



# Evaluation of the US COVID-19 vaccine allocation strategy

Audrey L. McCombs

M. R. Islam, T. Oraby, M. M. Chowdhury, M. al-Mamun, M. G. Tyshenko, C. Kadelka

Joint Statistical Meetings, American Statistical Association  
8 August 2022  
Washington, D.C.



Sandia National Laboratories is a multimission laboratory managed and operated by National Technology & Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.

# Introduction: Background to CDC Allocation



- Advisory Committee on Immunization Practices (ACIP)
  - Ensure equity in allocation and distribution
  - Reduce the burden on vulnerable groups
  - Maintain infrastructure and social order
  - Reduce mortality
  - Ensure transparency
- 4 Phases:
  - 1a. Healthcare personnel, and long-term care facility residents
  - 1b. Frontline essential workers, and 75+ year-olds
  - 1c. Other essential workers, 16-64 year-olds with comorbidities, and 65-74 year-olds
  - 2. 16-64 year-olds without comorbidities

# Introduction: Model Overview



- Compartmental Disease Model (20 compartments)
- US population stratified by characteristics in CDC allocation (17 groups)
- Four phases
- Important characteristics of COVID-19 pandemic:
  - Age-dependent susceptibility to infection
  - Age- and comorbidity-dependent CFR
  - Social contact rates
  - Case-dependent social-distancing levels
  - Speed of the vaccine roll-out
  - Vaccine hesitancy
  - Time-varying transmissibility of virus

# Introduction: Model Overview



- ~17.5 million potentially optimal strategies
  - $4^{17} \approx 1.7 \times 10^{10}$  total possible, then reduced
- Four primary metrics:
  - Total deaths
  - Total infections
  - Total cases
  - Years of life lost (YLL)
- Secondary metric: equitability
- Model parameters
  - Most derived from literature
  - Four parameters estimated using an elitist genetic algorithm, fit to data

# Methods: 17 Population Groups



Stratify the US population by:

1. Age (4 classes)
  1. 0-15 years
  2. 16-64 years
  3. 65-74 years
  4. 75+ years
2. Comorbidity (2 classes: with and without)
3. Job type (4 classes in 16-64 age group)
  1. Healthcare workers
  2. Frontline essential workers
  3. Other essential workers
  4. All others
4. Living situation (2 classes in 65-74 and 75+ age groups: congested living or not)

# Methods: 20 Compartments



**S:** Susceptible

**E:** Exposed, recently infected but not yet spreading the virus

**P:** Pre-clinical, not yet showing symptoms but spreading the virus

**A:** Asymptomatic, not symptomatic but spreading the virus

**C:** Clinical, symptomatic and spreading the virus

**Q:** Quarantine, symptomatic but not spreading the virus due to isolation or hospitalization

**RC:** Recovered after having shown symptoms

**RA:** Recovered after an asymptomatic infection

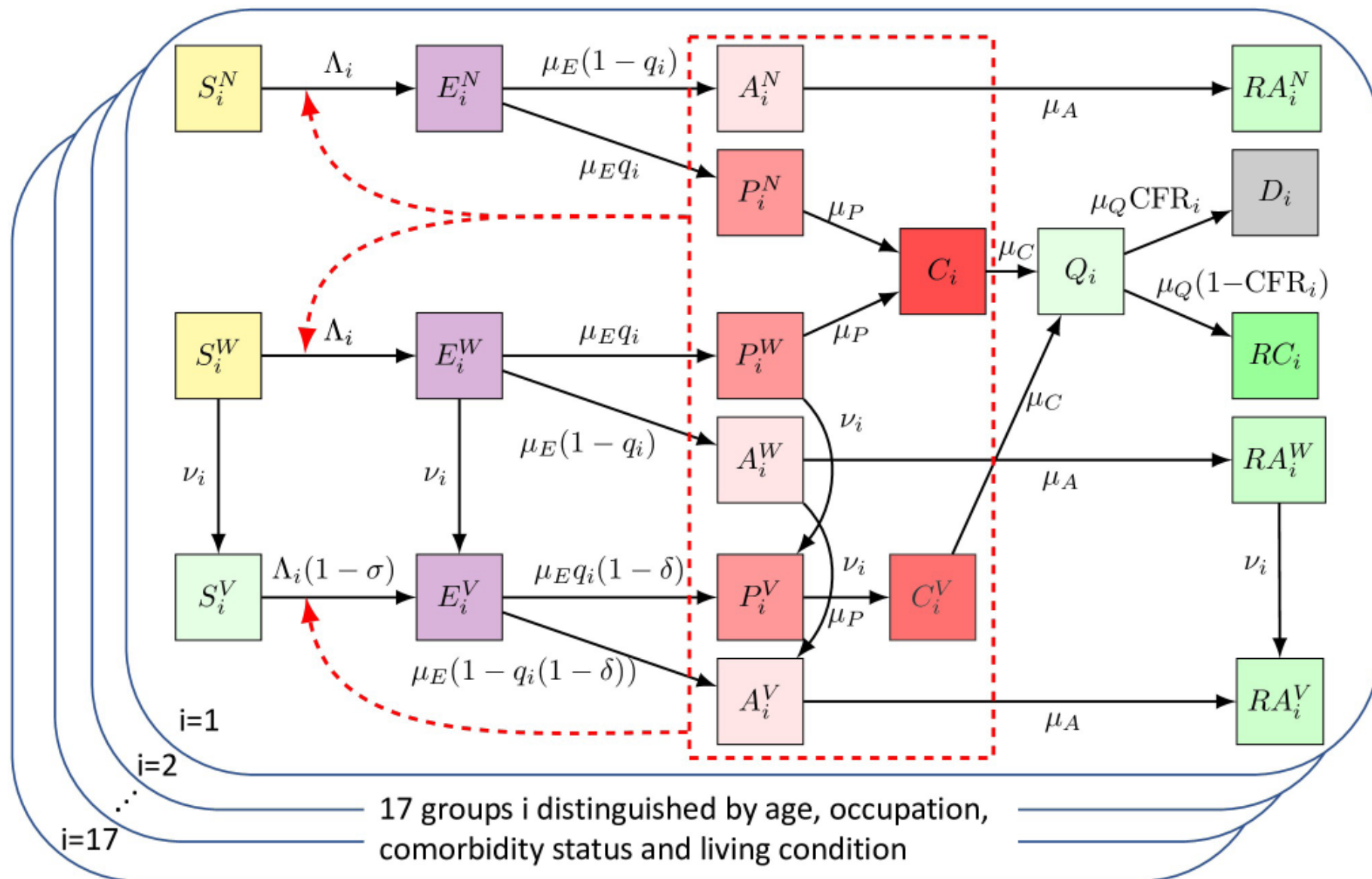
**D:** Dead

**N:** Not willing to be vaccinated

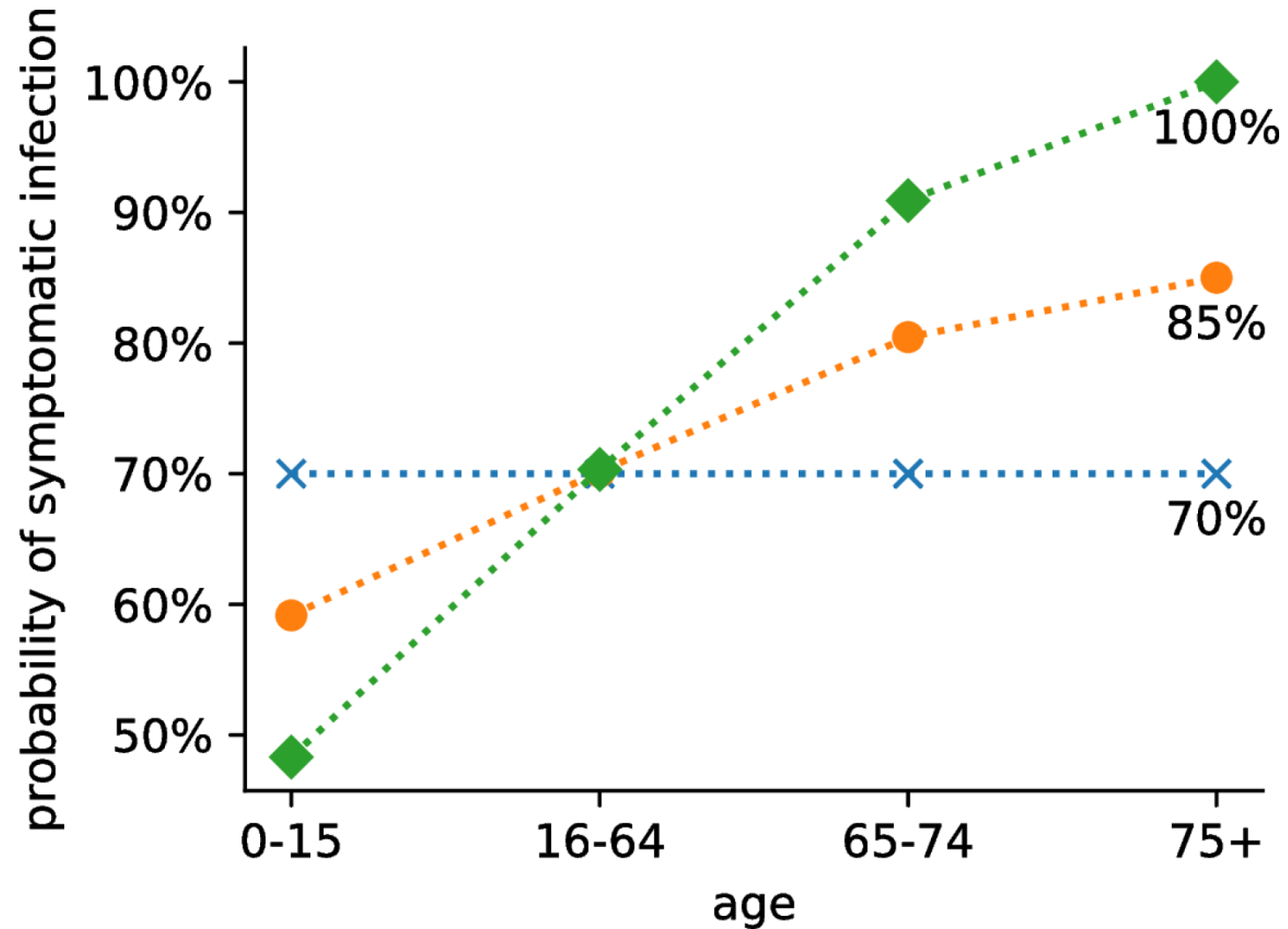
**W:** Willing to be vaccinated but not yet vaccinated

