

Estimation of all-cause excess mortality by age-specific mortality patterns of COVID-19 pandemic in Peru in 2020

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Summary

Background

All-cause excess mortality is a comprehensive approach for estimating the direct and indirect effects of COVID-19 on mortality. Estimates are usually derived from Civil Registration and Vital Statistics (CRVS) systems, but these do not include non-registered deaths, which may be affected by changes in mortality registration coverage over time.

Methods

We use quasi-Poisson models to estimate excess registered mortality in Peru during the first wave of the COVID-19 pandemic during 2020. We use logistic mixed-effects models to estimate the completeness of the new online registration system (SINADEF) at this time.

Findings

We estimate national under-registration of mortality to be 28.4% (95% CI 27.8% - 28.3%). We estimate total all-cause excess were registered during the period of analysis to be 153,832 (95% CI 129,070 - 161,454) of which 110,113 (95% CI 93,248 - 115,769) are excess registered deaths. Mortality in people aged 60 years and over accounts for 75.9% (95% CI 75.2% - 85.9%) of total excess deaths for both models, while there were fewer deaths than expected in younger age groups. Lima region, on the Pacific coast and including the national capital, accounts for 78,366 (95% CI 77,116 - 79,472) excess deaths, while the regions of Apurimac and Pasco, in the Andes, account for less than 400 deaths.

Interpretation

Estimating excess mortality in low- and middle-income countries (LMICs) like Peru must take under-registration of mortality into account. Combining demographic trends with data from administrative registries reduces uncertainty and measurement errors. In countries like Peru, this is likely to produce significantly higher estimates of excess mortality than studies that do not take these effects into account.

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Research in context

Evidence before this study

We searched PubMed, Google Scholar, medRxiv, and SocArXiv for studies published up to December 08, 2020, using the key words “excess mortality” and “under-registration” or “subregistration,” combined with “coronavirus” or “SARS-CoV-2” or “COVID-19.” We found studies estimating cumulative mortality in high-income countries in Europe and North America solely based on official death counts. We found few studies computing overall COVID-19 mortality in LMICs. Prior research shows a significant percentage of under-registration of deaths in LMICs.

Added value of this study

To our knowledge, we provide the first estimate of excess mortality associated with COVID-19 in a LMIC accounting for both registered and unregistered deaths. We develop an analytical strategy to address common challenges faced by LMICs, such as low completion rates of death certificates, missing data, inconsistency and variability of data across regions and age groups.

Implications of all the available evidence

Our approach shows the importance of accounting for unregistered deaths based on demographic trends to generate robust estimates of excess mortality associated with COVID-19. It suggests that previous reports of COVID-19 related mortality in Peru were substantial underestimates.

1 Introduction

Monitoring mortality is an essential part of the public health response to the COVID-19 pandemic. In many countries, robust COVID-19 mortality measurement and estimation have been hindered by shortcomings in diagnosis and attribution of cause of death, as well as incomplete registration of general mortality.¹ Disentangling the contribution of COVID-19 to overall mortality is especially challenging, as many people who die from COVID-19 have other conditions, such as cardiovascular disease and diabetes.² A further complication is the decision of some countries to apply an arbitrary maximum cut-off time between a positive COVID-19 test and death when deciding whether to attribute mortality to this cause.³ Also, it is widely documented that the pandemic has led to large numbers of deaths not directly attributable to COVID-19 (either exclusively or in part), due to the wider impacts of responses, including reduced access to treatment for other conditions.^{4,5} Conversely, there is evidence that the pandemic has, to a limited extent, reduced expected rates of mortality from causes such as road traffic injuries and homicides during lockdowns.⁶

In the absence of good data on these different mortality effects, robust estimation of all-cause excess mortality offers the most complete and reliable approach for gauging the overall impact of the pandemic on mortality in a defined population over a fixed period of time.⁷ Excess mortality refers to the number of additional deaths occurring over a time period when specific conditions apply (in this case, the presence of COVID-19), compared to the number of deaths we might reasonably expect over the same period based on historical data. It captures deaths directly and indirectly attributed to COVID-19 and corresponding responses, to provide an estimate of the overall mortality effect of the pandemic as it unfolds.⁸

