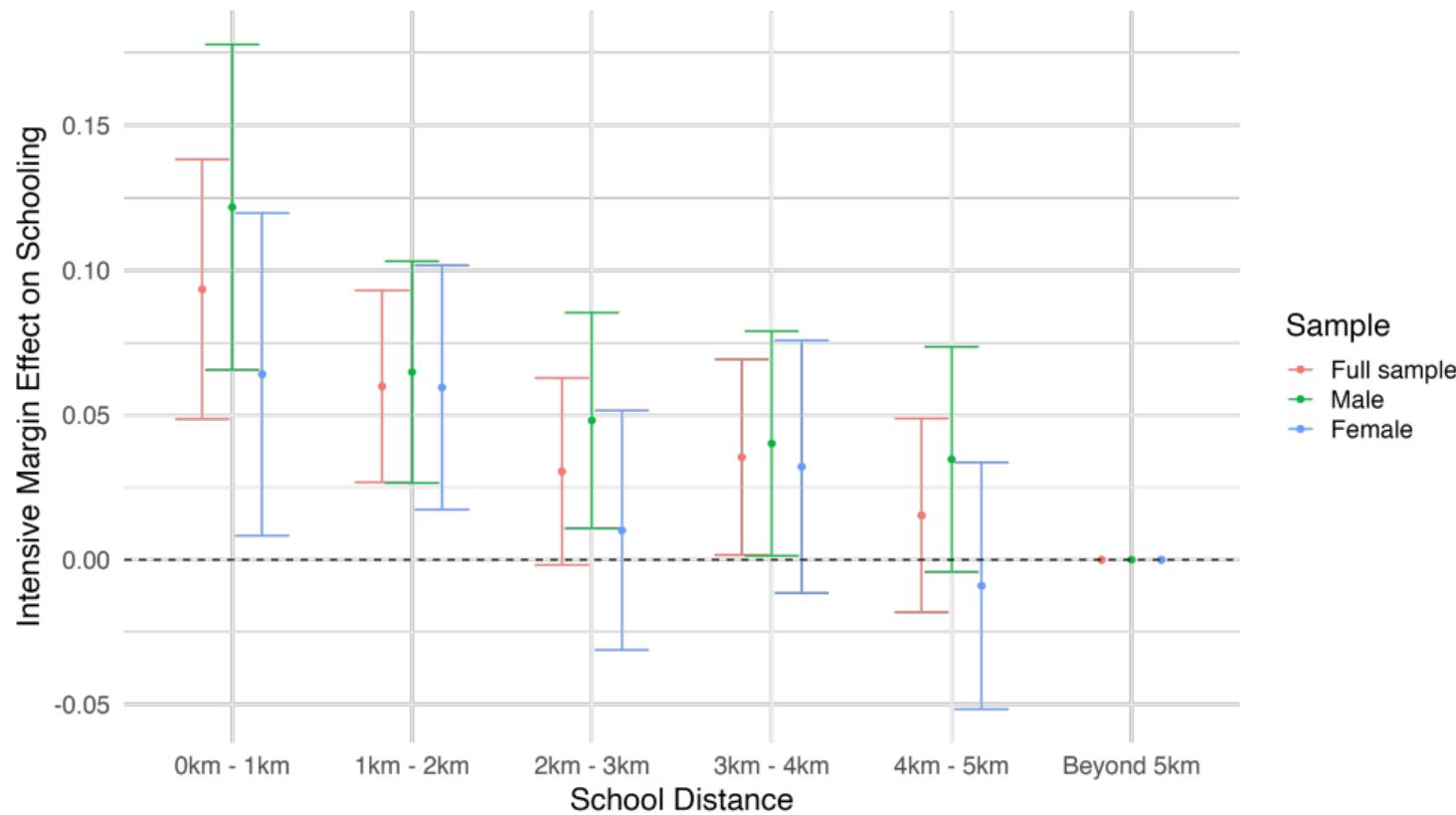


## Clean Path - District 21, Subdistrict 7, 1906



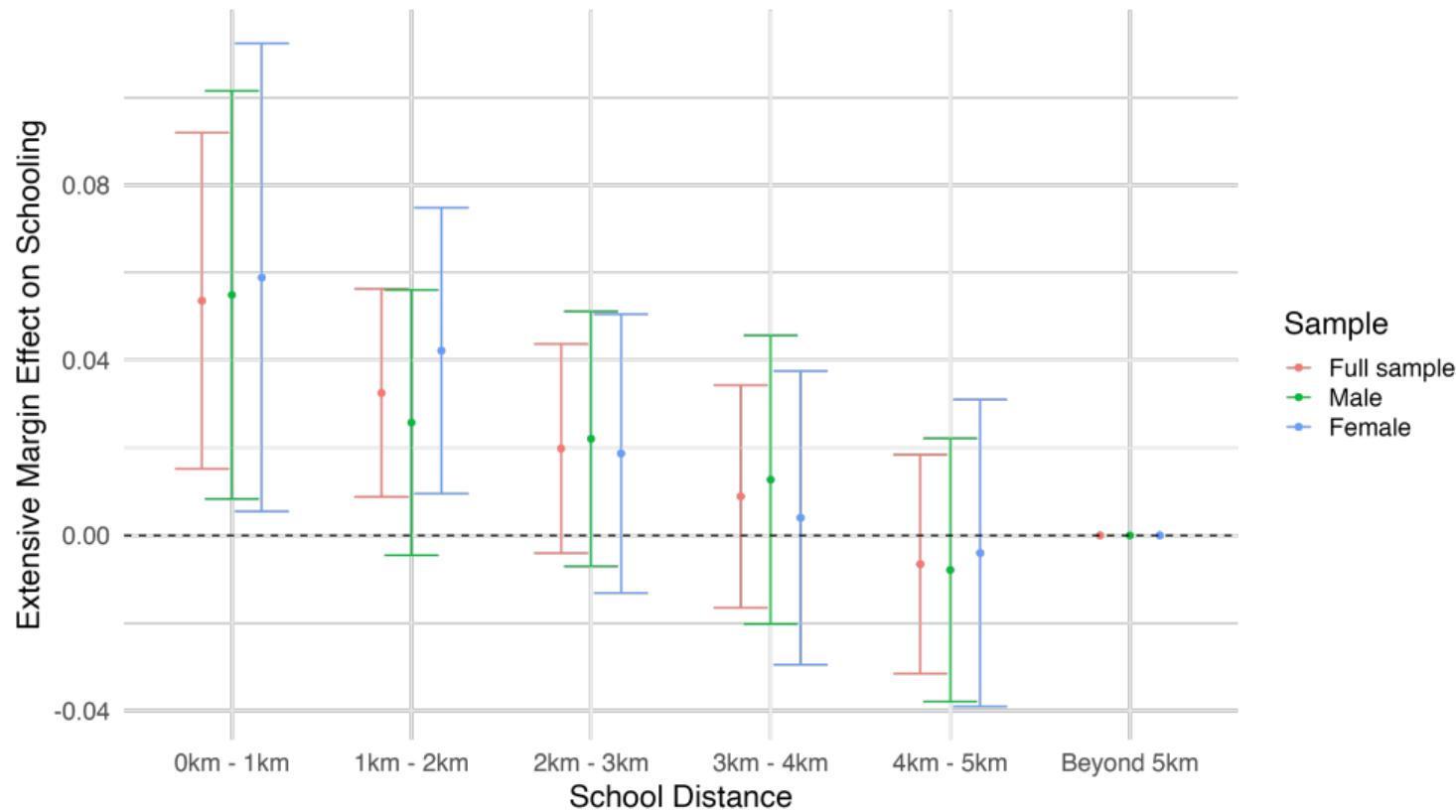
# Schooling vs Distance: Intensive Margin