**V:** Vaccinated

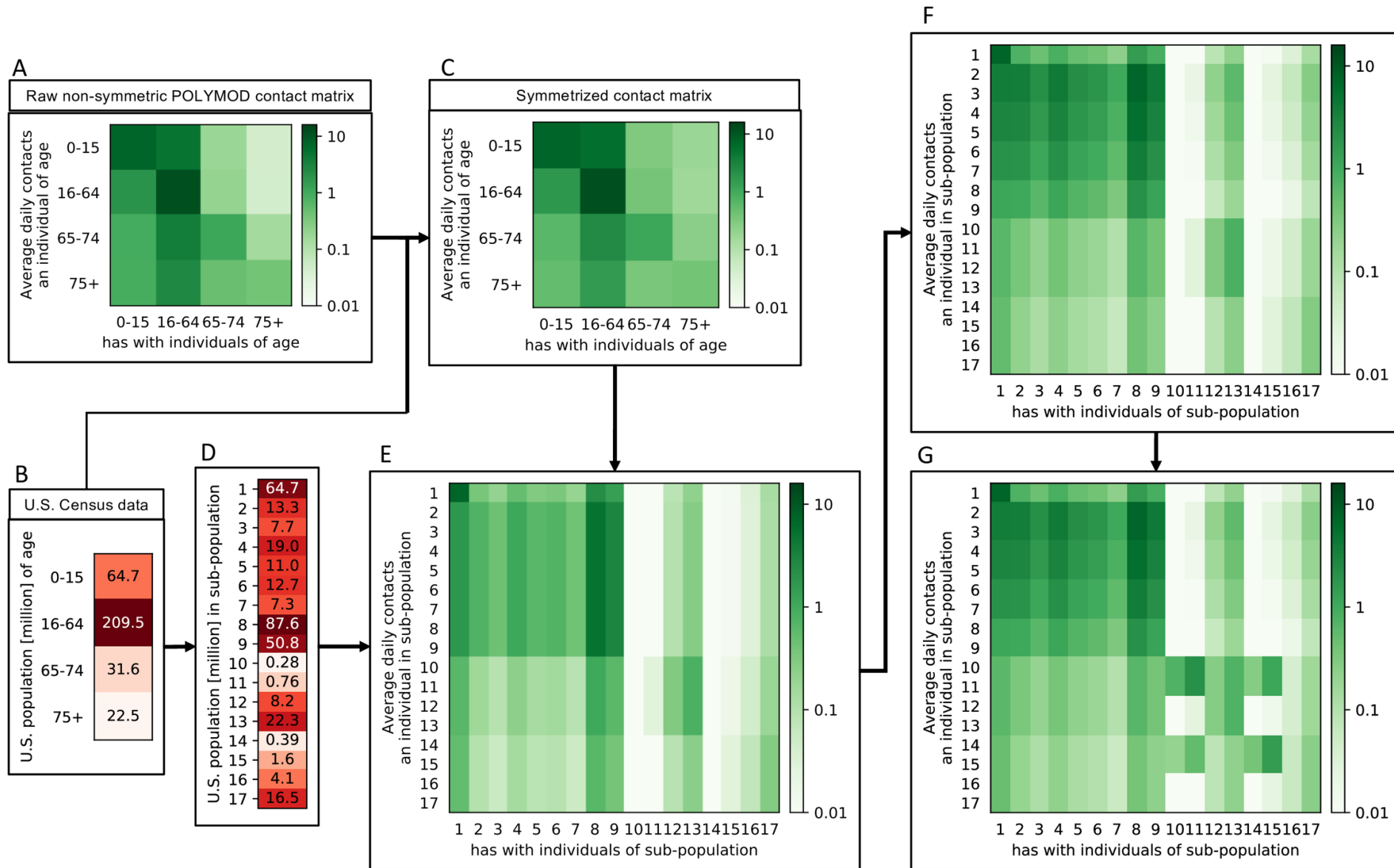
$S^N, S^W, S^V, E^N, E^W, E^V, A^N, A^W, A^V, RA^N, RA^W, RA^V, P^N, P^W, P^V, C, C^V, Q, RC, D$