Excess mortality estimates have been computed for high-income countries based on registered deaths by surveillance agencies^{9,10}, academia^{11–14} and news agencies.^{15–18} However, these types of data are often incomplete or inaccurate, especially in many low and middle-income countries (LMICs). The Global Burden Disease (GBD) project estimates that only 64% of global deaths were registered in 2015.¹⁹ In most LMICs, responsibility for mortality data is often divided between different national and subnational agencies.²⁰ This

can cause extended delays in national reporting and discrepancies between different sources.^{12,21} Disaggregation of summary data by different geographical areas or demographic groups is usually very limited.²² These shortcomings in mortality data explain a lack of published studies of excess mortality in LMICs.^{23,24}

We analyse excess mortality associated with COVID-19 pandemic in Peru during 2020. This country is well-suited to our analytical approach. Like much of Latin America, Peru has experienced high levels of COVID-19 mortality. Official sources report 37,723 deaths directly caused by COVID-19 between 18 March and December 31 2020. However, these only include cases with positive COVID-19 test results and rates of testing in Peru have been low compared to other Latin American countries.²⁵ Also, many tests have used low sensitivity devices, potentially generating false negatives.²⁶ Anonymised individual level data for mortality and COVID-19 are in the public domain in Peru. Also, the progressive implementation of electronic registration of deaths in recent years enables comparisons of spatial and temporal trends in death registration.²⁷ Other studies show unregistered mortality tends to be more prevalent among older people and in poorer regions.^{28,29} However, as in many LMICs, Peru does not report sub-national data disaggregated by age.

Potentially, our analytical design could be applied to other countries where mortality registration and cause of death data are very incomplete. It may offer a valuable contribution towards assessing the true global impact of the COVID-19 pandemic.

2 Data and methods

2.1 Data

We combine several demographic data sources to forecast mortality rates. These sources are (i) population projections from Peru's National Institute of Statistics and Information (INEI, for its acronym in Spanish) for 2020 disaggregated by region and age group (the most recent census was in 2017); (ii) crude mortality rate estimates disaggregated by region and year from 2015 to 2020 from INEI; (iii) individual-level registered deaths by region and age since 2017 from the Sistema Informático Nacional de Defunciones (SINADEF); and (iv) individual-level registered COVID-19 deaths by region and age in 2020 from the Ministry of Health (MoH).

2.2 Excess mortality methods

Our approach to estimate excess mortality decomposes the estimate of total excess deaths into three terms, namely 1) excess registered deaths, 2) excess unregistered deaths, 3) unregistered COVID-19 deaths. Figure 1 summarises data sources used (squares), analysis performed (diamonds) and different outputs (circles).

To estimate the first term, excess registered deaths, we fit quasi-Poisson regressions to weekly deaths since 2017 by age-groups and region, as follows:

$$\log(\text{Deaths}_i) = \beta_0 + \beta_1 \text{COVID-19} + \beta_2 \text{Year} + \beta_3 \epsilon_{t-1} + \sum \beta_k \phi_k(\text{Week}) + \log(\text{Pop}) + \epsilon_t \quad (1)$$

where we fit a natural cubic B-spline function $\phi_k(\text{Week})$ on weeks to address long-term trends and seasonality³⁰ and Year to deal with registration growth. Additionally, we use lagged residuals ϵ_{t-1} and the log of population in 2020 as an offset. Finally, we compute a dichotomous variable starting in the week corresponding to March 16 2020 as the first registered case of COVID-19 in the country.

Secondly, we compute the population attributable fraction³¹ based on the relative risk RR from parameter β_1 in equation (1), as follows:

$$\widehat{\text{Excess}_{\text{Reg}}} = (\text{RR}^{\circ} 1) / \text{RR} * n \quad (2)$$

where n is the weekly number of deaths,³² computed cumulatively over time.

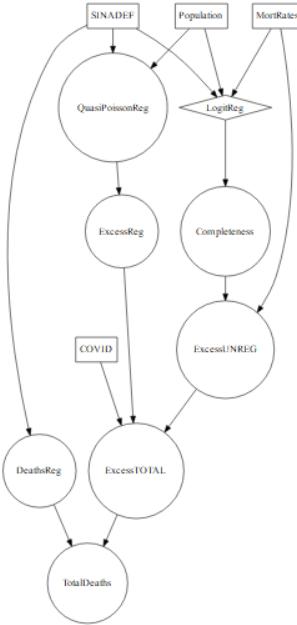


Figure 1: Flowchart: Data, analysis and outputs

To estimate our second term, we predict the logit of death registration completeness for years 2017 and 2019 (years for which data are available).²⁹ We exploit variability in mortality rates, population over 60 and rurality to address potential differences in terms of registration completeness. We model random-effects regressions by region using the equation:

