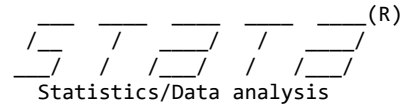


## Impact of non-pharmaceutical interventions against COVID-19

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Iteration 7: log likelihood = **-9080.8363**  
 name: <unnamed>  
 log: C:\Users\PaulH\Dropbox\001Covid19 files\Sensitivity analyses with varying time gaps.smcl  
 log type: smcl  
 opened on: 28 Feb 2021, 18:00:34

1 . menbreg cases ib0.mass ib0.business ib0.education ib0.non\_essential ib0.home ib0.masks dayfromfirst tests1mpop16ap

Fitting fixed-effects model:

Iteration 0: log likelihood = **-9994.792** (not concave)  
 Iteration 1: log likelihood = **-9728.5026**  
 Iteration 2: log likelihood = **-9423.4215**  
 Iteration 3: log likelihood = **-9289.1035**  
 Iteration 4: log likelihood = **-9287.8567**  
 Iteration 5: log likelihood = **-9287.854**  
 Iteration 6: log likelihood = **-9287.854**

Refining starting values:

Grid node 0: log likelihood = **-9150.7276**

Fitting full model:

Iteration 0: log likelihood = **-9150.7276** (not concave)  
 Iteration 1: log likelihood = **-9144.1122** (not concave)  
 Iteration 2: log likelihood = **-9137.8038**  
 Iteration 3: log likelihood = **-9107.7886**  
 Iteration 4: log likelihood = **-9081.8283**  
 Iteration 5: log likelihood = **-9080.841**  
 Iteration 6: log likelihood = **-9080.8363**  
 Iteration 7: log likelihood = **-9080.8363**

Mixed-effects nbinomial regression  
 Overdispersion: mean  
 Group variable: country\_code  
 Number of obs = 1,588  
 Number of groups = 30  
 Obs per group:  
 min = 42  
 avg = 52.9  
 max = 63

Integration method: mvaghermite  
 Integration pts. = 7

Log likelihood = **-9080.8363**  
 Wald chi2(38) = 3579.53  
 Prob > chi2 = 0.0000

cases	IRR	Std. Err.	z	P> z	[95% Conf. Interval]
<b>mass</b>					
1	1.317029	.1196723	3.03	0.002	1.102175 1.573765
2	1.125023	.1379665	0.96	0.337	.8846566 1.430698
3	.9909768	.1532647	-0.06	0.953	.7318406 1.34187
4	.7990745	.1476335	-1.21	0.225	.5563179 1.147761
5	.7410227	.1605023	-1.38	0.166	.484691 1.132917
6	.6604644	.1673839	-1.64	0.102	.4019077 1.085357
<b>business</b>					
1	1.183943	.1265193	1.58	0.114	.9602165 1.459797
2	.8705053	.1228561	-0.98	0.326	.6601456 1.147898
3	.689605	.1169872	-2.19	0.028	.4945376 .9616155
4	.6074017	.1245851	-2.43	0.015	.4063355 .907961
5	.4714363	.1140889	-3.11	0.002	.2933795 .7575586
6	.3212351	.0919537	-3.97	0.000	.1833016 .562963
<b>education</b>					
1	1.47373	.1447278	3.95	0.000	1.215698 1.786529
2	1.376326	.1895708	2.32	0.020	1.050702 1.802865
3	.9470467	.1642194	-0.31	0.754	.6741735 1.330366
4	.5207754	.1083542	-3.14	0.002	.3463752 .7829864
5	.2582716	.062911	-5.56	0.000	.1602275 .4163096
6	.1433753	.0402787	-6.91	0.000	.0826694 .2486591

non_essent~1						
1	1.142639	.1234961	1.23	0.217	.9245092	1.412234
2	1.146635	.1439661	1.09	0.276	.8965042	1.466554
3	1.018911	.1389382	0.14	0.891	.7799502	1.331085
4	.8253778	.1312559	-1.21	0.228	.6043524	1.127237
5	.759917	.1443942	-1.44	0.148	.5236327	1.102822
6	.7628685	.1966775	-1.05	0.294	.4602552	1.264447
home						
1	1.194981	.1243872	1.71	0.087	.9744476	1.465426
2	1.951905	.224044	5.83	0.000	1.558676	2.444339
3	2.275837	.2798113	6.69	0.000	1.788492	2.895979
4	2.550771	.3529303	6.77	0.000	1.944901	3.345381
5	2.489913	.4261743	5.33	0.000	1.780296	3.48238
6	2.393275	.5782635	3.61	0.000	1.490476	3.842908
masks						
1	.6583008	.0615543	-4.47	0.000	.5480662	.7907074
2	.528956	.0566382	-5.95	0.000	.4288218	.6524726
3	.5206164	.0675501	-5.03	0.000	.4037144	.6713693
4	.6819353	.1249588	-2.09	0.037	.4761763	.976604
5	1.146066	.2871408	0.54	0.586	.7013678	1.872724
6	1.059821	.3458243	0.18	0.859	.5590897	2.009016
dayfromfirst	1.138656	.0074957	19.72	0.000	1.124059	1.153442
tests1mpop~r	1.000055	8.54e-06	6.46	0.000	1.000038	1.000072
_cons	2.93e-07	5.09e-08	-86.48	0.000	2.08e-07	4.12e-07
ln(pop~2018)		1 (exposure)				
/lnalpha	-.7180626	.0377877			-.7921251	-.644
country_code						
var(_cons)	.2640212	.0740418			.1523803	.4574554

Note: Estimates are transformed only in the first equation.

Note: \_cons estimates baseline incidence rate (conditional on zero random effects).