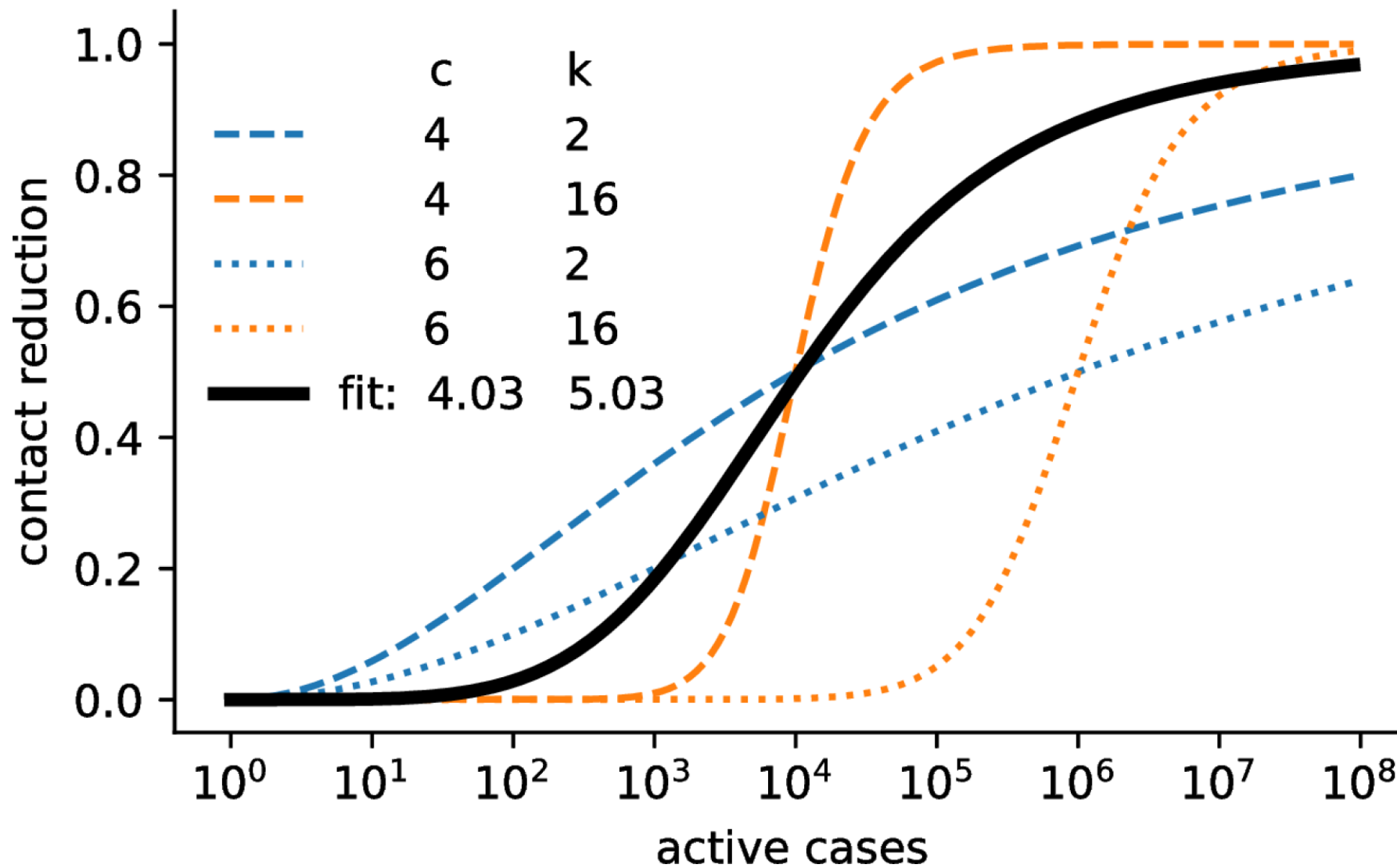
# Methods: Age-Dependent Susceptibility

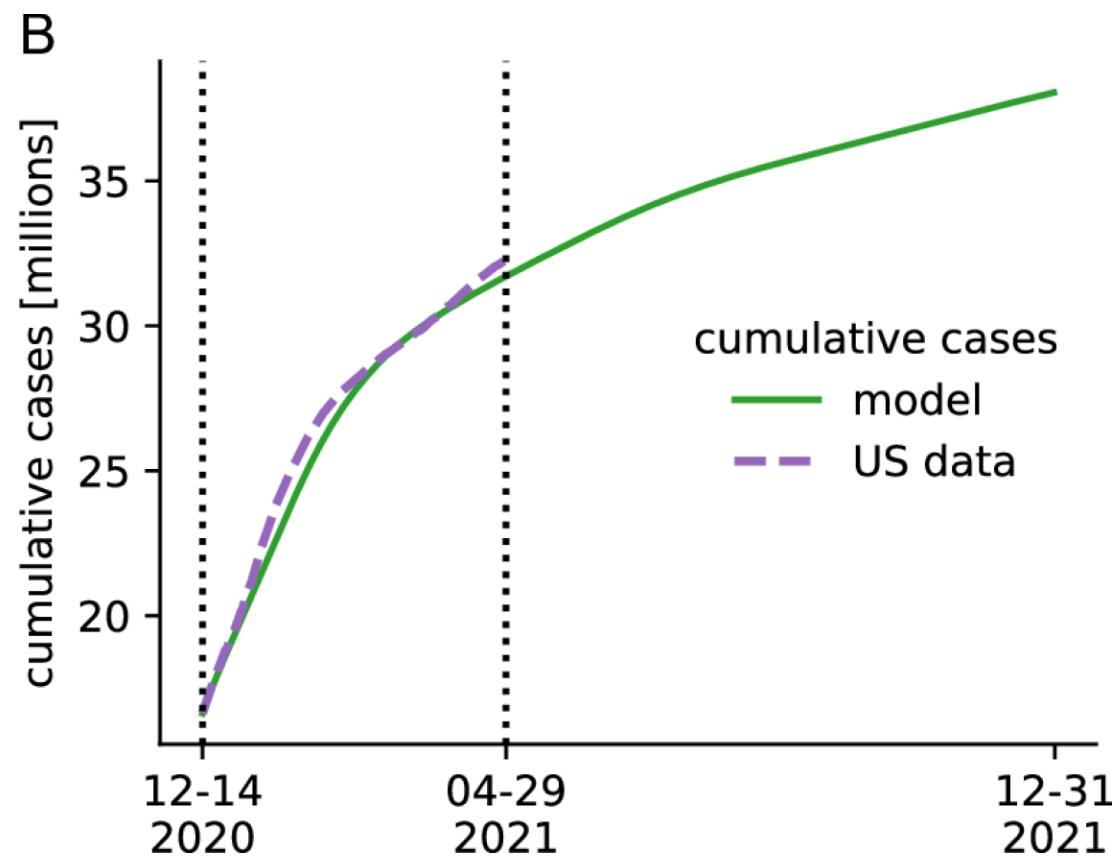
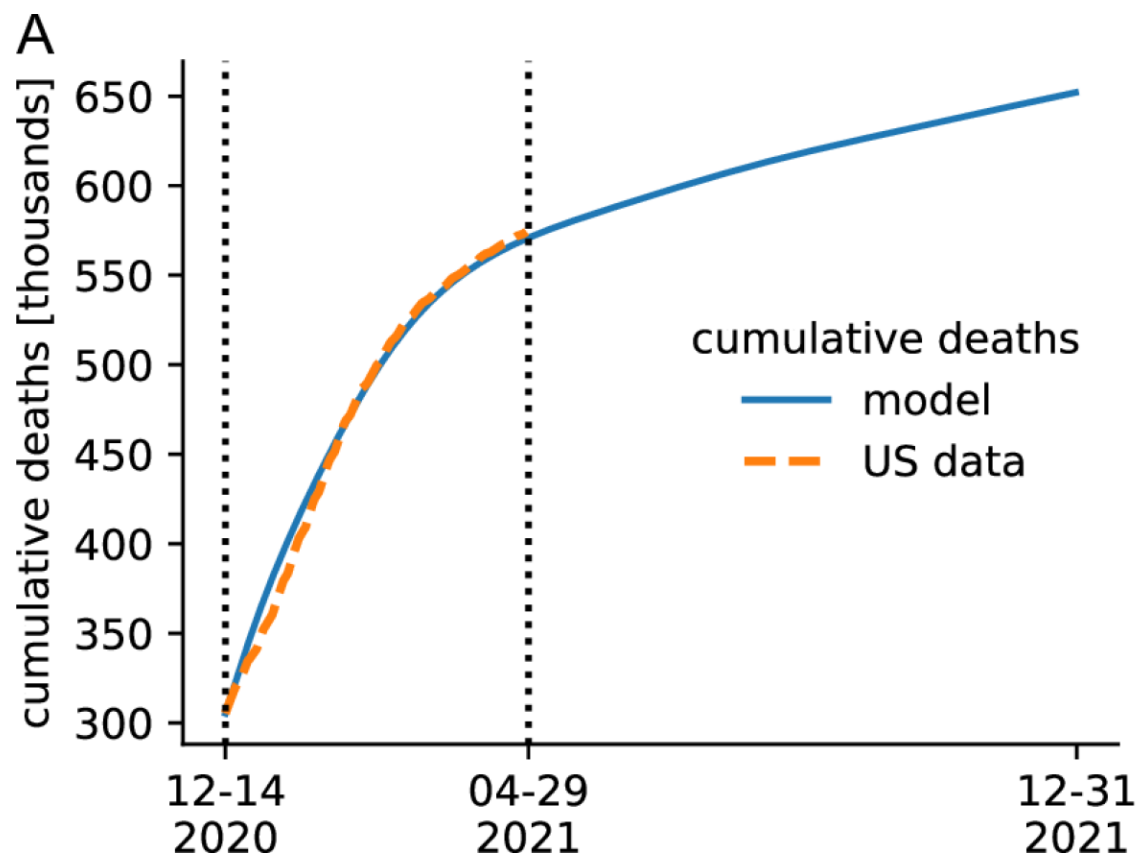