$$\text{logit}(\text{Reg complete}_{jk}) = \beta_0 + \beta_1 * \text{RegCDR} + \beta_2 * \text{RegCDR}^2 + \beta_3 * \text{complete} < 5 + \beta_4 * +60 + \beta_5 * \log(5q0) + \beta_6 * \text{LPG} + \beta_7 * k + \epsilon_{jk} + \gamma_j \quad (3)$$

where RegCDR and RegCDR² are the Crude Death Rates based on the Registration System, complete < 5 is the completeness registration rate for infants, $\log(5q0)$ is the logarithm under-five mortality rate and +60 represents the fraction of the population over 60. LPG is the share of households that use liquefied petroleum gas (LPG) for cooking, which is an important proxy of rurality in Peru, where 81.8% of rural households primarily use solid fuel compared to 9.8% of urban households, who typically cook with LPG.³³ Rurality is an important factor to explain delays and underregistration of deaths as the system requires access to internet and computers. Additionally, ϵ_{jk} in the error term and γ_j is the region-level random effect.

Then, we compute the completeness of deaths registration, $\widehat{\text{Completeness}}$ using the inverse logit of the predicted values from equation (3) and $\widehat{\text{Excess}_{\text{Reg}}}$ from equation (2), as follows:

$$\widehat{\mu}_{\text{Deaths}_{\text{Not reg}}} = (1 - \widehat{\text{Completeness}}) * \sum \widehat{\text{Excess}_{\text{Reg}}} \quad (4)$$

The third term, unregistered COVID-19 deaths, is computed to correct for situations where the proportion of cumulative cases of COVID-19 exceeds registered deaths for that period. It follows a deterministic approach conditional on excess registered deaths being lower than officially registered COVID-19 deaths for each region and specific age-group as follows:

$$\widehat{\mu}_{\text{Deaths}_{\text{COVID not reg}}} = \begin{cases} \widehat{\mu}_{\text{Deaths}_{\text{COVID Reg}}} - \widehat{\mu}_{\text{Excess}_{\text{Reg}}} & \text{if } \widehat{\mu}_{\text{Deaths}_{\text{COVID Reg}}} > \widehat{\mu}_{\text{Excess}_{\text{Reg}}}, \\ 0 & \text{if } \widehat{\mu}_{\text{Deaths}_{\text{COVID Reg}}} < \widehat{\mu}_{\text{Excess}_{\text{Reg}}}. \end{cases} \quad (5)$$

Finally, we estimate total excess deaths for different scenarios. A first scenario is when there is no solid evidence suggesting under-registration of deaths for some regions or age groups and, therefore, no scope to expand registration over time. This usually occurs in areas with very small populations. In those cases $\widehat{\mu}_{\text{Deaths}_{\text{Not reg}}}$ is set to 0 as, if not, they would add negative values to the sum. A second scenario is when some groups have fewer deaths than expected, due to effects such as reduced road traffic injuries caused by lockdowns. In these groups $\widehat{\text{Excess}_{\text{Reg}}} \leq 0$ is taken account of in our final estimates. A third scenario is related to younger age groups in some regions that have not been significantly affected by COVID-19 mortality. This case corresponds to when β_1 in equation (1) is not statistically significant and therefore we set $\widehat{\text{Excess}_{\text{Reg}}} = 0$.

Equation (6) summarises the estimation of $\widehat{\text{Excess}_{T_{\min/\text{mean}/\max}}}$ as follows:

$$\widehat{\text{Excess}_{T_{\min/\text{mean}/\max}}} = \widehat{\text{Excess}_{\text{Reg min/mean/max}}} + \widehat{\mu}_{\text{Deaths}_{\text{Not reg}}} + \text{Deaths}_{\text{COVID Not Reg}} \quad (6)$$

Finally, we estimate total mortality during 2020 by adding $\widehat{\text{Excess}_{T_{\min/\text{mean}/\max}}}$ from equation (6) and the counterfactual difference of SINADef deaths during 2020 and $\widehat{\text{Excess}_{\text{Reg min/mean/max}}}$, adjusted by Completeness as follows:

$$\text{Total}_{\widehat{\text{Mortality}} \text{ 2020}} = \widehat{\text{Excess}_{T_{\min/\text{mean}/\max}}} + \left(\left(\sum_{\text{week}=1}^{52} \text{SINADEF}_{2020} - \widehat{\text{Excess}_{\text{Reg min/mean/max}}} \right) * (1 + \widehat{\text{Completeness}}) \right) \quad (7)$$

3 Results

Estimates of registration completeness derived from our logistic regression model fit the data according to R² and Root Mean Square of Errors parameters. Model fit and goodness-of-fit are presented in Appendix 1. Figure 2 shows important variations in regional completeness rates: Amazonas and Loreto (in Amazonia), Lambayeque (on the coast), and Cajamarca and Pasco (in the Andes) show estimated completion of below 40%, while Ica (coast) and Madre De Dios (Peru's least populated regions in Amazonia) appear to have full registration completeness (see Appendix 1).