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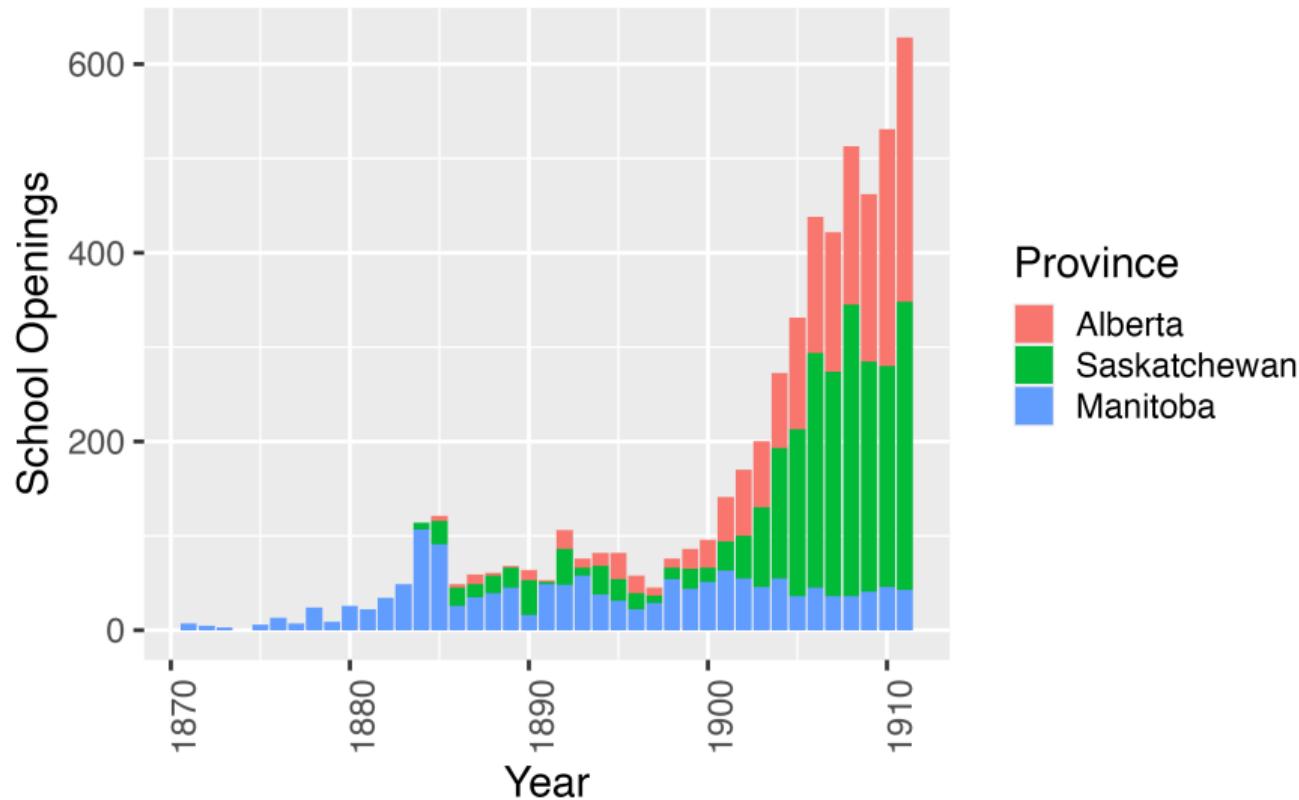
# Schooling vs Distance: Extensive Margin

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# Annual School Openings

[Back to DiD](#)