# Methods: Contact Matrix



# Methods: Case-Dependent Contact Reduction



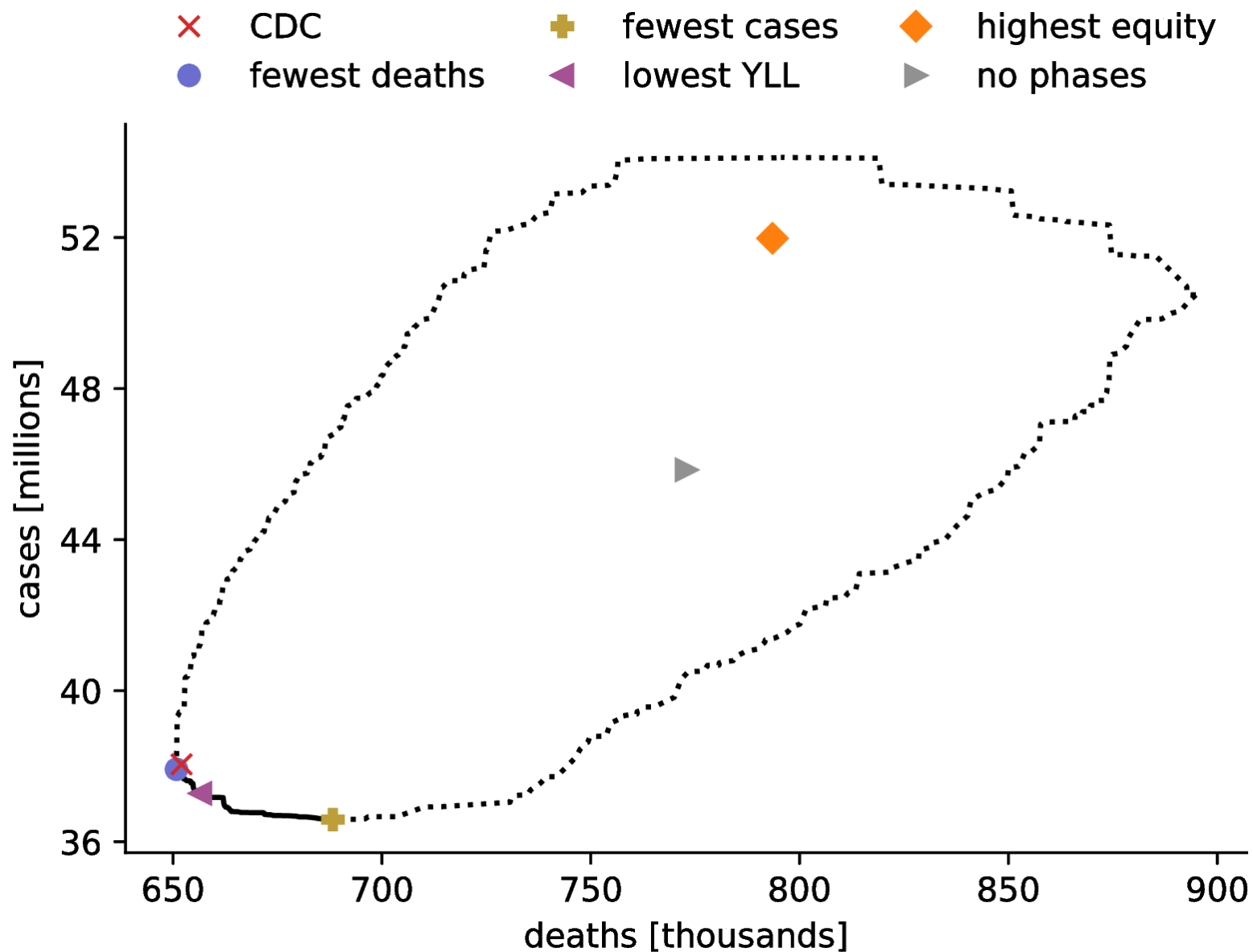


# Results: Comparison of CDC and Optimal Strategies

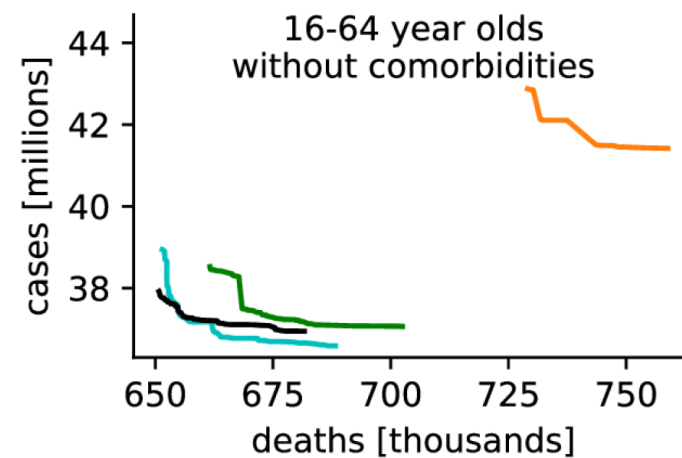
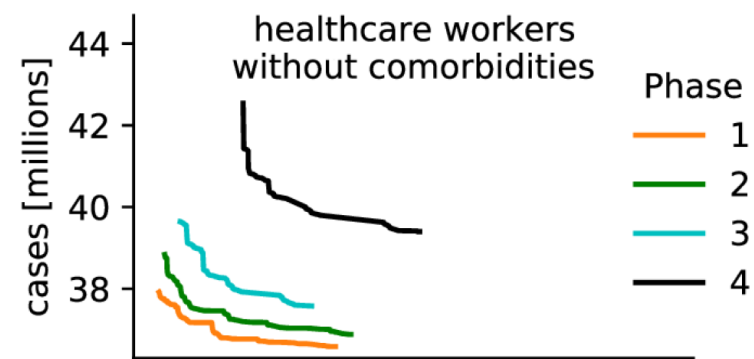
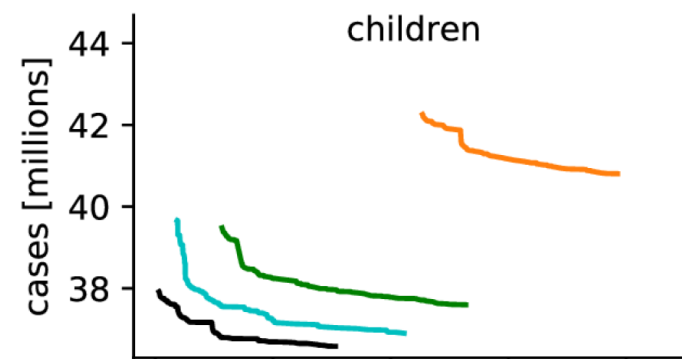
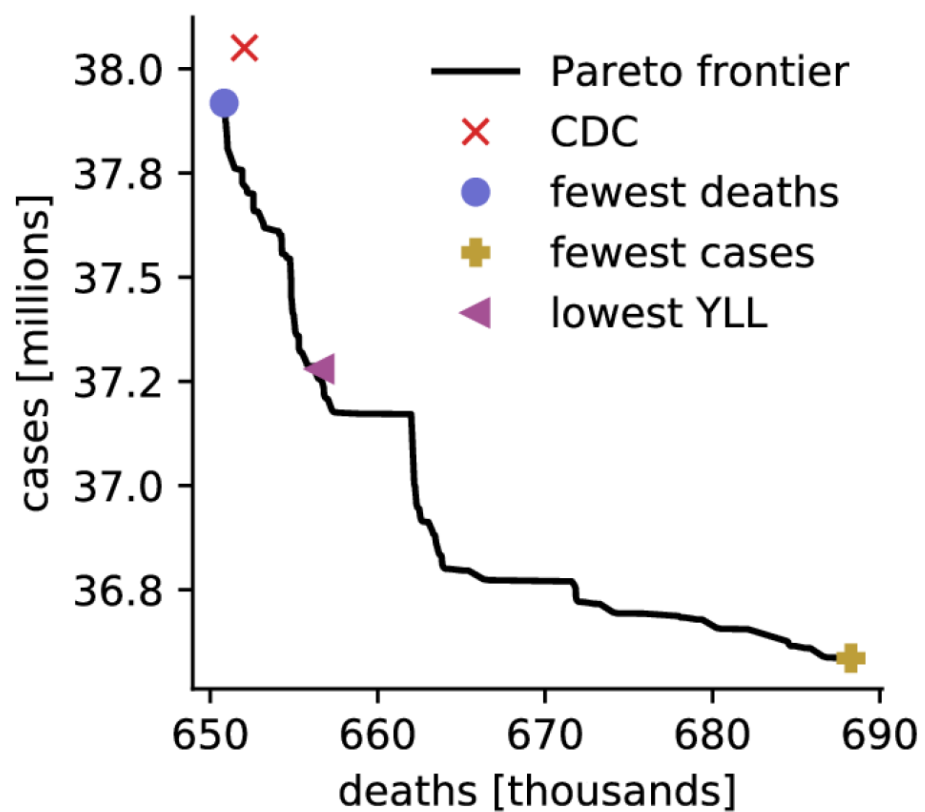


Age	Job / living situation	Comorbidity	Number of people [millions]	Sub-population ID in model	CDC allocation	fewest deaths [thousands]	lowest YLL [millions]	fewest cases [millions]	fewest infections [millions]
0-15	NA	NA	64.71	1	4	4	4	4	4
16-64	healthcare workers	no	13.29	2	1	1	1	1	1
		yes	7.71	3	1	1	1	1	1
	frontline essential workers	no	18.98	4	2	2	2	2	2
		yes	11.02	5	2	2	2	2	2
	other essential workers	no	12.66	6	3	3	3	2	2
		yes	7.34	7	3	3	2	2	2
	remaining people	no	87.61	8	4	4	4	3	3
		yes	50.85	9	3	3	3	3	3
65-74	congested living	no	0.28	10	1	2	3	3	3
		yes	0.76	11	1	1	2	3	3
	remaining people	no	8.20	12	3	3	4	4	4
		yes	22.34	13	3	3	3	4	4
75+	congested living	no	0.39	14	1	3	3	3	4
		yes	1.57	15	1	1	2	3	4
	remaining people	no	4.07	16	2	3	4	4	4
		yes	16.47	17	2	2	3	4	4
Respective outcome of specific allocation			CDC			652	11.6	38.1	56.4
			fewest deaths			651	11.6	37.9	56.2
			lowest YLL			657	11.5	37.3	55.3
			fewest cases			688	11.8	36.6	54.2
			fewest infections			695	11.9	36.6	54.2
% difference in outcome between specific and respective optimal allocation			CDC			0.19	0.97	4	4.07
			fewest deaths			0	0.67	3.64	3.74
			lowest YLL			0.88	0	1.9	2.08
			fewest cases			5.75	2.44	0	0.01
			fewest infections			6.73	2.84	0.03	0

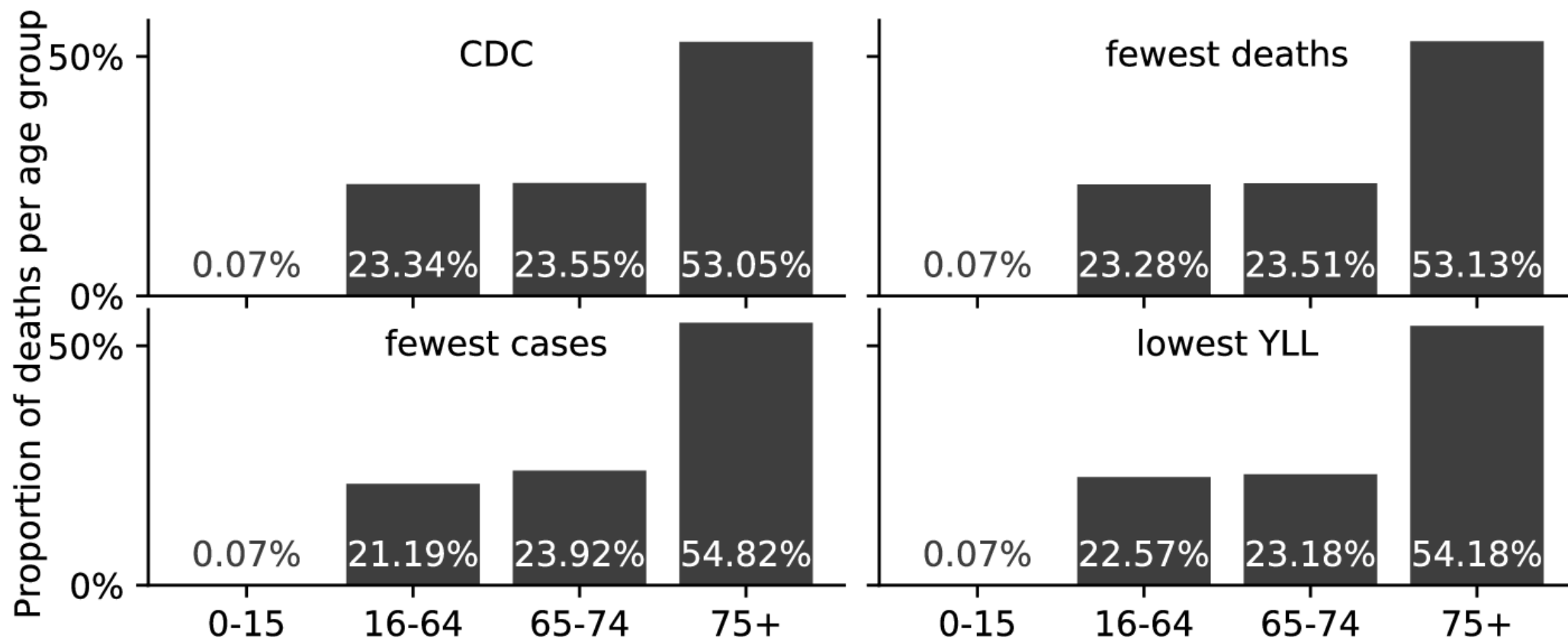
# Results: All Strategies



# Results: Pareto Frontier



# Results: Equity





1. All sub-populations exhibit the same level of vaccine hesitancy
2. Vaccine hesitance does not change over time
3. Uncertainty in key model parameters:
  1. Contact matrix
  2. Contagiousness of asymptomatic individuals and vaccinated individuals
4. No reinfections
5. Vaccine is fully effective immediately after 1 dose

Sensitivity analysis

# Extensions: Homophily & Contact Matrices



In the US, People of Color (POC) have been disproportionately affected by COVID-19



- As of 2022, POC comprise 34.1% of the US population, but suffered 42.2% of cases
- Studies from 2020:
  - Infection rate in predominately Black counties was 3x that of predominately white counties
  - Navajo Nation had more cases per capita than any US state
  - Black pop in Chicago is 30% but suffered over 50% of deaths
  - NYC 2x as many deaths per capita in Black and Latino pops than white pops
  - Across the US, the 20% of disproportionately Black counties account for 52% of cases and 58% of deaths

Why?