Table 1 summarises our estimates of excess mortality. The quasi-Poisson models show a good fit for our first term (weekly excess registered deaths) -across models. Excess registered mortality is estimated to be 110,113 (95% CI 93,248 - 115,769), of which 37,724 are reported as COVID-19 deaths. This represents an increase of 192 % (95% CI 147 % - 207 %) compared to MoH data.

Table 1: Summary of estimations, Peru, 2020.

Total Excess REGISTERED mortality based on Poisson models by region and age group	110113 (95% CI 93248 - 115769)
Completeness of CRVS deaths registration	71.6 % (95% CI 60.6 % - 75.3 %)
Excess TOTAL mortality	153832 (95% CI 129070 - 161454)
Counterfactual estimated deaths in 2020	101556 (95% CI 95900 - 118421)
Total estimated deaths in 2020	289949 (95% CI 257565 - 322334)

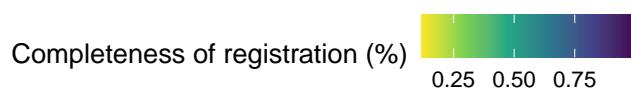


Figure 2: Completeness of registration

Table 2 shows estimates by region. Lima, which includes the capital, accounts for 78366 (95% CI 77116 - 79472) total excess deaths, and Amazonas, Apurimac, and Pasco show the lowest numbers. Table 3 shows excess mortality estimates by age group. Deaths among people aged 60 years and over accounts for 75.9% of total excess mortality. There was negative excess registered mortality for the two youngest age groups: under 9 and 10 to 19.

Table 2: Estimated total excess deaths by region

Region	Total excess (TE)	TE - Lower CI 95%	UpperCI 95%	Excess registered (ER)	ER - Lower CI 95%	ER - Upper CI 95%	Excess Covid-19
AMAZONAS	374.2	242.7	455.6	196.2	119.8	243.5	36.73
ANCASH	4594	4252	4875	3938	3641	4182	69.78
APURIMAC	-30.17	1653	91.54	-83.75	1094	1.362	89.59
AREQUIPA	6011	3784	6500	5513	3396	5977	210.7
AYACUCHO	419.1	2201	658.2	103.6	1296	263.6	264.3
CAJAMARCA	2683	-1933	3468	1429	-1262	1887	230.8
CALLAO	7838	7617	8011	6356	6175	6497	52.95
CUSCO	1331	875.4	1662	992.9	610.7	1270	147.3
HUANCAVELICA	15.9	550.4	823.9	565.6	430.2	653.9	24.25
HUANUCO	1592	1078	1959	1014	679.3	1254	36.34
ICA	4197	3918	4419	4097	3820	4320	93.33
JUNIN	3558	2975	4010	2876	2381	3260	171.7
LA	7517	6982	7947	6169	5722	6529	139.1
LIBERTAD							
LAMBAYEQUE	8101	8046	8149	4292	4262	4317	26.55
LIMA	78366	77116	79472	55172	54291	55951	63.73
LORETO	4654	3107	5030	2636	1714	2860	228.8
MADRE	403.9	343.8	443.2	348.9	289.3	387.9	52.1
DE DIOS							
MOQUEGUA	976.2	841.4	1068	803.4	689	881.1	29.25
PASCO	173.2	-11404	288.7	12.19	-7250	84.64	153.7
PIURA	11512	11263	11727	7547	7382	7690	151.1
PUNO	2541	710	3330	1677	347.7	2250	231.1
SAN	2447	1502	2867	1549	912.3	1832	147.2
MARTIN							
TACNA	599.2	364.3	741.5	466.9	279.8	580.2	12.9
TUMBES	1039	940.2	1111	883.9	797.8	946.4	22.98
UCAYALI	2219	2045	2346	1557	1431	1648	66.91
Total	153832	129070	161454	110113	93248	115769	2753