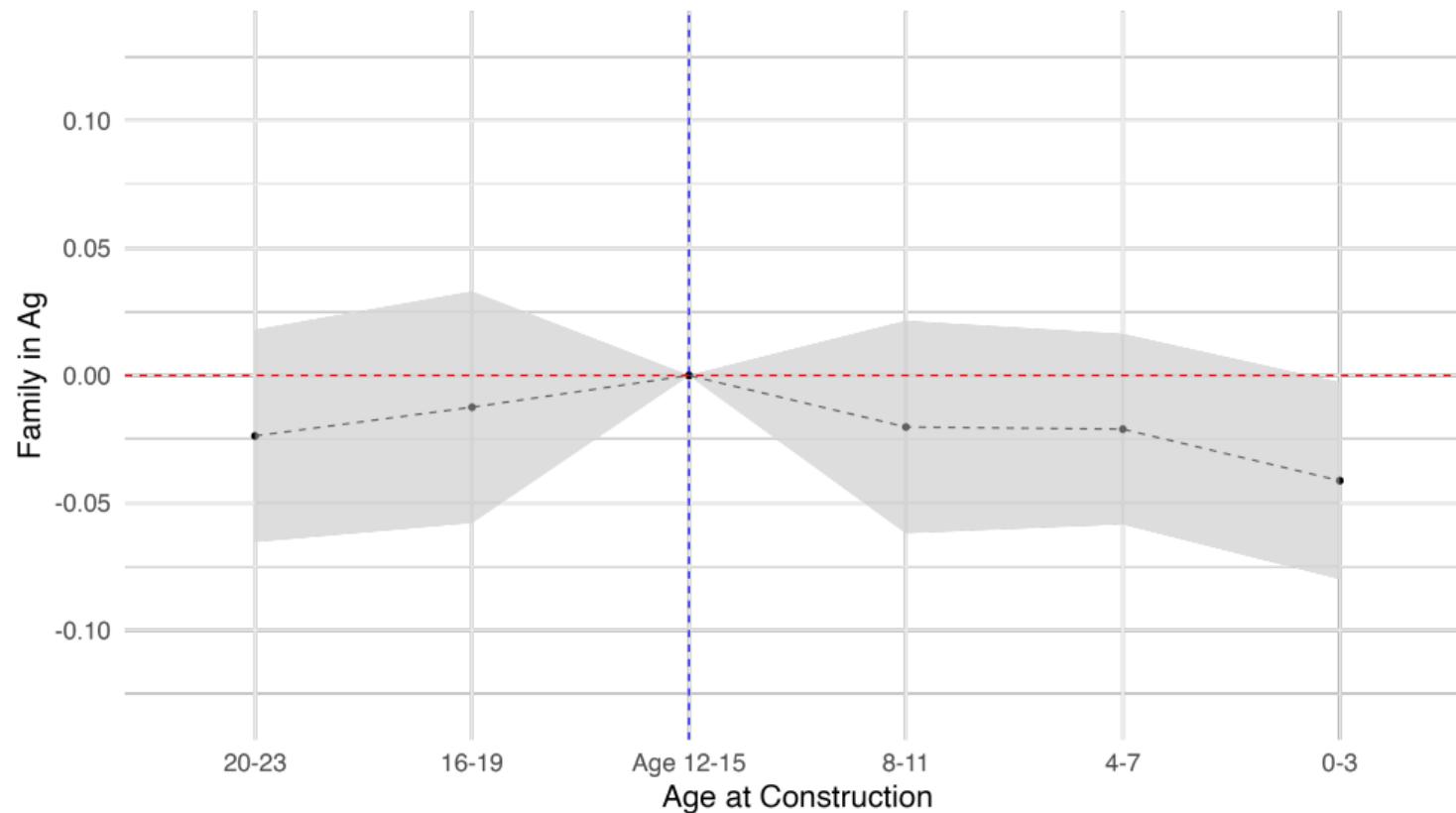


Dependent Variable:	Farm Work				
Model:	(1)	(2)	(3)	(4)	(5)
School Constr. × Cohort	-0.0405** (0.0167)	-0.0304* (0.0169)	-0.0318* (0.0169)	-0.0322* (0.0169)	-0.0326* (0.0169)
School Constr.	0.0108 (0.0172)	0.0047 (0.0174)	0.0055 (0.0174)	0.0040 (0.0175)	0.0042 (0.0173)
Cohort	-0.0191** (0.0092)	-0.0115 (0.0097)	0.0085 (0.0098)	0.0084 (0.0098)	0.0076 (0.0098)
Log RR Distance (1906)					0.0205*** (0.0046)
Link Prob.	0.4780*** (0.0247)	0.4656*** (0.0251)	0.4533*** (0.0248)	0.4539*** (0.0249)	0.4549*** (0.0249)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes	Yes
Fam. Bpl.		Yes	Yes	Yes	Yes
Homestead			Yes	Yes	Yes
Livestock			Yes	Yes	Yes
Ag. Productive.				Yes	Yes
Observations	14,944	14,944	14,944	14,944	14,944
R <sup>2</sup>	0.07941	0.08531	0.09614	0.09733	0.09893
Dependent variable mean	0.62018	0.62018	0.62018	0.62018	0.62018

Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses  
 Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