- Complex social and economic reasons, but:
  - POC tend to live in more crowded conditions
  - POC tend to work in more high-contact (high-risk) jobs

# Extensions: Trade-offs in Vaccination Priorities

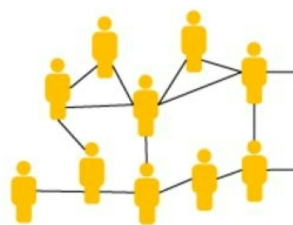


age	WA 	POC 	average daily contacts	CFR
0-15	55%	45%	14.0	0.01%
16-64	65%	35%	14.4	0.5%
... in high-contact jobs	$\geq 38\%$	$\leq 62\%$	14.4-28.7	0.5%
65-74	80%	20%	4.6	5.0%
75+	82%	18%	2.8	16.7%

Direct protection:  
Prioritize older people

Trade-off 1

Indirect protection:  
Prioritize active younger people



Homophily



WA are older:  
Prioritize WA

Trade-off 2

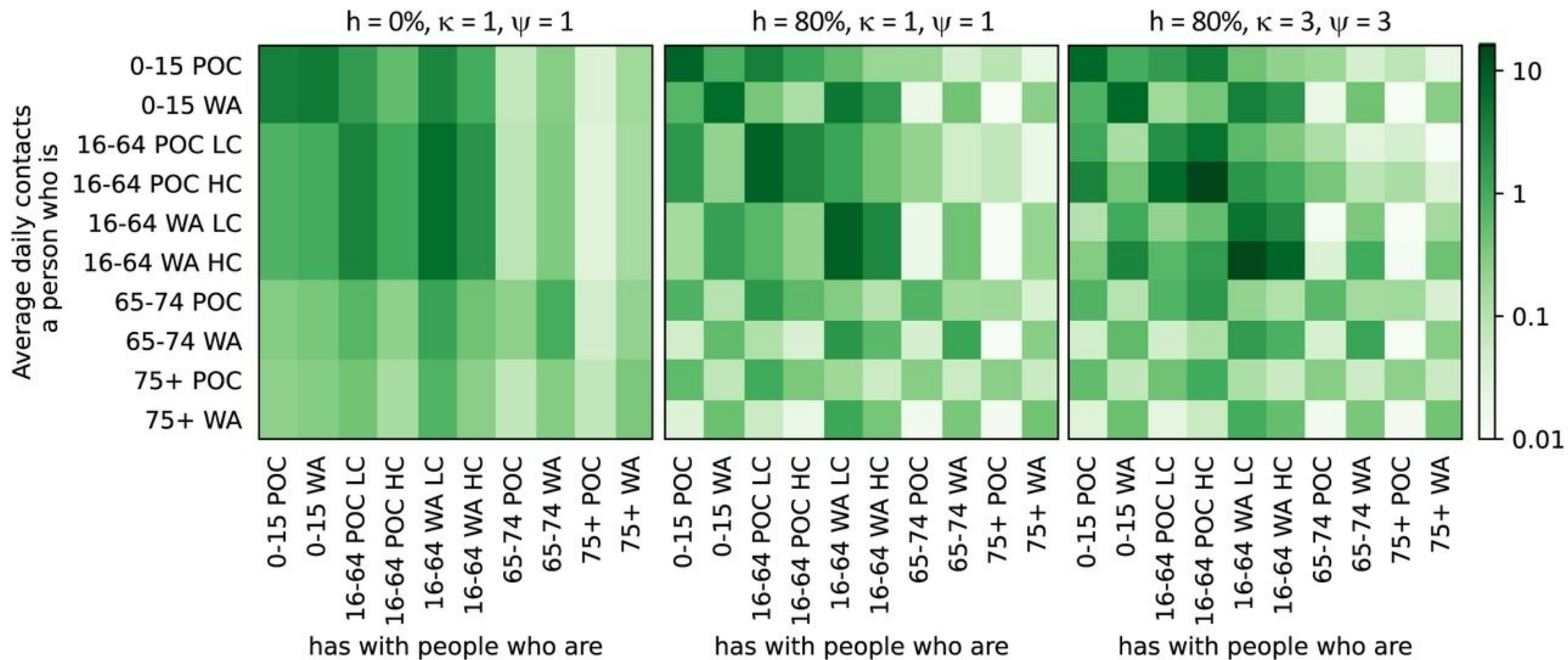
POC work more high-contact jobs:  
Prioritize POC

## Homophily:

The tendency of people from a particular demographic group to interact more frequently with people from the same group.

- Age
- Ethnicity
- Location
- Religion
- Political party
- Vaccination status
- Etc.

# Extensions: Contact Matrix and Parameters



## Excess predicted deaths:

Predicted deaths from best strategy compared to those under the best strategy that assigns WA and POC to the same allocation phase.

Scenario			Exact number of phases in vaccine allocation			
h	$\kappa$	$\psi$	2	3	4	5
80%	3	3	825	8774	11610	11226
	3	1	0	312	0	0
	1	1	0	0	0	570
0%	3	3	0	0	0	0
	3	1	0	0	0	0
	1	1	0	0	0	0



1. Optimal strategy depends on the goal of the vaccination campaign
2. The CDC allocation strategy was close to optimal
3. Allocation could be improved by prioritizing people with comorbidities in older populations
4. Accounting for ethnic homophily:
  1. Changes which strategy is optimal
  2. Better matches actual case counts and mortality
5. Essential model features:
  1. Levels of demographic homophily
  2. Case-dependent social distancing levels
  3. Age-dependent susceptibility
  4. Age-dependent clinical fraction
  5. Time-dependent transmission rate



Thank you





# Backup Slides



# Backup Slide: Strategies that Outperform CDC



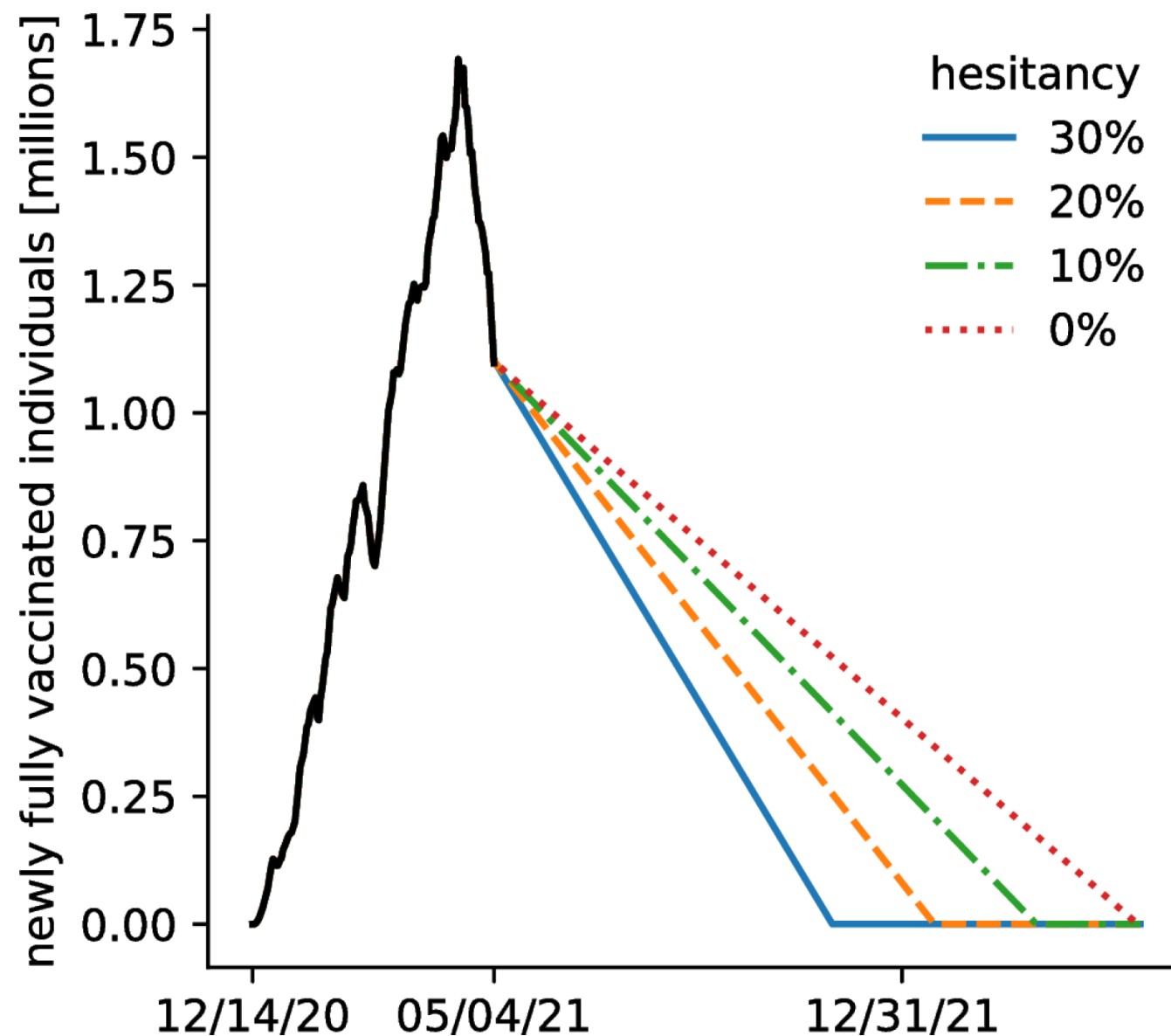
	deaths	cases	YLL	Phase assignment of sub-poulation																
	[thousands]	[millions]	[millions]	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17
CDC allocation	652.04	38.050	11.643	4	1	1	2	2	3	3	4	3	1	1	3	3	1	1	2	2
Allocations that Pareto-dominate the CDC allocation in all three metrics	650.83	37.918	11.608	4	1	1	2	2	3	3	4	3	2	1	3	3	3	1	3	2
	650.85	37.914	11.607	4	1	1	2	2	3	3	4	3	3	1	3	3	3	1	3	2
	650.90	37.881	11.606	4	1	1	2	2	3	3	4	3	2	2	3	3	2	1	3	2
	650.90	37.874	11.605	4	1	1	2	2	3	3	4	3	2	2	3	3	3	1	3	2
	650.94	37.870	11.604	4	1	1	2	2	3	3	4	3	3	2	3	3	3	1	3	2
	650.96	37.852	11.601	4	1	1	2	2	3	3	4	3	2	1	4	3	3	1	3	2
	650.98	37.848	11.600	4	1	1	2	2	3	3	4	3	3	1	4	3	3	1	3	2
	651.03	37.808	11.598	4	1	1	2	2	3	3	4	3	2	2	4	3	3	1	3	2
	651.07	37.805	11.597	4	1	1	2	2	3	3	4	3	3	2	4	3	3	1	3	2
	651.21	37.869	11.596	4	1	1	2	2	3	2	4	3	2	1	3	3	3	1	3	2
	651.24	37.865	11.596	4	1	1	2	2	3	2	4	3	3	1	3	3	3	1	3	2
	651.31	37.831	11.595	4	1	1	2	2	3	2	4	3	2	2	3	3	2	1	3	2
	651.31	37.824	11.593	4	1	1	2	2	3	2	4	3	2	2	3	3	3	1	3	2
	651.36	37.820	11.593	4	1	1	2	2	3	2	4	3	3	2	3	3	3	1	3	2
	651.40	37.811	11.591	4	1	1	2	2	3	2	4	3	2	1	4	3	3	1	3	2
	651.42	37.807	11.591	4	1	1	2	2	3	2	4	3	3	1	4	3	3	1	3	2
	651.43	37.765	11.592	4	1	1	2	2	3	3	4	3	2	2	4	3	3	1	4	2
	651.46	37.761	11.592	4	1	1	2	2	3	3	4	3	3	2	4	3	3	1	4	2
	651.50	37.773	11.590	4	1	1	2	2	3	2	4	3	2	2	4	3	2	1	3	2
	651.50	37.766	11.588	4	1	1	2	2	3	2	4	3	2	2	4	3	3	1	3	2
	651.54	37.763	11.588	4	1	1	2	2	3	2	4	3	3	2	4	3	3	1	3	2
	651.70	37.759	11.593	4	1	1	2	2	3	3	4	3	3	2	4	3	4	1	4	2
	651.81	37.772	11.587	4	1	1	2	2	3	2	4	3	2	1	4	3	3	1	4	2
	651.82	37.768	11.586	4	1	1	2	2	3	2	4	3	3	1	4	3	3	1	4	2
	651.90	37.757	11.590	4	1	1	2	2	3	3	4	3	2	1	4	3	2	2	3	2
	651.90	37.750	11.589	4	1	1	2	2	3	3	4	3	2	1	4	3	3	2	3	2
	651.91	37.734	11.585	4	1	1	2	2	3	2	4	3	2	2	4	3	2	1	4	2
	651.91	37.727	11.584	4	1	1	2	2	3	2	4	3	2	2	4	3	3	1	4	2
	651.95	37.723	11.584	4	1	1	2	2	3	2	4	3	3	2	4	3	3	1	4	2

