A data-driven, model-based approach to infer shifts in social contact patterns during an evolving pandemic

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Infectious Disease Outbreaks Webinar

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MATHEMATICS

OF COVID-19

TASK FORCE



COVID-19 in Canada

Non-pharmaceutical interventions:

- School closures;
- Workplace closures;
- Establishment closures;
- Limiting the size of gatherings.
- Hygienic measures;
- Respiratory etiquette;
- Face masks.



1. Coronavirus disease (COVID-19): Outbreak update - Canada.ca. https://www.canada.ca/en/public-health/services/diseases/2019-novel-coronavirus-infection.html (accessed Jan 26, 2020).

Outline

> Establish a methodology for modelling impacts of non-pharmaceutical interventions (NPIs) on social contact mixing.

Framework enables us to infer age-specific:

- Susceptibility to infection by SARS-CoV-2;
- Contact mixing in workplace, household, school and leisure settings;
- Transmission acquired in these settings.



2. Ontario COVID-19 Data Tool | Public Health Ontario. https://www.publichealthontario.ca/en/data-and-analysis/infectious-disease/covid-19-data-surveillance/covid-19-data-tool (accessed Nov 10, 2020).

Assessing the impacts of NPIs on contact mixing

In Ontario, we consider four distinct phases:

- February 26 March 14 (phase 0) [Monitoring and travel advisories]
- \succ March 14 17 (phase 1) [Public school closure]

- March 18 23 (phase 2) [State of emergency and physical distancing]
- March 24 May 16 (phase 3) [Non-essential workplace closure]

Modelling strategy

- **1**. Introduce transmission model;
- Identify a social contact matrix representative of 2. mixing in Ontario, Canada;
- Capture the distinct phases of public health 3. interventions taken by the government.

Transmission model

Mechanistic model (ODEs) with discrete age structure.

Age-stratified parameters

- Susceptibility to infection by SARS-CoV-2;
- Diagnosis rates among symptomatic individuals;
- Social contact mixing matrix.

$$\begin{split} S_{i} &:= -\sum_{j=1}^{n} (\beta_{i}C_{ij} + q(1 - \beta_{i})C_{ij})S_{i}(I_{j} + \theta A_{j})/N_{j} + \lambda S_{qi}, \\ E_{i} &:= \sum_{j=1}^{n} \beta_{i}C_{ij}(1 - q)S_{i}(I_{j} + \theta A_{j})/N_{j} - \sigma E_{i}, \\ I_{i} &:= \sigma \varrho E_{i} - (\delta_{Ii} + \gamma_{I})I_{i}, \\ A_{i} &:= \sigma (1 - \varrho)E_{i} - \gamma_{A}A_{i}, \\ S_{qi} &:= \sum_{j=1}^{n} (1 - \beta_{i})C_{ij}qS_{i}(I_{j} + \theta A_{j})/N_{j} - \lambda S_{qi}, \\ E_{qi} &:= \sum_{j=1}^{n} \beta_{i}C_{ij}qS_{i}(I_{j} + \theta A_{j})/N_{j} - \delta_{q}E_{qi}, \\ D_{i} &:= \delta_{Ii}I_{i} + \delta_{q}E_{qi} - (\alpha + \gamma_{D})D_{i}, \\ R_{i} &:= \gamma_{I}I_{i} + \gamma_{A}A_{i} + \gamma_{D}D_{i}. \end{split}$$





3. Tang, B., Scarabel, F., Bragazzi, N. L., McCarthy, Z., Glazer, M., Xiao, Y., ... & Wu, J. (2020). De-Escalation by Reversing the Escalation with a Stronger Synergistic Package of Contact Tracing, Quarantine, Isolation and Personal Protection: Feasibility of Preventing a COVID-19 Rebound in Ontario, Canada, as a Case Study. *Biology*, 9(5), 100.

for each age group i = 1, ..., n,

Established setting-specific contact matrices

2. Identifying social contact matrices representative of mixing in Ontario, Canada.





5. Prem, K., Cook, A. R., & Jit, M. (2017). Projecting social contact matrices in 152 countries using contact surveys and demographic data. *PLoS Computational Biology*, *13*(9), e1005697.