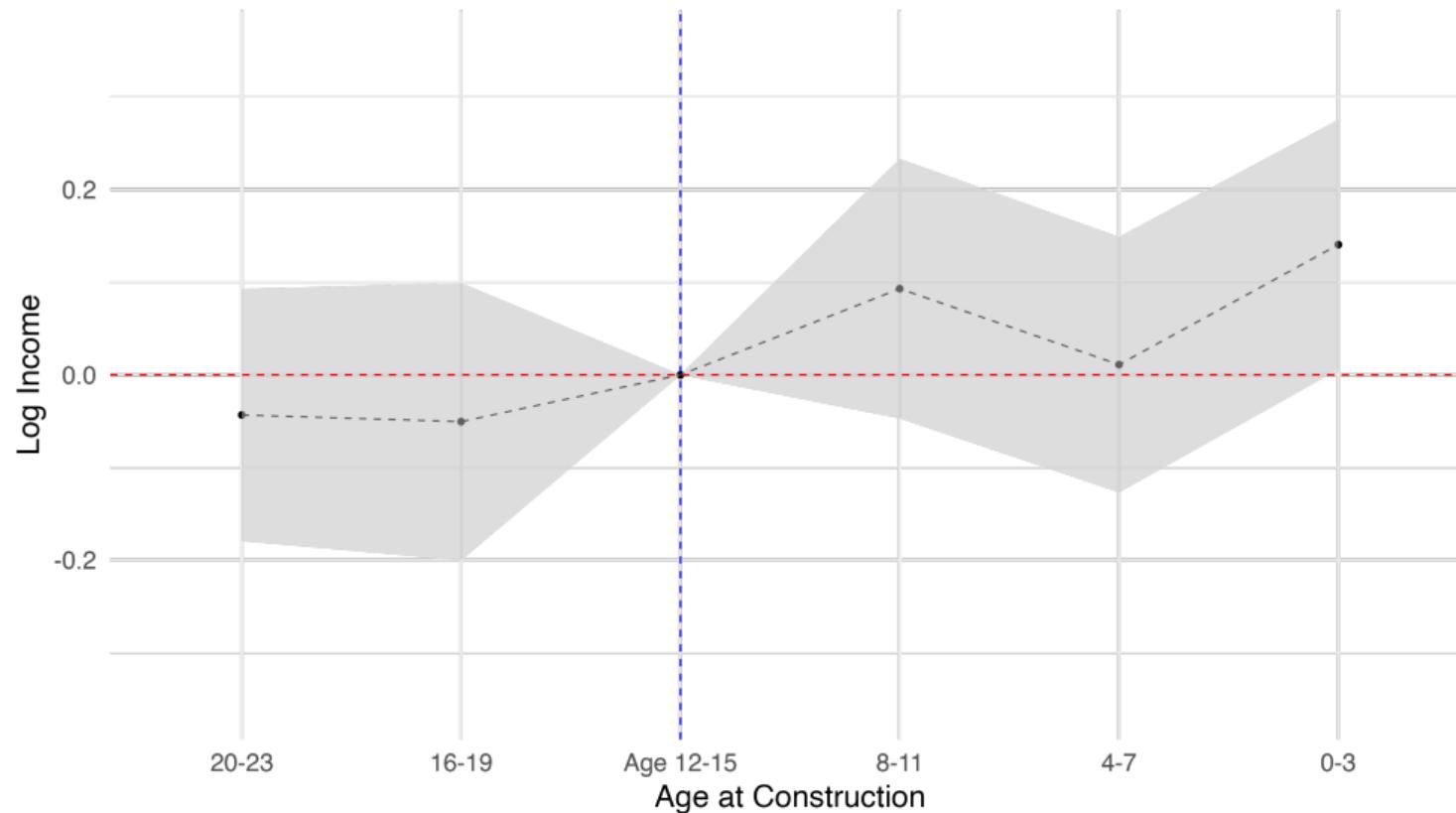
# Family in Agriculture Event Study

Back



# Income Event Study

Back



Dependent Variables: Model:	Migration Dist. (km)	Asinh Migration. Dist.	Log Migration Dist.	On Family Farm
	(1)	(2)	(3)	(4)
School Constr. × Cohort	28.62*** (10.18)	0.1542* (0.0850)	0.1268* (0.0729)	-0.0129 (0.0174)
School Constr.	-2.942 (8.529)	-0.0327 (0.0789)	-0.0290 (0.0712)	0.0004 (0.0157)
Cohort	-4.966 (6.259)	-0.1574*** (0.0500)	0.0155 (0.0447)	0.0385*** (0.0095)
Log RR Distance (1906)	-3.407 (2.448)	-0.0244 (0.0227)	-0.0044 (0.0197)	0.0108*** (0.0038)
Link Prob.	-971.6*** (27.55)	-5.512*** (0.1231)	-4.698*** (0.1039)	0.5013*** (0.0188)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes
Fam. Bpl. + Homestead	Yes	Yes	Yes	Yes
Livestock + Ag.	Yes	Yes	Yes	Yes
Observations	14,944	14,944	13,217	14,944
R <sup>2</sup>	0.28142	0.22182	0.24087	0.12132
Dependent variable mean	224.04	4.2159	4.0652	0.23367

*Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses*

*Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1*

Dependent Variable:	Employee				
Model:	(1)	(2)	(3)	(4)	(5)
School Constr. × Cohort	0.0176 (0.0179)	0.0103 (0.0179)	0.0115 (0.0178)	0.0120 (0.0178)	0.0124 (0.0177)
School Constr.	0.0050 (0.0155)	0.0113 (0.0154)	0.0096 (0.0153)	0.0130 (0.0155)	0.0129 (0.0154)
Cohort	0.0970*** (0.0101)	0.0878*** (0.0108)	0.0710*** (0.0108)	0.0711*** (0.0108)	0.0717*** (0.0108)
Log RR Distance (1906)					-0.0181*** (0.0045)
Link Prob.	-0.3419*** (0.0219)	-0.3348*** (0.0221)	-0.3247*** (0.0220)	-0.3260*** (0.0221)	-0.3269*** (0.0221)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes	Yes
Fam. Bpl.		Yes	Yes	Yes	Yes
Homestead			Yes	Yes	Yes
Livestock			Yes	Yes	Yes
Ag. Productive.				Yes	Yes
Observations	14,944	14,944	14,944	14,944	14,944
R <sup>2</sup>	0.07386	0.07982	0.08833	0.09050	0.09181
Dependent variable mean	0.34475	0.34475	0.34475	0.34475	0.34475

*Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses*  
*Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1*

Dependent Variable:	Employer				
Model:	(1)	(2)	(3)	(4)	(5)
School Constr. × Cohort	-0.0035 (0.0155)	-0.0026 (0.0156)	-0.0021 (0.0157)	-0.0028 (0.0156)	-0.0028 (0.0156)
School Constr.	0.0118 (0.0139)	0.0102 (0.0139)	0.0117 (0.0139)	0.0108 (0.0140)	0.0108 (0.0140)
Cohort	-0.0802*** (0.0066)	-0.0815*** (0.0072)	-0.0697*** (0.0073)	-0.0694*** (0.0073)	-0.0695*** (0.0073)
Log RR Distance (1906)					0.0016 (0.0034)
Link Prob.	0.1212*** (0.0164)	0.1207*** (0.0167)	0.1130*** (0.0165)	0.1143*** (0.0165)	0.1144*** (0.0165)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes	Yes
Fam. Bpl.		Yes	Yes	Yes	Yes
Homestead			Yes	Yes	Yes
Livestock			Yes	Yes	Yes
Ag. Productive.				Yes	Yes
Observations	14,944	14,944	14,944	14,944	14,944
R <sup>2</sup>	0.07100	0.07377	0.07988	0.08248	0.08249
Dependent variable mean	0.13597	0.13597	0.13597	0.13597	0.13597

*Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses*  
*Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1*

Dependent Variable:	Own Account				
Model:	(1)	(2)	(3)	(4)	(5)
School Constr. × Cohort	-0.0065 (0.0194)	0.0054 (0.0198)	0.0053 (0.0199)	0.0063 (0.0199)	0.0061 (0.0199)
School Constr.	-0.0166 (0.0166)	-0.0280* (0.0165)	-0.0283* (0.0166)	-0.0292* (0.0167)	-0.0291* (0.0166)
Cohort	-0.1493*** (0.0091)	-0.1304*** (0.0098)	-0.1250*** (0.0100)	-0.1255*** (0.0100)	-0.1260*** (0.0099)
Log RR Distance (1906)					0.0120*** (0.0045)
Link Prob.	0.1872*** (0.0222)	0.1922*** (0.0221)	0.1876*** (0.0221)	0.1879*** (0.0222)	0.1885*** (0.0222)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes	Yes
Fam. Bpl.		Yes	Yes	Yes	Yes
Homestead			Yes	Yes	Yes
Livestock			Yes	Yes	Yes
Ag. Productive.				Yes	Yes
Observations	14,292	14,292	14,292	14,292	14,292
R <sup>2</sup>	0.09792	0.11031	0.11388	0.11487	0.11541
Dependent variable mean	0.35502	0.35502	0.35502	0.35502	0.35502

*Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses*  
*Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1*

Dependent Variable:	Log Income				
Model:	(1) 3km	(2) 4km	(3) 5km	(4) 6km	(5) 7km
School Constr. × Cohort	0.1010* (0.0592)	0.0815 (0.0560)	0.1226** (0.0591)	0.1299** (0.0587)	0.1316** (0.0632)
School Constr.	0.0134 (0.0498)	0.0346 (0.0517)	0.0050 (0.0546)	-0.0259 (0.0560)	-0.0106 (0.0588)
Cohort	-0.2508*** (0.0305)	-0.2490*** (0.0312)	-0.2585*** (0.0308)	-0.2603*** (0.0307)	-0.2590*** (0.0308)
Log RR Distance (1906)	-0.0314** (0.0123)	-0.0315** (0.0123)	-0.0318** (0.0123)	-0.0320*** (0.0123)	-0.0323*** (0.0124)
Link Prob.	-0.2032*** (0.0621)	-0.2024*** (0.0621)	-0.2018*** (0.0621)	-0.2023*** (0.0621)	-0.2022*** (0.0621)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes	Yes
Fam. Bpl. + Homestead	Yes	Yes	Yes	Yes	Yes
Livestock + Ag.	Yes	Yes	Yes	Yes	Yes
Observations	5,750	5,750	5,750	5,750	5,750
R <sup>2</sup>	0.17262	0.17262	0.17288	0.17257	0.17269
Dependent variable mean	6.5231	6.5231	6.5231	6.5231	6.5231

Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses

Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1



Dependent Variable:	Log Income			
Model:	(1) Baseline	(2) Vs Close	(3) Vs Far	(4) Vs Urban
School Constr. × Cohort	0.1226** (0.0591)	0.1725** (0.0720)	0.0988 (0.0706)	0.1490** (0.0624)
School Constr.	0.0050 (0.0546)	-0.0245 (0.0682)	0.0388 (0.0553)	-0.4199*** (0.0582)
Cohort	-0.2585*** (0.0308)	-0.3148*** (0.0396)	-0.2112*** (0.0543)	-0.3872*** (0.0336)
Log RR Distance (1906)	-0.0318** (0.0123)	-0.0482*** (0.0167)	-0.0216 (0.0196)	
Link Prob.	-0.2018*** (0.0621)	-0.1819** (0.0759)	-0.1329 (0.0876)	0.2660*** (0.0674)
Enumeration Subdistrict (1906)	Yes	Yes	Yes	
Birthplace	Yes	Yes	Yes	Yes
Father Birthplace	Yes	Yes	Yes	Yes
Mother Birthplace	Yes	Yes	Yes	Yes
Homestead	Yes	Yes	Yes	
Enumeration District (1906)				Yes
Fam. Bpl. + Homestead	Yes	Yes	Yes	partial
Livestock + Ag.	Yes	Yes	Yes	
Observations	5,750	3,857	3,095	3,986
R <sup>2</sup>	0.17288	0.19322	0.23183	0.12405
Dependent variable mean	6.5231	6.5445	6.5006	6.8325