## Backup Slide: Age- and Comorbidity-Dependent CFR



- CDC age-structured (death count / case count)
  - 0.0129%, 0.4533%, 4.9781%, 16.7279%
- US population-level estimates of comorbidity prevalence
  - 18.60%, 36.72%, 73.15%, 80.18%
- 51.71% cases with comorbidities (health insurance claims)
- 83.29% deaths with comorbidities (health insurance claims)
- Persons with comorbidities have 4.65x higher CFR
- **CFR without comorbidities:**
  - **0.0129%, 0.1935%, 1.1355%, 4.2560%**
- **CFR with comorbidities:**
  - **0.0129%, 0.8997%, 6.3012%, 19.7907%**

# Backup Slide: Speed of Vaccine Roll-Out

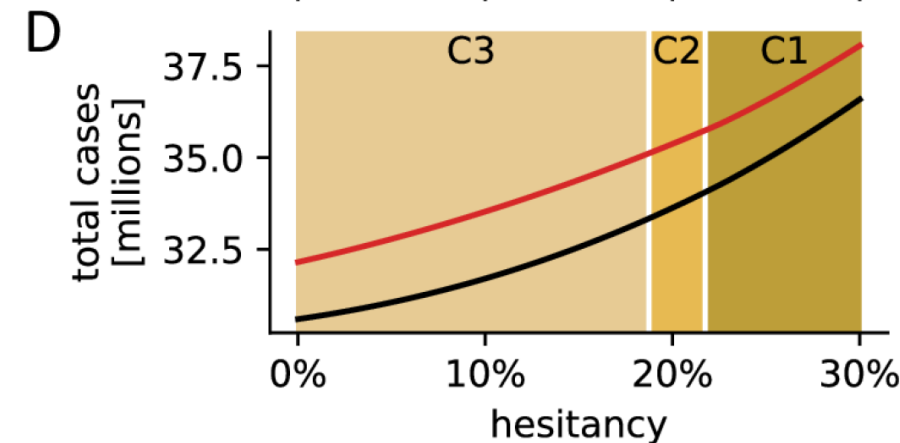
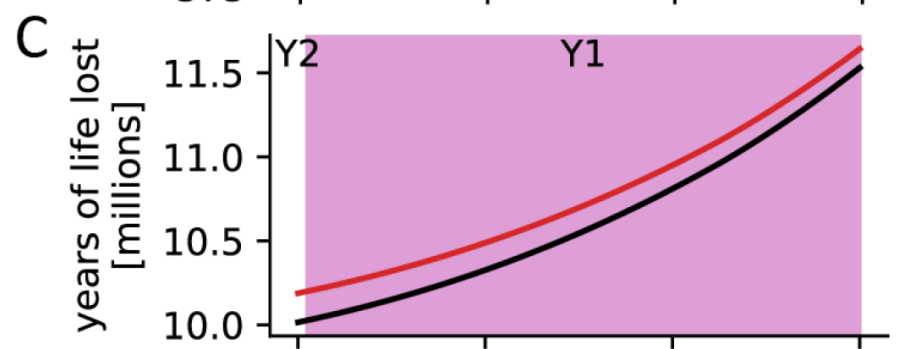
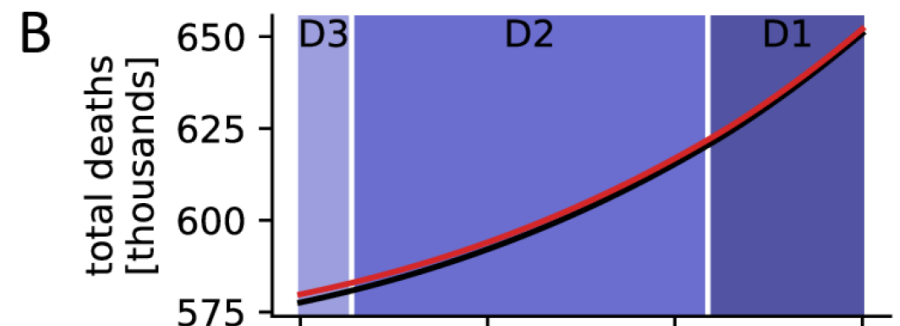