Baseline contact mixing in Ontario

2. Social contact matrices representative of mixing in Ontario, Canada.

 $C_{i,j} \rightarrow C'_{i',j'}$ Original (Canada) \rightarrow Target (Ontario)

A suitable contact matrix:

- i. Modified age group subdivisions (6 age groups);
- ii. Reciprocity condition is satisfied;
- iii. Accounts for the specific age structure of Ontario, Canada, in 2019;
- iv. Mean connectivity is representative of individual mean contact rate in Ontario, Canada.

Representative of contact mixing in Ontario, Canada in the absence of physical distancing measures.

6. Arregui, S., Aleta, A., Sanz, J., & Moreno, Y. (2018). Projecting social contact matrices to different demographic structures. *PLoS Computational Biology*, *14*(12), e1006638.



Heatmaps of estimated social contact matrices in Ontario, Canada. A) Households, B) Workplaces, C) Schools, D) Communities and other locations and E) contact mixing in all four settings combined.

Shifting mixing patterns in response to public health interventions

3. Capture the distinct phases of public health interventions taken by the government.

Ontario's response: A series of public health interventions taken such as school closure, physical distancing advisories, non-essential workplace closure.

We consider four distinct phases:

- February 26 March 14 (phase 0) [Monitoring and travel advisories]
- March 14 17 (phase 1) [Public school closure]
- March 18 23 (phase 2)

[State of emergency and physical distancing]

March 24 – May 16 (phase 3) [Non-essential workplace closure]

$$C(t) = \begin{cases} C^{0}, \ T_{S} < t < T_{0}, & \text{(phase 0)}, \\ C^{1}, \ T_{0} < t < T_{1}, & \text{(phase 1)}, \\ C^{2}, \ T_{1} < t < T_{2}, & \text{(phase 2)}, \\ C^{3}(t), \ T_{2} < t < T, & \text{(phase 3)}, \end{cases}$$

Shifting mixing patterns in response to public health interventions

Modelling the escalation in terms of setting-specific contact mixing.

Phase 0 (Monitoring and international travel advisories)

 $C^0 = C^H + C^W + C^C + C^S.$

Phase 1: Public school closure

$$C^{1} = (1 + p_{1}^{H})C^{H} + C^{W} + (1 + p_{1}^{C})C^{C} + 0 C^{S}.$$

Phase 2: Physical distancing advisories

 $C^{2} = (1 + p_{2}^{H})(1 + p_{1}^{H})C^{H} + C^{W} + (1 - p_{2}^{C})(1 + p_{1}^{C})C^{C}.$

Phase 3: Closure of non-essential workplaces

$$C^{3}(t) = \left[(1+p_{3}^{H}) - e^{-r_{H}(t-T_{2})}p_{3}^{H} \right] (1+p_{2}^{H})(1+p_{1}^{H})C^{H}$$

+ $[p_3^W e^{-r_W(t-T_2)} + (1 - p_3^W)]C^W + [p_3^C e^{-r_C(t-T_2)} + (1 - p_3^C)](1 - p_2^C)(1 + p_1^C)C^C$

Model fitting

Fit to age-stratified incidence:

- Six lists of time series;
- Feb 26 2020 May 16 2020;
- Single fitting procedure to estimate all age-dependent and ageindependent model parameters.

Parameter uncertainty quantification





Cumulative incidence according to age class.

- A) ages 0-5; B) ages 6-13; C) ages 14-17; D) ages 18, 24; E) ages 25, 64 and E) ages 65
- **D**) ages 18-24; **E**) ages 25-64 and **F**) ages 65+.

Effective reproduction number in Ontario

 R_t declines below 1 between April 5 and April 12 following the implementation of a package of non-pharmaceutical interventions.



Estimated effective reproduction number R_t .