Clustered (Enumeration Subdistrict (1906)) standard errors in parentheses

Dependent Variable:	Log Income					
Model:	(1) 0-5 vs 6-11	(2) 0-5 vs 12-17	(3) 0-5 vs 18-23	(4) 6-11 vs 12-17	(5) 6-11 vs 18-23	(6) 12-17 vs 18-23
<i>Variables</i>						
School Constr. × Cohort	0.0468 (0.0528)	0.1226** (0.0591)	0.1831*** (0.0529)	0.1101* (0.0618)	0.0947* (0.0551)	0.0663 (0.0627)
School Constr.	-0.0258 (0.0488)	0.0050 (0.0546)	-0.1015** (0.0505)	-0.0711 (0.0577)	-0.1556*** (0.0537)	-0.0158 (0.0555)
Cohort	-0.1857*** (0.0273)	-0.2585*** (0.0308)	-0.2225*** (0.0338)	-0.0921*** (0.0292)	-0.0480 (0.0313)	0.0187 (0.0314)
Log RR Distance (1906)	-0.0251** (0.0120)	-0.0318** (0.0123)	-0.0177 (0.0125)	-0.0332*** (0.0116)	-0.0275** (0.0118)	-0.0372*** (0.0132)
Link Prob.	-0.1089* (0.0632)	-0.2018*** (0.0621)	-0.1134* (0.0627)	-0.2240*** (0.0667)	-0.1455** (0.0641)	-0.2348*** (0.0666)
<i>Fixed-effects</i>						
Enumeration Subdistrict (1906)	Yes	Yes	Yes	Yes	Yes	Yes
Fam. Bpl. + Homestead	Yes	Yes	Yes	Yes	Yes	Yes
Livestock + Ag.	Yes	Yes	Yes	Yes	Yes	Yes
<i>Fit statistics</i>						
Observations	6,223	5,750	6,091	5,253	5,594	5,121
R <sup>2</sup>	0.15385	0.17288	0.17415	0.16089	0.15167	0.15958
Dependent variable mean	6.4950	6.5231	6.5400	6.6499	6.6607	6.7075

Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses

Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

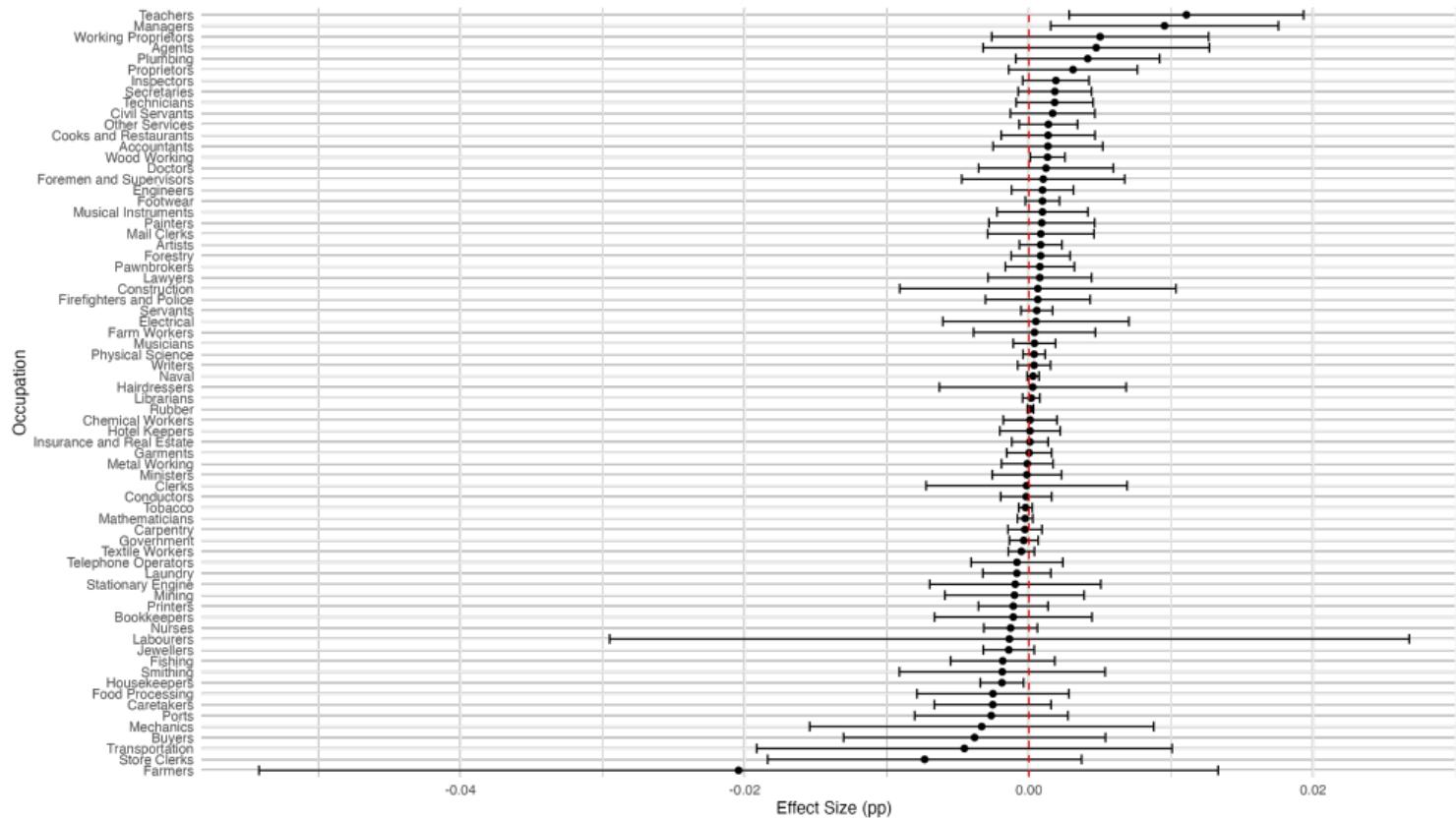
Dependent Variable: Model:	Family in Ag		
	(1)	(2)	(3)
	Full Sample	Within 5km of Railway	Beyond 5km of Railway
School Constr. × Cohort	-0.0342** (0.0153)	0.0108 (0.0367)	-0.0503*** (0.0191)
School Constr.	0.0047 (0.0165)	0.0165 (0.0406)	0.0047 (0.0185)
Cohort	0.0626*** (0.0092)	0.0537*** (0.0138)	0.0684*** (0.0127)
Link Prob.	0.4146*** (0.0245)	0.3764*** (0.0406)	0.4408*** (0.0299)
Enumeration Subdistrict (1906)	Yes	Yes	Yes
Fam. Bpl. + Homestead	Yes	Yes	Yes
Livestock + Ag.	Yes	Yes	Yes
Observations	14,944	5,371	9,573
R <sup>2</sup>	0.09796	0.16115	0.11051
Dependent variable mean	0.69279	0.65388	0.71461