LR test vs. nbinomial model:  $\chi^2(01) = 414.04$  Prob >=  $\chi^2 = 0.0000$

2 . menbreg cases ib0.mass4 ib0.business4 ib0.education4 ib0.non\_essential4 ib0.home4 ib0.masks4 dayfromfirst tests1mpop~r

Fitting fixed-effects model:

Iteration 0: log likelihood = -9986.8369 (not concave)  
 Iteration 1: log likelihood = -9718.087  
 Iteration 2: log likelihood = -9455.4816  
 Iteration 3: log likelihood = -9260.6121  
 Iteration 4: log likelihood = -9258.192  
 Iteration 5: log likelihood = -9258.184  
 Iteration 6: log likelihood = -9258.184

Refining starting values:

Grid node 0: log likelihood = -9117.515

Fitting full model:

Iteration 0: log likelihood = -9117.515 (not concave)  
 Iteration 1: log likelihood = -9110.9696 (not concave)  
 Iteration 2: log likelihood = -9104.6382  
 Iteration 3: log likelihood = -9059.2946  
 Iteration 4: log likelihood = -9020.1352  
 Iteration 5: log likelihood = -9015.6862  
 Iteration 6: log likelihood = -9014.9553  
 Iteration 7: log likelihood = -9014.9141  
 Iteration 8: log likelihood = -9014.914

Mixed-effects nbinomial regression Number of obs = 1,588

Overdispersion: mean

Group variable: country\_code Number of groups = 30

Obs per group:  
 min = 42  
 avg = 52.9  
 max = 63

Integration method: mvaghermite Integration pts. = 7

Log likelihood = -9014.914 Wald chi2(74) = 4123.30  
 Prob > chi2 = 0.0000

cases	IRR	Std. Err.	z	P> z	[95% Conf. Interval]
mass4					
1	1.185258	.1176035	1.71	0.087	.9757871 1.439696
2	.9096095	.1069167	-0.81	0.420	.7224431 1.145266
3	.7401525	.1030979	-2.16	0.031	.5633198 .9724951
4	.6979264	.112031	-2.24	0.025	.5095362 .9559701
5	.5606689	.1017057	-3.19	0.001	.3929144 .8000461
6	.4650396	.0938509	-3.79	0.000	.313117 .6906741
7	.3496517	.0782323	-4.70	0.000	.2255198 .542109
8	.3327507	.0814023	-4.50	0.000	.2060088 .5374675
9	.3380323	.0910834	-4.03	0.000	.1993428 .5732128
10	.2857476	.0835348	-4.28	0.000	.1611181 .5067817
11	.2210622	.0707774	-4.71	0.000	.1180282 .414041
12	.210928	.0782178	-4.20	0.000	.1019732 .4362974
13	.1535516	.0684175	-4.21	0.000	.0641191 .3677234
14	.0618684	.0348283	-4.94	0.000	.0205255 .1864855
business4					
1	1.196626	.1353321	1.59	0.112	.9587203 1.493567
2	.9023284	.126786	-0.73	0.465	.6851133 1.188411
3	.789744	.1262493	-1.48	0.140	.5773143 1.08034
4	.6610521	.1180563	-2.32	0.020	.465823 .9381029
5	.5675012	.1113741	-2.89	0.004	.3862918 .833716
6	.583267	.1282284	-2.45	0.014	.3790835 .8974285
7	.4548825	.1127513	-3.18	0.001	.279842 .7394106
8	.4323347	.1172199	-3.09	0.002	.2541159 .7355433
9	.2882276	.0863427	-4.15	0.000	.1602308 .5184719
10	.2190637	.073024	-4.56	0.000	.1139792 .4210319
11	.1651703	.0647106	-4.60	0.000	.0766385 .3559727
12	.0091631	.0104551	-4.11	0.000	.0009791 .0857558
education4					
1	1.456755	.1490419	3.68	0.000	1.192064 1.780221
2	1.54878	.2073416	3.27	0.001	1.19134 2.013462
3	1.457869	.2329389	2.36	0.018	1.065892 1.993994
4	1.092813	.2020773	0.48	0.631	.7605824 1.570167
5	.8684146	.1800988	-0.68	0.496	.5783598 1.303936
6	.5471565	.1261643	-2.62	0.009	.348209 .8597716
7	.3983053	.1022937	-3.58	0.000	.2407734 .6589064
8	.2216427	.0626413	-5.33	0.000	.1273751 .385676
9	.1645677	.0512526	-5.79	0.000	.0893816 .3029988
10	.1116608	.0379659	-6.45	0.000	.0573433 .2174296
11	.0856576	.0331237	-6.35	0.000	.0401427 .1827785
12	.0908273	.0582404	-3.74	0.000	.0258469 .3191722
13	.0548464	.0509094	-3.13	0.002	.008893 .338258
non_essential4					
1	1.139119	.1308484	1.13	0.257	.9094802 1.42674
2	1.258193	.1638699	1.76	0.078	.9747309 1.624088
3	1.144833	.1587013	0.98	0.329	.8724596 1.502239
4	1.017415	.1440874	0.12	0.903	.7708147 1.342908
5	1.047637	.1545559	0.32	0.752	.7845749 1.398901
6	.8795601	.1447269	-0.78	0.435	.637096 1.2143
7	.8038324	.1438234	-1.22	0.222	.5660656 1.141469
8	.7328725	.1437931	-1.58	0.113	.4989059 1.07656
9	1.067588	.2528525	0.28	0.782	.6711214 1.698268
10	1.020293	.3168747	0.06	0.948	.5550926 1.875359
11	2.773997	1.677676	1.69	0.092	.8478341 9.07614
home4					
1	1.050294	.1224825	0.42	0.674	.8356919 1.320006