Contact mixing pattern estimated for each escalation phase

Household contact pattern emerges following the implementation of public health measures.

Consistent with results from contact survey-based studies.



Age



25.64

Age

65×





Mean contact rate by escalation phase



Setting	Daily contact rate pre-	t rate pre-Running daily contact ratecontacts/day)(Change relative to pre-intervention)					
	interventions (contacts/day)						
		Phase 1	Phase 2	Phase 3			
				(May 16)			
School	1.52	0 (-100%)	0 (-100%)	0 (-100%)			
Workplace	3.66	3.66 (0%)	3.66 (0%)	1.49 (-59%)			
Community	4.14	4.42 (+7%)	2.97 (-28%)	0.64 (-85%)			
Household	2.95	3.34 (+13%)	4.29 (+45%)	4.45 (+51%)			
Total	12.27	11.42 (-7%)	10.92 (-11%)	6.58 (-46%)			

Table 1 | Estimated mean daily contact rate by setting escalation phase

Shift to household contact observed, by magnitude of contacts.

3 ×10⁴ w н 200 С 2.5 S 100 Cumulative cases 2 2126 1.5 Date 1 0.5 0 2125 ana 3/17 5/16 3/24 Date

Estimated infections acquired in Ontario according to setting

Community contacts initially contributed to more infections than contacts from remaining three locations, while household contacts played the primary role in contributing new infections after the closure of non-essential workplaces on March 24.

Estimated infections by setting acquired and age class

Cumulative infections by setting (and age class).Model-estimated cumulative infections acquired inA) Workplaces, B) Household, C) Community and other locations and D) Schools.





Cumulative infections by age class (and setting).
A) ages 0-5; B) ages 6-13; C) ages 14-17;
D) ages 18-24; E) ages 25-64 and F) ages 65+.

Age-dependent model parameters

Parameter		Age class				
	0-5	6-13	14-17	18-24	25-64	65+
Percentage of symptomatic individuals diagnosed	31%	38%	37%	20%	64%	67%
Susceptibility to infection	1.9%	1.6%	2.5%	9.2%	13.9%	50.2%

Percentage of symptomatic diagnosed individuals and susceptibility (i.e., probability of infection upon contact, β_i). The reported values are obtained from the mean of the fitting results of 1000 bootstrap realizations. The fraction of symptomatic diagnosed individuals is calculated from $\frac{\delta_{I,l}}{\delta_{I,l}+\gamma_I}$, where $\delta_{I,i}$ and γ_I denote the diagnosis and recovery rate, respectively.

We estimated that the susceptibility to SARS-CoV-2 infection increases with age.

Estimated an overall increasing trend with age in the percentage of symptomatic individuals diagnosed.



Empirical distributions of the age-specific susceptibility to infection.

Caveats to consider

Reported case data subject to error.

Testing protocols have been variable in Ontario.

Heterogeneities in geographical location.

Health care workers and long-term care homes were disproportionately affected.

Conclusions

Retrospective analysis:

Framework for intervention evaluation.

Proactive analysis:

Framework for evaluating reopening measures and vaccine distribution strategies.

Data-driven approach can provide insights and infer contact mixing in a timely manner, without the usage of participant-based surveys.

Methodology can be applied in many regions worldwide.

Projects in progress

1. Comparative analysis of contact shifts between Toronto and the rest of Ontario: insights for more region specific intervention strategies;

2. Age-specific immunization program optimization integrating with economic recovery;

3. Assess the impacts of relaxation and reopening measures in terms of contact mixing.

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- 3. Tang, B., Scarabel, F., Bragazzi, N. L., McCarthy, Z., Glazer, M., Xiao, Y., ... & Wu, J. (2020). De-Escalation by Reversing the Escalation with a Stronger Synergistic Package of Contact Tracing, Quarantine, Isolation and Personal Protection: Feasibility of Preventing a COVID-19 Rebound in Ontario, Canada, as a Case Study. *Biology*, 9(5), 100
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