Clustered (Enumeration Subdistrict (1906)) standard-errors in parentheses  
 Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1



# Switching to High Skill Services

Back



Dependent Variable:	Ln Farm Area				
Model:	(1) 1906	(2)	(3) 1921	(4)	(5) 1931
Asinh Horses	0.0559*** (0.0049)				
Asinh Milk Cows	-0.0004 (0.0051)				
Asinh Hogs	-0.0112*** (0.0035)				
Asinh Cattle	0.0234*** (0.0032)				
Asinh Sheep	0.0057 (0.0059)				
Family Size	-0.0024 (0.0017)	-0.0063*** (0.0009)	-0.0044* (0.0023)	0.0030*** (0.0008)	0.0012 (0.0019)
Fam. Adult Males	0.0271*** (0.0049)	0.0206*** (0.0028)	0.0152** (0.0070)	0.0202*** (0.0026)	0.0131*** (0.0047)
Homestead	-0.3169*** (0.0186)	-0.1730*** (0.0072)	-0.1624*** (0.0181)	-0.1699*** (0.0062)	-0.1637*** (0.0139)
Asinh Wheat	-0.0138 (0.1390)	0.1437** (0.0617)	-0.0287 (0.1968)	0.0074 (0.0862)	0.0635 (0.1766)
Asinh Oats	0.0756 (0.2470)	-0.7318*** (0.1506)	-0.6965** (0.3189)	-0.2634 (0.1626)	-0.3276 (0.2926)
Asinh Grass	-0.0049 (0.0152)	-0.2574*** (0.0599)	-0.1207* (0.0681)	0.0626 (0.0581)	0.0454 (0.0558)
Asinh Flax	0.0050 (0.0450)	0.0970 (0.0669)	0.0754 (0.0757)	-0.1193 (0.0761)	-0.1052 (0.1003)
Asinh Barley	-0.1012 (0.2248)	0.7859*** (0.1405)	0.8071*** (0.3037)	0.2907** (0.1320)	0.3371 (0.2523)
Employer		0.0405*** (0.0137)	-0.0096 (0.0293)	0.0290*** (0.0049)	0.0207 (0.0127)
Employee		-0.2295*** (0.0149)	-0.2749*** (0.0336)	-0.2082*** (0.0111)	-0.2459*** (0.0160)
Log Income			-0.0181*** (0.0068)		-0.0233*** (0.0084)

Model:	Farm Area			
	(1) Movers 1906-1921	(2) On FF	(3) Movers 1906-1931	(4) On FF
Farm Area (1906)	0.0395*** (0.0111)	0.0883*** (0.0112)	0.0295* (0.0152)	0.0863*** (0.0160)
Homestead (1906)	0.0440* (0.0233)	0.0150 (0.0263)	-0.0392 (0.0322)	-0.0042 (0.0383)
Homestead (1921)	-0.2108*** (0.0224)	-0.1662*** (0.0243)		
Homestead (1931)			-0.1303*** (0.0274)	-0.1251*** (0.0370)
Enum. Sub. (1921)	Yes	Yes		
Enum. Sub. (1921)			Yes	Yes
Observations	6,079	7,203	4,517	4,541
R <sup>2</sup>	0.20421	0.21970	0.46946	0.43374

Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1



Dependent Variable:	Ln Farm Area				
Model:	(1) 3km	(2) 4km	(3) 5km	(4) 6km	(5) 7km
Early $\times$ Close $\times$ Year = 1931	0.0226 (0.0259)	0.0484* (0.0260)	0.0529* (0.0304)	0.0739* (0.0407)	0.1096** (0.0540)
Early $\times$ Close $\times$ Year = 1921	-0.0149 (0.0259)	0.0106 (0.0257)	0.0078 (0.0288)	0.0225 (0.0386)	0.0625 (0.0571)
Early $\times$ Close	-0.0216 (0.0223)	-0.0340 (0.0221)	-0.0133 (0.0243)	-0.0064 (0.0322)	-0.0399 (0.0457)
Homestead	-0.2125*** (0.0050)	-0.2135*** (0.0050)	-0.2131*** (0.0050)	-0.2131*** (0.0050)	-0.2130*** (0.0050)
RR Dist in Year	0.0403*** (0.0035)	0.0407*** (0.0035)	0.0408*** (0.0035)	0.0408*** (0.0035)	0.0409*** (0.0035)
Family Size	-0.0005 (0.0005)	-0.0006 (0.0005)	-0.0006 (0.0005)	-0.0006 (0.0005)	-0.0006 (0.0005)
Fam. Adult Males	0.0044*** (0.0016)	0.0045*** (0.0016)	0.0045*** (0.0016)	0.0045*** (0.0016)	0.0046*** (0.0016)
Other Terms	Yes	Yes	Yes	Yes	Yes
School District	Yes	Yes	Yes	Yes	Yes
Observations	439,027	439,027	439,027	439,027	439,027
R <sup>2</sup>	0.15431	0.15384	0.15382	0.15383	0.15378