2	1.442408	.1781364	2.97	0.003	1.13231	1.837431
3	2.18335	.2813717	6.06	0.000	1.696007	2.810729
4	2.491129	.3259727	6.98	0.000	1.927585	3.219428
5	2.716038	.3724418	7.29	0.000	2.075932	3.553517
6	2.71914	.4031361	6.75	0.000	2.03345	3.636048
7	3.307177	.5339675	7.41	0.000	2.410043	4.538267
8	2.940677	.5413366	5.86	0.000	2.049997	4.218338
9	2.660378	.6168096	4.22	0.000	1.688853	4.190779
10	2.581587	.8114433	3.02	0.003	1.39424	4.78009
11	.8914859	.7986507	-0.13	0.898	.1540115	5.16031
masks4						
1	.655428	.0705672	-3.92	0.000	.5307375	.8094132
2	.5767203	.0670066	-4.74	0.000	.4592697	.7242068
3	.5849548	.0734933	-4.27	0.000	.4572755	.7482843
4	.4436336	.0613972	-5.87	0.000	.3382371	.581872
5	.6561941	.1000257	-2.76	0.006	.4867221	.8846746
6	.6569734	.1262464	-2.19	0.029	.4507937	.9574536
7	.7840156	.2008676	-0.95	0.342	.4745083	1.295405
8	.8409724	.2470652	-0.59	0.556	.472837	1.495726
9	1.362676	.4150369	1.02	0.310	.7501322	2.47541
10	1.381992	.6027965	0.74	0.458	.5877992	3.249243
11	2.888456	2.179965	1.41	0.160	.6580364	12.6789
dayfromfirst	1.172339	.0085949	21.69	0.000	1.155614	1.189306
tests1mpop16apr	1.000052	.0000104	5.06	0.000	1.000032	1.000073
_cons	2.54e-07	5.24e-08	-73.46	0.000	1.69e-07	3.80e-07
ln(popdata2018)	1	(exposure)				
/lnalpha	-.816131	.0384955			-.8915807	-.7406813
country_code						
var(_cons)	.3986612	.1128118			.2289464	.6941832

Note: Estimates are transformed only in the first equation.

Note: \_cons estimates baseline incidence rate (conditional on zero random effects).

LR test vs. nbinomial model: chibar2(01) = 486.54 Prob >= chibar2 = **0.0000**

3 . menbreg cases ib0.mass10 ib0.business10 ib0.education10 ib0.non\_essential10 ib0.home10 ib0.masks10 dayfromfirst te

Fitting fixed-effects model:

Iteration 0: log likelihood = -9994.7123 (not concave)  
 Iteration 1: log likelihood = -9729.7216  
 Iteration 2: log likelihood = -9400.1142  
 Iteration 3: log likelihood = -9314.0225  
 Iteration 4: log likelihood = -9310.3497  
 Iteration 5: log likelihood = -9310.341  
 Iteration 6: log likelihood = -9310.341

Refining starting values:

Grid node 0: log likelihood = -9172.6869

Fitting full model:

Iteration 0: log likelihood = -9172.6869 (not concave)  
 Iteration 1: log likelihood = -9166.0565 (not concave)  
 Iteration 2: log likelihood = -9159.8536  
 Iteration 3: log likelihood = -9116.4445  
 Iteration 4: log likelihood = -9108.8947  
 Iteration 5: log likelihood = -9108.8151  
 Iteration 6: log likelihood = -9108.8147

Mixed-effects nbinomial regression  
 Overdispersion: mean  
 Group variable: country\_code  
 Number of obs = 1,588  
 Number of groups = 30

Obs per group:  
 min = 42  
 avg = 52.9  
 max = 63

Integration method: mvaghermite

Integration pts. = 7

Log likelihood = -9108.8147

Wald chi2(33) = 3378.08  
 Prob > chi2 = 0.0000

cases	IRR	Std. Err.	z	P> z	[95% Conf. Interval]	
mass10						
1	1.344834	.1218836	3.27	0.001	1.125962	1.606253
2	1.142497	.1495364	1.02	0.309	.8839848	1.476607
3	.8847125	.1507808	-0.72	0.472	.6334797	1.235582
4	.6812934	.1465837	-1.78	0.074	.4468831	1.038662
5	.4624673	.1235176	-2.89	0.004	.2739926	.7805906
6	.1330454	.0568681	-4.72	0.000	.0575662	.307491
business10						
1	1.242759	.1301701	2.07	0.038	1.012115	1.525964
2	.9461983	.1383282	-0.38	0.705	.7104637	1.26015
3	.788907	.148699	-1.26	0.208	.5452367	1.141475
4	.5075358	.1206429	-2.85	0.004	.3185174	.8087236
5	.3524099	.1169214	-3.14	0.002	.1839244	.675238
education10						
1	1.480886	.1430543	4.06	0.000	1.225448	1.789567
2	1.165597	.1648493	1.08	0.279	.8834126	1.537917
3	.5976606	.1088434	-2.83	0.005	.4182514	.8540274
4	.3167193	.0719422	-5.06	0.000	.2029204	.4943373
5	.211947	.0606602	-5.42	0.000	.120951	.3714027
non_essential10						
1	1.121458	.1166766	1.10	0.271	.9145844	1.375124
2	1.05786	.1320182	0.45	0.652	.8283241	1.351001
3	.9092745	.1379363	-0.63	0.531	.6754113	1.224113
4	1.167886	.2336179	0.78	0.438	.7890969	1.728504
5	3.964275	1.980284	2.76	0.006	1.489227	10.55277
home10						
1	1.273257	.1249543	2.46	0.014	1.050464	1.543302
2	1.982338	.2200223	6.17	0.000	1.594783	2.464073
3	1.872801	.2487633	4.72	0.000	1.443534	2.429721
4	1.56168	.285629	2.44	0.015	1.09121	2.23499
5	.4451837	.2615056	-1.38	0.168	.1407779	1.40781
masks10						
1	.5949662	.0519296	-5.95	0.000	.5014158	.7059705
2	.4794569	.052917	-6.66	0.000	.3861924	.5952446
3	.4821065	.0823776	-4.27	0.000	.3449039	.6738882
4	.7418492	.1820478	-1.22	0.224	.458599	1.200047
5	1.567337	1.186874	0.59	0.553	.3552924	6.914153
dayfromfirst	1.123644	.0068083	19.24	0.000	1.110379	1.137067
tests1mpop16apr	1.000054	8.33e-06	6.43	0.000	1.000037	1.00007
_cons	3.34e-07	5.66e-08	-87.86	0.000	2.39e-07	4.65e-07
ln(popdata2018)	1	(exposure)				
/lnalpha	-.6825385	.0376508			-.7563328	-.6087442
country_code						
var(_cons)	.2505226	.0693896			.1455732	.4311339