# Backup Slide: Vaccine Hesitancy

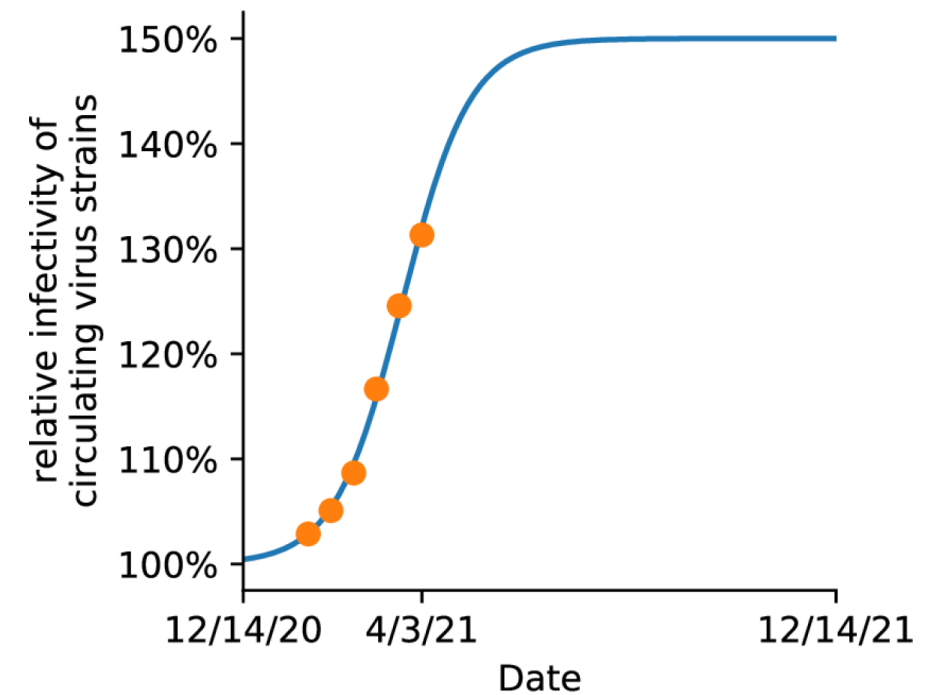
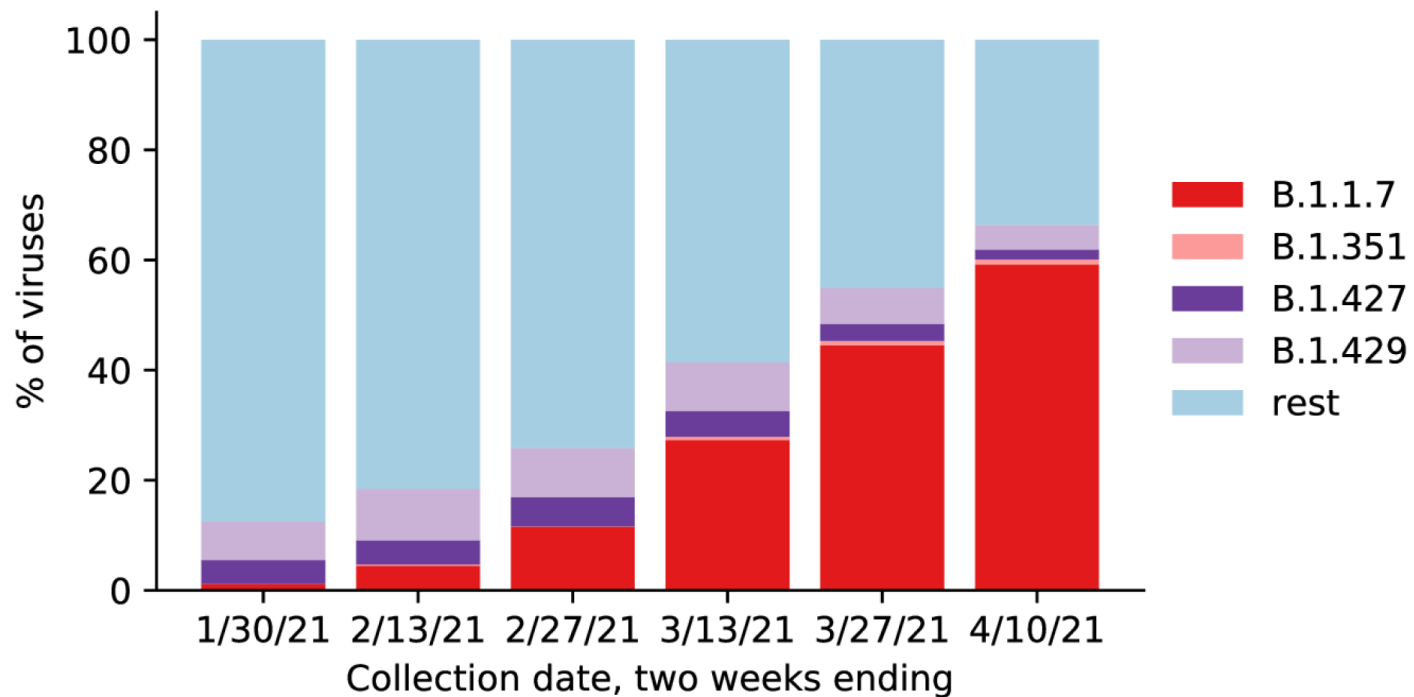


**A**

		objective				minimize deaths				minimize YLL				minimize cases				CDC
		hesitancy				0.3	0.2	0.1	0	0.3	0.2	0.1	0	0.3	0.2	0.1	0	
Sub-population	1	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4
	2	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
	3	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
	4	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
	5	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
	6	3	3	3	3	3	3	3	3	2	2	2	2	2	2	2	2	3
	7	3	3	3	3	2	2	2	3	2	2	2	2	2	2	2	2	3
	8	4	4	4	4	4	4	4	4	3	3	3	3	3	3	3	3	4
	9	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3
	10	2	2	2	3	3	3	3	3	3	3	3	3	3	3	3	3	1
	11	1	2	2	2	2	2	2	2	3	2	2	2	2	2	2	2	1
	12	3	3	3	4	4	4	4	4	4	4	4	4	4	4	4	4	3
	13	3	3	3	3	3	3	3	3	4	4	4	4	4	4	4	4	3
	14	3	3	3	3	3	3	3	3	3	4	3	3	3	3	3	3	1
	15	1	1	1	1	2	2	2	2	3	3	3	3	3	3	3	3	1
	16	3	3	3	3	4	4	4	4	4	4	4	4	4	4	4	4	2
	17	2	2	2	2	3	3	3	3	4	4	4	4	4	4	4	4	2
		↑	↑	↑		↑	↑	↑		↑	↑	↑		↑	↑	↑		↑
		D1	D2	D3		Y1	Y2			C1	C2	C3						CDC



# Backup Slide: Emergence of Variants



# Backup Slide: Model Parameters



Parameter	Description	Value	Source
$N_i$	number of people in sub-population $i$	see <a href="#">Table 2</a>	[25]
$X_{ij}$	average daily number of contacts a person in sub-population $i$ has with sub-population $j$	see <a href="#">S2G Fig</a>	[21, 22]
$c$	log10 value of active cases at which overall contacts are reduced by 50%	$c = 4.0346$ (see <a href="#">S1 Table</a> for fitted values used in the sensitivity analysis)	fitted (see Model calibration)
$k$	sensitivity of contact reduction to changes in active cases (shape of the Hill function)	$k = 5.0266$ (see <a href="#">S1 Table</a> for fitted values used in the sensitivity analysis)	fitted (see Model calibration)
$\beta_i$	age-dependent susceptibility to infection	see <a href="#">S1 Table</a>	fitted (see Model calibration)
$1/\mu_E$	incubation period	3.7 days	[26]
$q_i$	age-dependent clinical fraction	varied, see <a href="#">S1 Fig</a>	[27]
$1/\mu_A$	average time of virus spread by truly asymptomatic individuals	5 days	[17]
$1/\mu_P$	average time of virus spread before symptom onset	2.1 days	[17]
$1/\mu_C$	average time of virus spread after symptom onset	2.723 days	estimated from CDC raw data
$1/\mu_Q + 1/\mu_C$	average time between symptom onset and possible death	22 days	estimated from U.S. deaths and case counts [28]
$CFR_i$	sub-population-dependent case fatality ratio	see Case fatality rates	calculated from [29, 30]
$f_A$	relative contagiousness of truly asymptomatic individuals	75% (25% and 100% in sensitivity analysis)	[27]
$f_V$	relative contagiousness of vaccinated individuals	50% (0% and 100% in sensitivity analysis)	no data
none	vaccine hesitancy	30%	[31, 32]
$\xi(t)$	daily number of available vaccines	see <a href="#">S4 Fig</a>	[33]
none	vaccine effectiveness: reduction of symptomatic infections among vaccinated (compared to non-vaccinated)	90%	[34]
$\sigma$ and $\delta$	reduction in infections and symptomatic infections (when infected) among vaccinated (compared to non-vaccinated) individuals	70% and 66.7% (varied such that $1 - (1 - \sigma)(1 - \delta) = 90\%$ in sensitivity analysis)	[35]

# Backup Slide: Genetic Algorithm



Minimize fitness function:

- 50 iterations of 1000 parameter sets
- 300 parents, 700 children
- 50% crossover probability
- 10% random mutation probability
- 1% elite ratio (top 10 par sets)
- 100 separate runs
- Weighting ensures good fit at end

$$f(\text{deaths, cases}) = \text{wSSE}(\text{deaths}) + \text{wSSE}(\text{cases})$$

where

$$\text{wSSE}(\text{deaths}) = \sum_{d=\text{December 14, 2020}}^{\text{April 29, 2021}} w_d \cdot (\text{observed minus predicted deaths up to day } d)^2$$

$$\text{wSSE}(\text{cases}) = \sum_{d=\text{December 14, 2020}}^{\text{April 29, 2021}} w_d \cdot (\text{observed minus predicted cases up to day } d)^2$$

Quadratically-increasing weights:

$$w_{\text{December 14 2020}} = 1, w_{\text{December 15 2020}} = 4, w_{\text{December 16 2020}} = 9, \dots$$

# Backup Slide: Model Parameters



Varied parameters			Fitted parameters				wSSE
$f_A$	$f_V$	$q_{75+}$	$b_0$	$b_1$	$c$	$k$	
0.75	0.5	0.85	0.0397	0.0044	4.0346	5.0266	4.66E+13
0.75	0.5	0.7	0.0001	0.0045	4.2812	5.2282	5.56E+13
0.75	0.5	1	0.0810	0.0031	4.0382	4.9903	4.54E+13
0.25	0.5	0.85	0.0727	0.0072	4.1001	5.9356	4.33E+13
1	0.5	0.85	0.0249	0.0030	4.2441	4.8065	4.80E+13
0.75	0	0.85	0.0764	0.0081	4.3286	7.3542	3.58E+13
0.75	1	0.85	0.0157	0.0017	4.0277	2.8875	5.83E+13