Clustered (Enum. Sub.) standard-errors in parentheses

Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

Dependent Variable:	Ln Farm Area				
Model:	(1) 1904	(2) 1905	(3) 1906	(4) 1907	(5) 1908
Early $\times$ Close $\times$ Year = 1931	0.0720* (0.0408)	0.0675* (0.0347)	0.0529* (0.0304)	0.0610** (0.0311)	0.0343 (0.0317)
Early $\times$ Close $\times$ Year = 1921	-0.0117 (0.0348)	0.0009 (0.0318)	0.0078 (0.0288)	0.0228 (0.0289)	-0.0025 (0.0303)
Early $\times$ Close	0.0060 (0.0286)	-0.0089 (0.0258)	-0.0133 (0.0243)	-0.0234 (0.0248)	0.0128 (0.0274)
Homestead	-0.2125*** (0.0050)	-0.2129*** (0.0050)	-0.2131*** (0.0050)	-0.2132*** (0.0050)	-0.2132*** (0.0050)
RR Dist in Year	0.0420*** (0.0035)	0.0414*** (0.0035)	0.0408*** (0.0035)	0.0406*** (0.0035)	0.0407*** (0.0035)
Family Size	-0.0007 (0.0005)	-0.0006 (0.0005)	-0.0006 (0.0005)	-0.0005 (0.0005)	-0.0005 (0.0005)
Fam. Adult Males	0.0048*** (0.0016)	0.0046*** (0.0016)	0.0045*** (0.0016)	0.0044*** (0.0016)	0.0044*** (0.0016)
Other Terms	Yes	Yes	Yes	Yes	Yes
School District	Yes	Yes	Yes	Yes	Yes
Observations	439,027	439,027	439,027	439,027	439,027
R <sup>2</sup>	0.15403	0.15387	0.15382	0.15385	0.15390

Clustered (Enum. Sub.) standard-errors in parentheses  
Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1

Dependent Variable:	Ln Farm Area	
Model:	(1) Baseline	(2) Drop 1906
Early $\times$ Close $\times$ Year = 1931	0.0529* (0.0304)	0.0459** (0.0216)
Early $\times$ Close	-0.0133 (0.0243)	-0.0045 (0.0149)
Early $\times$ Close $\times$ Year = 1921	0.0078 (0.0288)	
Homestead	-0.2131*** (0.0050)	-0.1850*** (0.0047)
RR Dist in Year	0.0408*** (0.0035)	0.0519*** (0.0032)
Family Size	-0.0006 (0.0005)	-0.0014** (0.0006)
Fam. Adult Males	0.0045*** (0.0016)	0.0068*** (0.0017)
Other Terms	Yes	Yes
School District	Yes	Yes
Observations	439,027	358,129
R <sup>2</sup>	0.15382	0.17134

*Clustered (Enum. Sub.) standard-errors in parentheses  
Signif. Codes: \*\*\*: 0.01, \*\*: 0.05, \*: 0.1*

## Why do Schools Open Sooner in Some Places?

- Rural common schools were constructed by local communities, who organized together.
- Goldin and Katz argue that smaller communities were more homogenous, had higher levels of social capital, and therefore constructed schools sooner (Goldin & Katz, 2008).
- Dippel and Ottinger makes a similar argument, that small decentralized school boards facilitated increased population heterogeneity (Dippel et al., 2020).
- Can test this by looking at the local determinants of school opening.
  - Look at the set of schools that opened between 1900-1912
  - Link schools to the characteristics of the people that lived nearby (in 1906 when observe locations).
- Standardize variables to compare, each variable is expressed in standard deviations.

	School Year Established		
	(1)	(2)	(3)
RR Distance 1899 (Log)	0.5537*** (0.1440)	0.5375*** (0.1436)	0.5328*** (0.1441)
School Distance 1899 (Log)	0.2362* (0.1363)	0.2338* (0.1363)	0.2316* (0.1360)
Local Children 1906 (Count)	-0.4225*** (0.1197)	-0.4704*** (0.1201)	-0.4915*** (0.1211)
Local Households 1906 (Count)	-0.1476 (0.0998)	-0.1502 (0.0987)	-0.1467 (0.0991)
Cattle 1906 (Count)	0.0799 (0.0618)	0.0762 (0.0600)	0.0783 (0.0601)
Hogs 1906 (Count)	-0.2889*** (0.1092)	-0.2603** (0.1086)	-0.2613** (0.1089)
Milk Cows 1906 (Count)	-0.2925** (0.1175)	-0.2647** (0.1169)	-0.2560** (0.1177)
Area Entirely Homestead (Indicator)	0.1021* (0.0569)	0.0809 (0.0570)	0.0821 (0.0569)
Birthplace HHI (Percentage)	0.1668*** (0.0596)	0.1691*** (0.0590)	0.1568*** (0.0593)
Religion HHI (Percentage)	0.1901*** (0.0653)	0.1546** (0.0651)	0.1522** (0.0651)
Anglo-North American (Percentage)		-0.3393*** (0.0666)	-0.3130*** (0.0699)
Methodist (Percentage)			-0.1027 (0.0624)
Presbyterian (Percentage)			-0.1120* (0.0607)
Lutheran (Percentage)			-0.1215* (0.0676)
Observations	2,926	2,926	2,926
R <sup>2</sup>	0.44815	0.45350	0.45483
Geographic Controls fixed effects	✓	✓	✓

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