Note: Estimates are transformed only in the first equation.  
 Note: \_cons estimates baseline incidence rate (conditional on zero random effects).  
 LR test vs. nbinomial model: chibar2(01) = 403.05 Prob >= chibar2 = 0.0000

4 . testparm ib0.mass10

```
( 1) [cases]1.mass10 = 0
( 2) [cases]2.mass10 = 0
( 3) [cases]3.mass10 = 0
( 4) [cases]4.mass10 = 0
( 5) [cases]5.mass10 = 0
( 6) [cases]6.mass10 = 0

      chi2( 6) =    64.89
      Prob > chi2 =    0.0000
```

5 . menbreg cases ib0.mass14 ib0.business14 ib0.education14 ib0.non\_essential14 ib0.home14 ib0.masks14 dayfromfirst to  
note: 4.home14 omitted because of collinearity

Fitting fixed-effects model:

```
Iteration 0: log likelihood = -10004.262 (not concave)
Iteration 1: log likelihood = -9743.3976
Iteration 2: log likelihood = -9461.711
Iteration 3: log likelihood = -9345.9307
Iteration 4: log likelihood = -9345.2049
Iteration 5: log likelihood = -9345.2044
```

Refining starting values:

Grid node 0: log likelihood = -9214.5766

Fitting full model:

```
Iteration 0: log likelihood = -9214.5766 (not concave)
Iteration 1: log likelihood = -9207.8849 (not concave)
Iteration 2: log likelihood = -9201.7211
Iteration 3: log likelihood = -9168.2085
Iteration 4: log likelihood = -9162.6838
Iteration 5: log likelihood = -9161.9602
Iteration 6: log likelihood = -9161.9209
Iteration 7: log likelihood = -9161.9208
```

```
Mixed-effects nbinomial regression          Number of obs   =       1,588
Overdispersion:          mean                Number of groups =         30
Group variable:    country_code              Obs per group:
                                                min =         42
                                                avg =        52.9
                                                max =         63

Integration method: mvaghermite              Integration pts. =         7

Log likelihood = -9161.9208                  Wald chi2(24)    =       3035.84
                                                Prob > chi2      =         0.0000
```

cases	IRR	Std. Err.	z	P> z	[95% Conf. Interval]	
mass14						
1	1.546279	.1388527	4.85	0.000	1.296736	1.843843
2	1.44411	.194042	2.73	0.006	1.109752	1.879206
3	1.202014	.2197375	1.01	0.314	.8400459	1.719949
4	.8038274	.1994619	-0.88	0.379	.4942486	1.307315
business14						
1	1.204076	.12594	1.78	0.076	.9808952	1.478036
2	.9057518	.1367136	-0.66	0.512	.6737975	1.217556
3	.6316922	.1282855	-2.26	0.024	.4242692	.9405231
4	.3694655	.15298	-2.40	0.016	.1641071	.8318027
education14						
1	1.565542	.1509728	4.65	0.000	1.295923	1.891255
2	1.037304	.15069	0.25	0.801	.7802814	1.378989
3	.4584379	.0886586	-4.03	0.000	.3138082	.6697253
4	.3896325	.1151281	-3.19	0.001	.218345	.6952919

non_essential14						
1	1.090845	.1123048	0.84	0.398	.8915174	1.334738
2	.9366864	.1212635	-0.51	0.613	.7267709	1.207232
3	.811321	.1463667	-1.16	0.246	.5696806	1.155458
4	1.809959	1.183796	0.91	0.364	.5022778	6.52219
home14						
1	1.478777	.1440292	4.02	0.000	1.221793	1.789812
2	1.825602	.2120113	5.18	0.000	1.453965	2.292229
3	1.762915	.287684	3.47	0.001	1.280344	2.42737
4	1	(omitted)				
masks14						
1	.5638435	.0473464	-6.82	0.000	.4782803	.6647138
2	.4510717	.0561698	-6.39	0.000	.3533861	.5757601
3	.7172957	.1678056	-1.42	0.156	.4534887	1.134567
dayfromfirst	1.098089	.0058299	17.62	0.000	1.086722	1.109576
tests1mpop16apr	1.000056	7.86e-06	7.09	0.000	1.00004	1.000071
_cons	3.78e-07	6.11e-08	-91.54	0.000	2.75e-07	5.19e-07
ln(popdata2018)	1	(exposure)				
/lnalpha	-.6096359	.0371429			-.6824347	-.5368371
country_code						
var(_cons)	.2206711	.0606155			.1288046	.3780588

Note: Estimates are transformed only in the first equation.

Note: \_cons estimates baseline incidence rate (conditional on zero random effects).