# Backup Slide: Sensitivity Analysis



Age	Job or living situation	Comorbidity	Number of people [millions]	Group ID in model	CDC allocation	deaths [thousands]							YLL [millions]							cases [millions]							q <sub>75+</sub>		
						0.85	0.7	1	0.85	0.85	0.85	0.85	0.85	0.7	1	0.85	0.85	0.85	0.85	0.85	0.7	1	0.85	0.85	0.85	0.85		f <sub>A</sub>	
						0.75	0.75	0.75	0.25	1	0.75	0.75	0.75	0.25	1	0.75	0.75	0.75	0.25	1	0.75	0.75	0.75	0.25	1	0.75		0.75	f <sub>A</sub>
						0.5	0.5	0.5	0.5	0.5	0	1	0.5	0.5	0.5	0.5	0.5	0	1	0.5	0.5	0.5	0.5	0.5	0	1		f <sub>V</sub>	
0-15	NA	NA	64.71	1	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4					
16-64	healthcare workers	no	13.29	2	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1				
		yes	7.71	3	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1				
	frontline essential workers	no	18.98	4	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2				
		yes	11.02	5	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2				
	other essential workers	no	12.66	6	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3				
		yes	7.34	7	3	3	3	2	2	3	3	3	3	2	2	2	2	2	2	2	2	2	2	2	2				
	remaining people	no	87.61	8	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4				
		yes	50.85	9	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3	3				
65-74	congested living	no	0.28	10	1	2	2	2	2	2	2	3	3	3	3	3	3	3	3	3	3	3	3	3	3				
		yes	0.76	11	1	1	1	1	1	1	1	1	2	2	2	2	2	2	2	2	3	2	3	2	3				
	remaining people	no	8.20	12	3	3	3	3	3	3	3	4	4	4	4	4	4	4	4	4	4	4	4	4	4				
		yes	22.34	13	3	3	3	2	2	3	3	3	3	3	3	3	3	3	3	3	4	4	4	4	4				
75+	congested living	no	0.39	14	1	3	3	2	2	3	2	3	3	3	3	3	3	3	3	3	3	4	3	3	3	4			
		yes	1.57	15	1	1	1	1	1	1	1	1	2	2	2	2	2	2	2	2	3	3	3	3	3				
	remaining people	no	4.07	16	2	3	3	3	3	3	3	3	4	4	4	4	4	4	4	4	4	4	4	4	4				
		yes	16.47	17	2	2	2	1	1	2	2	2	3	3	3	3	3	3	3	3	4	4	4	4	4				
Value for optimal allocation						650.8	645.5	651.6	658.5	647.3	661.7	620.4	11.53	11.55	11.5	11.69	11.46	11.75	10.92	36.59	36.58	36.78	37.28	36.21	37.57	33.96			
Value for CDC allocation						652	646.6	653.2	659.7	648.6	662.9	622	11.64	11.67	11.61	11.79	11.57	11.86	11.05	38.05	38.08	38.2	38.74	37.69	39.08	35.4			
% difference						0.187	0.182	0.235	0.181	0.192	0.185	0.249	0.974	1	0.984	0.892	1.012	0.971	1.167	4.003	4.104	3.869	3.909	4.066	4.029	4.244			



## Backup Slide: Ethnic Homophily Model



- 10 subpopulations:
  - 4 age groups
  - 2 ethnicity groups (WA and POC)
  - 2 occupation levels for age group 16-64 (HC and LC)
- 3 parameters (built into contact matrix):
  - $h$ : ethnic homophily (0% & 80%)
  - $\psi$ : relative proportion of POC (vs. WA) in high-contact jobs (1 & 3)
  - $\kappa$ : relative contact levels for employees in high-contact jobs vs. low-contact jobs (1 & 3)
- Varying number of allocation phases:
  - 1-5, and 10
- Global optimization approach over 2.9 million possibly optimal strategies

## Backup Slide: Calculating Relative Homophily



$p \in (0, 1)$  Proportion of population in WA group

$\phi \in [0, 1]$  Proportion of all contacts between people of the same ethnicity group

$$\mathbb{E}(\phi) = p^2 + (1 - p)^2$$

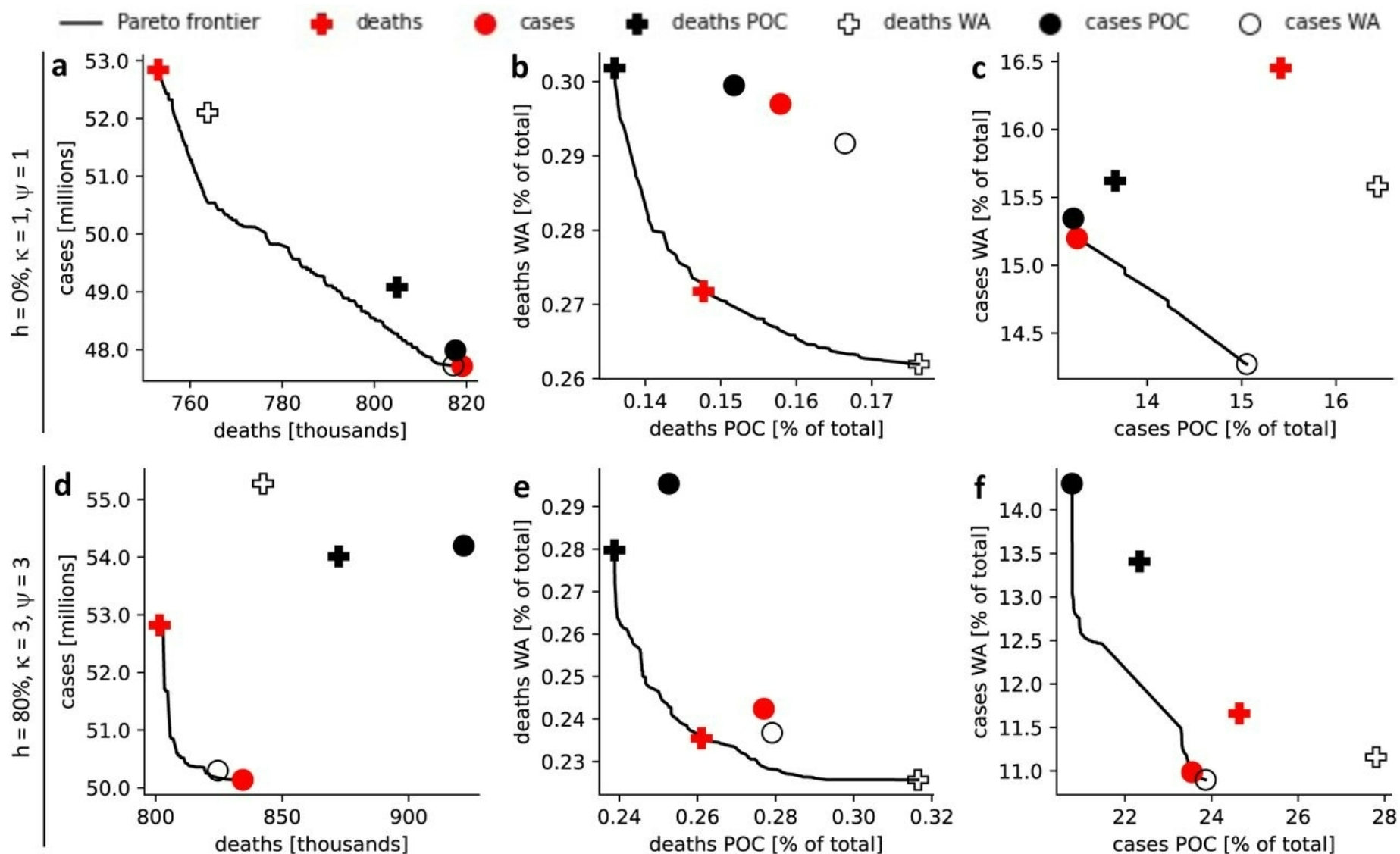
$$\text{homophily } h = \begin{cases} \frac{\phi - \mathbb{E}(\phi)}{1 - \mathbb{E}(\phi)} \in [0, 1] & \text{if } \phi \geq \mathbb{E}(\phi), \\ \frac{\phi - \mathbb{E}(\phi)}{\mathbb{E}(\phi)} \in [-1, 0) & \text{if } \phi < \mathbb{E}(\phi). \end{cases}$$

If  $p = \frac{2}{3}$ , then  $E(\phi) = \frac{5}{9}$ .  $\phi = \frac{7}{9}$  therefore corresponds to 50% homophily.

$\phi = 1$  corresponds to 100% homophily: complete segregation of WA and POC.

$\phi = 0$  corresponds to 100% heterophily, where the contact graph is bipartite.

# Backup slide: Results from Ethnic Homophily Study



# Backup Slide: Results from Ethnic Homophily Study



exact number of phases			4						5						10					
age	ethnicity	occupation	3	1	1	3	1	1	3	1	1	3	1	1	3	1	1	3	1	1
0-15	POC	n.a.	4	4	4	4	4	4	5	5	5	5	5	4	10	10	10	9	9	9
	WA	n.a.	4	4	4	4	4	4	5	5	5	5	5	5	8	9	9	10	10	10
16-64	POC	low-contact	4	4	3	4	4	3	5	4	4	4	4	3	9	8	8	7	7	6
		high-contact	2	2	3	2	2	3	2	2	4	2	2	3	4	4	6	3	3	5
	WA	low-contact	3	4	3	4	4	3	4	4	4	4	4	3	7	7	7	8	8	8
		high-contact	1	2	3	2	2	3	1	2	3	2	2	3	2	3	5	4	4	7
65-74	POC	n.a.	3	3	2	3	3	2	3	3	2	3	3	2	5	5	4	5	5	3
	WA	n.a.	3	3	2	3	3	2	3	3	2	3	3	2	6	6	3	6	6	4
75+	POC	n.a.	1	1	1	1	1	1	1	1	1	1	1	1	1	2	2	1	1	1
	WA	n.a.	2	1	1	1	1	1	2	1	1	1	1	1	3	1	1	2	2	2

## Backup Slide: Citations



- Islam, M.R., Oraby, T., McCombs, A., Chowdhury, M.M., Al-Mamun, M., Tyshenko, M.G. and Kadelka, C., 2021. Evaluation of the United States COVID-19 vaccine allocation strategy. *PloS one*, 16(11), p.e0259700.
- Kadelka, C., Islam, M.R., McCombs, A., Alston, J. and Morton, N., 2022. Ethnic homophily affects vaccine prioritization strategies. *medRxiv*.
- Kadelka, C., 2022. Projecting social contact matrices to populations stratified by binary attributes with known homophily. *arXiv preprint arXiv:2207.12328*.
- Kadelka, C. and McCombs, A., 2021. Effect of homophily and correlation of beliefs on COVID-19 and general infectious disease outbreaks. *PloS one*, 16(12), p.e0260973.