LR test vs. nbinomial model:  $\text{chibar2}(01) = 366.57$  Prob >=  $\text{chibar2} = 0.0000$

6 . menbreg deaths ib0.mass ib0.business ib0.education ib0.non\_essential ib0.home ib0.masks dayfromfirst tests1mpop16apr

Fitting fixed-effects model:

Iteration 0: log likelihood = -6092.0787 (not concave)  
 Iteration 1: log likelihood = -5450.907 (not concave)  
 Iteration 2: log likelihood = -5093.0957 (not concave)  
 Iteration 3: log likelihood = -5002.9839  
 Iteration 4: log likelihood = -4827.041 (backed up)  
 Iteration 5: log likelihood = -4626.4817  
 Iteration 6: log likelihood = -4620.6277  
 Iteration 7: log likelihood = -4620.566  
 Iteration 8: log likelihood = -4620.566

Refining starting values:

Grid node 0: log likelihood = -4349.6314

Fitting full model:

Iteration 0: log likelihood = -4349.6314  
 Iteration 1: log likelihood = -4132.8538  
 Iteration 2: log likelihood = -4098.1002  
 Iteration 3: log likelihood = -4096.2717  
 Iteration 4: log likelihood = -4096.2635  
 Iteration 5: log likelihood = -4096.2635

Mixed-effects nbinomial regression  
 Overdispersion: mean  
 Group variable: country\_code  
 Number of obs = 1,588  
 Number of groups = 30  
 Obs per group:  
 min = 42  
 avg = 52.9  
 max = 63

Integration method: mvaghermite Integration pts. = 7



Log likelihood = -4096.2635      Wald chi2(38) = 4664.97  
 Prob > chi2 = 0.0000

deaths	IRR	Std. Err.	z	P> z	[95% Conf. Interval]	
mass						
1	.7563676	.120079	-1.76	0.079	.5541129	1.032446
2	.5832463	.1076164	-2.92	0.003	.406251	.8373551
3	.5948864	.132347	-2.33	0.020	.384648	.9200357
4	.5578361	.1458002	-2.23	0.026	.3342183	.9310714
5	.5042393	.1518256	-2.27	0.023	.2794739	.9097709
6	.4930222	.1732332	-2.01	0.044	.2476167	.9816419
business						
1	1.066555	.1597097	0.43	0.667	.7952827	1.43036
2	1.073205	.19744	0.38	0.701	.7483161	1.539147
3	.7191153	.1582015	-1.50	0.134	.4672387	1.106772
4	.4951598	.130937	-2.66	0.008	.2948892	.8314419
5	.415128	.1298009	-2.81	0.005	.2249216	.7661838
6	.3742309	.1370291	-2.68	0.007	.1825836	.7670388
education						
1	2.509827	.3650106	6.33	0.000	1.887345	3.337615
2	3.140665	.6171348	5.82	0.000	2.136791	4.616163
3	2.756948	.6459132	4.33	0.000	1.741824	4.363679
4	2.019311	.5457603	2.60	0.009	1.188909	3.429714
5	1.096814	.3385769	0.30	0.765	.5989282	2.008591
6	.5517744	.1932168	-1.70	0.089	.277775	1.096049
non_essential						
1	1.396027	.2180175	2.14	0.033	1.027926	1.895946
2	1.405768	.2414743	1.98	0.047	1.003921	1.968466
3	1.418574	.2602883	1.91	0.057	.9900763	2.032523
4	1.440418	.3027939	1.74	0.083	.9540155	2.174812
5	1.044527	.2531821	0.18	0.857	.649528	1.679739
6	.7685437	.2336021	-0.87	0.386	.4235866	1.394424
home						
1	1.304813	.2010532	1.73	0.084	.964695	1.764844
2	2.007811	.3310027	4.23	0.000	1.453436	2.773638
3	2.2306	.3901335	4.59	0.000	1.583237	3.14266
4	1.985532	.3800538	3.58	0.000	1.364417	2.889393
5	1.835626	.4052875	2.75	0.006	1.190824	2.829573
6	1.213501	.3408576	0.69	0.491	.6997581	2.104419
masks						
1	.9091138	.0910322	-0.95	0.341	.7471108	1.106246
2	.8940483	.103534	-0.97	0.333	.7125065	1.121846
3	.9701593	.1395542	-0.21	0.833	.7318137	1.286132
4	1.399045	.3072944	1.53	0.126	.9096386	2.151764
5	1.357023	.4367027	0.95	0.343	.7222082	2.549833
6	1.453493	.6602568	0.82	0.410	.5966917	3.540591
dayfromfirst	1.167611	.0100284	18.04	0.000	1.14812	1.187433
tests1mpop16apr	1.000021	.0000182	1.17	0.243	.9999855	1.000057
_cons	2.98e-09	1.07e-09	-54.76	0.000	1.47e-09	6.01e-09
ln(popdata2018)	1	(exposure)				
/lnalpha	-1.205326	.0668716			-1.336392	-1.07426
country_code						
var(_cons)	1.190973	.3250558			.6975593	2.033401

Note: Estimates are transformed only in the first equation.

Note: \_cons estimates baseline incidence rate (conditional on zero random effects).

LR test vs. nbinomial model: chibar2(01) = 1048.60      Prob >= chibar2 = 0.0000

7 . menbreg deaths ib0.mass4 ib0.business4 ib0.education4 ib0.non\_essential4 ib0.home4 ib0.masks4 dayfromfirst tests1m

Fitting fixed-effects model:

Iteration 0: log likelihood = **-6085.9998** (not concave)  
Iteration 1: log likelihood = **-5456.6922** (not concave)  
Iteration 2: log likelihood = **-5087.2181** (not concave)  
Iteration 3: log likelihood = **-4993.0782** (not concave)  
Iteration 4: log likelihood = **-4947.6529**  
Iteration 5: log likelihood = **-4649.61**  
Iteration 6: log likelihood = **-4596.5313**  
Iteration 7: log likelihood = **-4595.8369**  
Iteration 8: log likelihood = **-4595.8358**  
Iteration 9: log likelihood = **-4595.8358**

Refining starting values:

Grid node 0: log likelihood = **-4329.3104**

Fitting full model:

Iteration 0: log likelihood = **-4329.3104**  
Iteration 1: log likelihood = **-4114.8771**  
Iteration 2: log likelihood = **-4062.0213**  
Iteration 3: log likelihood = **-4048.4743**  
Iteration 4: log likelihood = **-4048.4537**  
Iteration 5: log likelihood = **-4048.4537**

Mixed-effects nbinomial regression  
Overdispersion: mean  
Group variable: country\_code

Number of obs = **1,588**  
Number of groups = **30**  
Obs per group:  
min = **42**  
avg = **52.9**  
max = **63**

Integration method: mvaghermite

Integration pts. = **7**

Log likelihood = **-4048.4537**

Wald chi2(74) = **4914.42**  
Prob > chi2 = **0.0000**

deaths	IRR	Std. Err.	z	P> z	[95% Conf. Interval]	
mass4						
1	.5847977	.1084222	-2.89	0.004	.4066229 .8410455	
2	.5879855	.1128876	-2.77	0.006	.4035934 .8566218	
3	.3668961	.0810156	-4.54	0.000	.2380051 .5655876	
4	.3894459	.0994524	-3.69	0.000	.2360895 .642418	
5	.3189652	.0926378	-3.93	0.000	.1805206 .5635855	
6	.2822302	.0922415	-3.87	0.000	.1487321 .5355529	
7	.2450735	.0896624	-3.84	0.000	.1196401 .5020144	
8	.2146618	.0865236	-3.82	0.000	.097423 .4729859	
9	.1879309	.0837082	-3.75	0.000	.0784974 .4499255	
10	.1686319	.0816859	-3.67	0.000	.0652551 .4357775	
11	.1707839	.0900275	-3.35	0.001	.0607774 .4799008	
12	.1472727	.0855079	-3.30	0.001	.0471962 .4595552	
13	.1240041	.0795963	-3.25	0.001	.0352422 .4363239	
14	.0943584	.071085	-3.13	0.002	.021554 .4130801	
business4						
1	1.097531	.1790413	0.57	0.568	.7971858 1.511034	
2	.9655254	.1806591	-0.19	0.851	.6691071 1.393259	
3	1.130661	.2418595	0.57	0.566	.7434505 1.719542	
4	.8317088	.2021432	-0.76	0.448	.516524 1.339221	
5	.6129191	.1689121	-1.78	0.076	.3571278 1.05192	
6	.5121274	.1606383	-2.13	0.033	.2769382 .9470506	
7	.4135227	.1467851	-2.49	0.013	.2062313 .8291708	
8	.3753815	.1481539	-2.48	0.013	.1731907 .8136194	
9	.3705288	.1619581	-2.27	0.023	.157312 .8727343	
10	.3681416	.1775462	-2.07	0.038	.143054 .9473923	
11	.3111824	.1710421	-2.12	0.034	.105962 .9138603	

12	.1111157	.1535649	-1.59	0.112	.0074027	1.667859
education4						
1	2.056265	.3068929	4.83	0.000	1.534756	2.754984
2	2.623156	.4945699	5.11	0.000	1.812753	3.795856
3	2.668537	.6027761	4.35	0.000	1.713964	4.15475
4	2.425121	.6178535	3.48	0.001	1.471875	3.995728
5	2.349112	.6718227	2.99	0.003	1.341123	4.114707
6	1.985382	.6332169	2.15	0.032	1.062582	3.70959
7	1.367885	.4844265	0.88	0.376	.6832864	2.738397
8	.8678009	.3380697	-0.36	0.716	.4044054	1.862186
9	.609483	.2606812	-1.16	0.247	.2635696	1.409379
10	.3384913	.15844	-2.31	0.021	.1352445	.8471793
11	.2831938	.1458306	-2.45	0.014	.1032185	.7769802
12	.1266484	.0854567	-3.06	0.002	.0337479	.475283
13	.0735043	.062757	-3.06	0.002	.0137901	.3917934
non_essential4						
1	1.378051	.229698	1.92	0.054	.9939952	1.910497
2	1.334065	.2391555	1.61	0.108	.9388226	1.895704
3	1.300853	.2380027	1.44	0.151	.9088523	1.861929
4	1.325037	.2503383	1.49	0.136	.914979	1.918867
5	1.326435	.2586255	1.45	0.147	.9051498	1.943799
6	1.41912	.3067426	1.62	0.105	.9290346	2.167735
7	1.314033	.3094178	1.16	0.246	.8282709	2.084684
8	1.058531	.2738149	0.22	0.826	.6375577	1.757469
9	.9468984	.2855925	-0.18	0.856	.5242927	1.710145
10	.681326	.2583866	-1.01	0.312	.3239997	1.432733
11	1.030781	.6757346	0.05	0.963	.2852047	3.725431
home4						
1	1.301498	.2272646	1.51	0.131	.9242905	1.832645
2	1.547739	.2765779	2.44	0.015	1.090409	2.196877
3	2.382469	.4240976	4.88	0.000	1.680765	3.377129
4	2.368919	.4322144	4.73	0.000	1.65671	3.3873
5	2.575174	.4875535	5.00	0.000	1.776847	3.732185
6	2.409067	.4857608	4.36	0.000	1.622609	3.576712
7	2.317956	.5088787	3.83	0.000	1.50742	3.564314
8	2.19568	.5343065	3.23	0.001	1.362806	3.537563
9	1.783916	.5160842	2.00	0.045	1.011866	3.145037
10	1.516464	.5632429	1.12	0.262	.7322846	3.140394
11	.8600512	.7007593	-0.19	0.853	.1741709	4.24691
masks4						
1	.9292004	.1070975	-0.64	0.524	.7413135	1.164708
2	.9410561	.1158773	-0.49	0.622	.739269	1.197922
3	.9575459	.1279443	-0.32	0.745	.7369274	1.244212
4	1.024722	.1585697	0.16	0.875	.7566373	1.387793
5	1.11171	.1904436	0.62	0.536	.7946477	1.555278
6	1.590809	.3522862	2.10	0.036	1.030667	2.455372
7	1.490327	.4674364	1.27	0.203	.8059439	2.755867
8	1.038825	.4130926	0.10	0.924	.4764953	2.264779
9	3.151694	1.254835	2.88	0.004	1.444247	6.877754
10	2.579486	1.956182	1.25	0.211	.583465	11.40385
11	2.780446	3.416485	0.83	0.405	.2501477	30.90524
dayfromfirst	1.206331	.01368	16.54	0.000	1.179815	1.233444
tests1mpop16apr	1.000018	.0000192	0.92	0.357	.99998	1.000055
_cons	2.61e-09	9.77e-10	-52.74	0.000	1.25e-09	5.43e-09
ln(popdata2018)	1	(exposure)				
/lnalpha	-1.326459	.0685182			-1.460752	-1.192165
country_code						
var(_cons)	1.306631	.3616677			.7595306	2.247814

Note: Estimates are transformed only in the first equation.

Note: \_cons estimates baseline incidence rate (conditional on zero random effects).

LR test vs. nbinomial model:  $\text{chibar2}(01) = 1094.76$  Prob >= chibar2 = 0.0000

8 . menbreg deaths ib0.mass10 ib0.business10 ib0.education10 ib0.non\_essential10 ib0.home10 ib0.masks10 dayfromfirst t

Fitting fixed-effects model:

Iteration 0: log likelihood = **-6096.7024** (not concave)  
 Iteration 1: log likelihood = **-5442.2489** (not concave)  
 Iteration 2: log likelihood = **-5099.6838** (not concave)  
 Iteration 3: log likelihood = **-5013.7263**  
 Iteration 4: log likelihood = **-4717.0454** (backed up)  
 Iteration 5: log likelihood = **-4649.6372**  
 Iteration 6: log likelihood = **-4648.6644**  
 Iteration 7: log likelihood = **-4648.66**  
 Iteration 8: log likelihood = **-4648.66**

Refining starting values:

Grid node 0: log likelihood = **-4387.6526**

Fitting full model:

Iteration 0: log likelihood = **-4387.6526**  
 Iteration 1: log likelihood = **-4188.023**  
 Iteration 2: log likelihood = **-4148.1391**  
 Iteration 3: log likelihood = **-4146.1946**  
 Iteration 4: log likelihood = **-4146.1839**  
 Iteration 5: log likelihood = **-4146.1839**

Mixed-effects nbinomial regression  
 Overdispersion: mean  
 Group variable: **country\_code**  
 Number of obs = **1,588**  
 Number of groups = **30**  
 Obs per group:  
     min = **42**  
     avg = **52.9**  
     max = **63**  
 Integration method: **mvaghermite**  
 Integration pts. = **7**  
 Wald chi2(33) = **4358.84**  
 Prob > chi2 = **0.0000**  
 Log likelihood = **-4146.1839**

deaths	IRR	Std. Err.	z	P> z	[95% Conf. Interval]	
mass10						
1	.7933842	.1263504	-1.45	0.146	.5806648	1.084031
2	.6462756	.1314059	-2.15	0.032	.4338551	.9626996
3	.6323483	.1586684	-1.83	0.068	.3866993	1.034045
4	.4859425	.1477044	-2.37	0.018	.2678295	.8816809
5	.3990637	.1444065	-2.54	0.011	.1963468	.8110744
6	.1616749	.0796189	-3.70	0.000	.0615826	.4244512
business10						
1	1.354599	.202174	2.03	0.042	1.011041	1.8149
2	1.20092	.2335491	0.94	0.346	.8203075	1.758133
3	.8332331	.2080098	-0.73	0.465	.5108237	1.359133
4	.7346226	.2298941	-0.99	0.324	.397821	1.356566
5	.6529924	.2649127	-1.05	0.293	.2948383	1.446213
education10						
1	2.618531	.3908756	6.45	0.000	1.954323	3.508479
2	3.272596	.651684	5.95	0.000	2.21508	4.834986
3	2.401223	.5825608	3.61	0.000	1.492528	3.863158
4	1.280701	.372342	0.85	0.395	.7243953	2.264228
5	.8868284	.3091561	-0.34	0.730	.4478198	1.756207
non_essential10						
1	1.304915	.2045242	1.70	0.090	.9597769	1.774166
2	1.37905	.2423596	1.83	0.067	.9772074	1.946138
3	1.455649	.3032486	1.80	0.072	.9676759	2.189695
4	1.271133	.324759	0.94	0.348	.7704057	2.09731
5	1.779882	.9400789	1.09	0.275	.6321393	5.011522

home10						
1	1.487145	.2242743	2.63	0.009	1.106585	1.998581
2	1.876879	.313318	3.77	0.000	1.353133	2.603348
3	1.403147	.2671175	1.78	0.075	.9661836	2.03773
4	.8840621	.208023	-0.52	0.600	.5574321	1.402083
5	.1816888	.1060076	-2.92	0.003	.0579013	.570122
masks10						
1	.8454197	.0820308	-1.73	0.084	.6990054	1.022502
2	.8645704	.1104758	-1.14	0.255	.6730275	1.110626
3	1.043792	.2166648	0.21	0.836	.6949062	1.56784
4	1.230409	.4119592	0.62	0.536	.6383439	2.371616
5	1.272183	1.557947	0.20	0.844	.1153845	14.02659
dayfromfirst	1.147815	.0090551	17.48	0.000	1.130204	1.165701
tests1mpop16apr	1.000016	.000019	0.82	0.410	.9999784	1.000053
_cons	3.40e-09	1.26e-09	-52.51	0.000	1.64e-09	7.04e-09
ln(popdata2018)	1	(exposure)				
/lnalpha	-1.085879	.065591			-1.214435	-.957323
country_code						
var(_cons)	1.294418	.3538466			.7575077	2.211882

Note: Estimates are transformed only in the first equation.

Note: \_cons estimates baseline incidence rate (conditional on zero random effects).

LR test vs. nbinomial model: chibar2(01) = 1004.95 Prob >= chibar2 = **0.0000**

9 . menbreg deaths ib0.mass14 ib0.business14 ib0.education14 ib0.non\_essential14 ib0.home14 ib0.masks14 dayfromfirst  
note: 4.home14 omitted because of collinearity

Fitting fixed-effects model:

Iteration 0: log likelihood = **-6105.2707** (not concave)  
 Iteration 1: log likelihood = **-5341.976** (not concave)  
 Iteration 2: log likelihood = **-5109.2826** (not concave)  
 Iteration 3: log likelihood = **-5046.7401** (not concave)  
 Iteration 4: log likelihood = **-4944.9043**  
 Iteration 5: log likelihood = **-4764.9866**  
 Iteration 6: log likelihood = **-4664.3758**  
 Iteration 7: log likelihood = **-4663.255**  
 Iteration 8: log likelihood = **-4663.2537**  
 Iteration 9: log likelihood = **-4663.2537**

Refining starting values:

Grid node 0: log likelihood = **-4394.9043**

Fitting full model:

Iteration 0: log likelihood = **-4394.9043**  
 Iteration 1: log likelihood = **-4204.3637**  
 Iteration 2: log likelihood = **-4169.7704**  
 Iteration 3: log likelihood = **-4169.0356**  
 Iteration 4: log likelihood = **-4169.0308**  
 Iteration 5: log likelihood = **-4169.0308**

Mixed-effects nbinomial regression

Overdispersion: **mean**

Group variable: **country\_code**

Number of obs = **1,588**

Number of groups = **30**

Obs per group:

min = **42**

avg = **52.9**

max = **63**

Integration method: **mvaghermite**

Integration pts. = **7**

Log likelihood = **-4169.0308** Wald chi2(24) = **4122.31**  
 Prob > chi2 = **0.0000**

deaths	IRR	Std. Err.	z	P> z	[95% Conf. Interval]	
<b>mass14</b>						
1	.8952258	.1423886	-0.70	0.487	.6554601	1.222697
2	.9902703	.20305	-0.05	0.962	.6625507	1.480091
3	.8994396	.2289916	-0.42	0.677	.5460866	1.481435
4	.7018597	.2232273	-1.11	0.266	.3762917	1.309109
<b>business14</b>						
1	1.323085	.1970869	1.88	0.060	.9880811	1.77167
2	.8965988	.1750896	-0.56	0.576	.6114681	1.314687
3	.6933552	.1757276	-1.44	0.148	.4219128	1.139433
4	1.083379	.4883515	0.18	0.859	.447803	2.621042
<b>education14</b>						
1	2.777188	.4169053	6.80	0.000	2.069307	3.727227
2	2.789269	.5449698	5.25	0.000	1.901874	4.090712
3	1.343229	.324296	1.22	0.222	.8368432	2.156034
4	.5776422	.1929843	-1.64	0.100	.3001105	1.111826
<b>non_essential14</b>						
1	1.375352	.2158628	2.03	0.042	1.011153	1.870729
2	1.356355	.2466374	1.68	0.094	.949713	1.937111
3	.9347891	.2141295	-0.29	0.768	.596666	1.464522
4	.3028598	.1800789	-2.01	0.045	.0944325	.9713188
<b>home14</b>						
1	1.658136	.249702	3.36	0.001	1.234344	2.227431
2	1.784602	.3096292	3.34	0.001	1.270157	2.507409
3	1.438098	.3076567	1.70	0.089	.9455587	2.187201
4	1 (omitted)					
<b>masks14</b>						
1	.7787291	.0730946	-2.66	0.008	.647872	.9360168
2	.7451864	.1077411	-2.03	0.042	.561301	.9893136
3	.6632024	.2066264	-1.32	0.187	.3601197	1.221364
dayfromfirst	1.134572	.0075746	18.91	0.000	1.119823	1.149516
tests1mpop16apr	1.00002	.0000182	1.11	0.265	.9999846	1.000056
_cons	3.38e-09	1.21e-09	-54.49	0.000	1.67e-09	6.81e-09
ln(popdata2018)	1 (exposure)					
/lnalpha	-.9963278	.0634911			-1.120768	-.8718875
<b>country_code</b>						
var(_cons)	1.193087	.324013			.7006589	2.031596

Note: Estimates are transformed only in the first equation.  
 Note: \_cons estimates baseline incidence rate (conditional on zero random effects).  
 LR test vs. nbinomial model: chibar2(01) = 988.45 Prob >= chibar2 = **0.0000**

```
10 . log close
    name: <unnamed>
    log: C:\Users\Pau1H\Dropbox\001Covid19 files\Sensitivity analyses with varying time gaps.smcl
    log type: smcl
    closed on: 28 Feb 2021, 18:07:26
```