Table 3: Estimated total excess deaths by age-group

Age range	Total excess (TE)	TE - Lower CI 95%	UpperCI 95%	Excess registered (ER)	ER - Lower CI 95%	ER - Upper CI 95%	Excess Covid-19
< 9	323.9	-3495	1578	-795	-3087	51.87	1525
10-19	154.9	-1123	216.3	-54.61	-1269	3.748	212.4
20-29	1024	802.2	1180	290.4	134.4	400.7	611.4
30-39	3129	2677	3438	2083	1749	2311	174.2
40-49	10203	-1921	10712	7284	-403.1	7663	131.5
50-59	22196	21296	22844	16124	15461	16605	9.589
60-69	37244	35975	38196	27062	26110	27779	65.91

Age range	Total excess (TE)	TE -		Excess registered (ER)	ER - Lower CI 95%	ER - Upper CI 95%	Excess Covid-19
		TE - Lower CI 95%	UpperCI 95%				
70-79	37972	36402	39203	27612	26423	28545	19.23
> 80	41586	38456	44086	30506	28129	32410	3.594
Total	153832	129070	161454	110113	93248	115769	2753

Our second term, excess unregistered deaths, is estimated to be 40,966 deaths (95% CI 33,069 - 42,932). This represents 28.4 % (95% CI 25 % - 39.4 %) additional deaths compared to estimates based only on registered deaths. Coastal regions show the highest values where Lima accounts for 23,194 additional deaths, followed by Piura (3,809) and Lambayeque (3,965). Our third term adds 2,753 deaths corresponding to cases when reported COVID-19 deaths exceed our estimate of adjusted excess registered mortality. This mainly occurs for some smaller groups at younger ages. The regions with the highest under reported COVID-19 values are the Andean regions of Ayacucho (264), Cajamarca and Puno (both with 231).

Combining all these terms, our estimate of total excess deaths during 2020 is 153,832 (95% CI 129,070 - 161,454) and our estimate of total deaths for 2020 is 289,949 (95% CI 257,565 - 322,334). This is 51.5 % (95% CI 29.9 % - 60.4 %) more than deaths projected by INEI for 2020.

Figure 3 shows age-standardised rates per 1,000 people by direct standardization methods,³⁴ using INEI estimates as the standard population. We find that Moquegua, Callao, and Lima present the highest rates ranging from 9.65 to 11.4, while regions of Amazonas, Ayacucho, and Cusco show the lowest rates ranging from 4.59 to 6.86. We find that Amazonas shows -1.59 deaths per 1000 in comparison to INEI's projections for 2020, which suggests INEI's overestimations of mortality rates for Amazonas.

4 Discussion

The COVID-19 pandemic has underscored the need for methodological advances in population health measurement, research, critical data quality assessment and improvement of health information systems to obtain reliable information on the impact of the pandemic from multiple data sources linked to COVID-19. In this regards, our study provides a framework, tools and analytical strategy to estimate the excess mortality caused by the COVID-19 pandemic in Peru and, potentially, other countries where the quality of mortality data is medium to lower.³⁵ The scope for applying a similar approach will vary between countries. In countries such as India, where mortality and cause of death data are of very low quality considerably lower than in Peru, achieving robust estimates by any method will be hugely challenging.³⁶ In other countries, such as Brazil, data quality is similar to Peru's and a recent reform to its CRVS system is likely to have boosted registration rates where they are relatively complete.^{37,38}

Other published studies of excess COVID-19 mortality fall into two categories. Some studies provide estimates for countries where mortality data are relatively complete and reliable.³⁹⁻⁴² As such, they do not apply specific methods to address data gaps. Studies of excess COVID-19 mortality for countries with less complete data do not take unregistered deaths.⁴³⁻⁴⁵

We estimate that overall excess mortality in Peru was 308 % (95% CI 242 % - 328 %) higher than the level officially attributed to COVID-19. Registered excess mortality was 192 % (95% CI 147 % - 207 %) higher. This is a larger differential than reported by studies for high-income countries. For example, separate studies of the USA show differentials of between 28 and 33%.^{11,12} Our estimates are higher than those previously published for Peru, which are only based on registered deaths, apply simpler methods.^{46,47} Our estimates of regional variations for excess mortality are in line with,⁴⁶ who speculate they may be attributable to the effect of altitude of COVID-19 case fatality.

Despite making up a relatively low share of Peru's population (12.5% in 2020), people aged 60 or more accounted for 75.9% of excess mortality. No other published study for Peru provides data disaggregated by age groups. A study of registered excess mortality in six Brazilian cities reports that people aged 60 and over accounted for 71.1% of the total.⁴⁴ An analysis of European countries reports that 91% of excess COVID-19

Age-standardised mortality rate

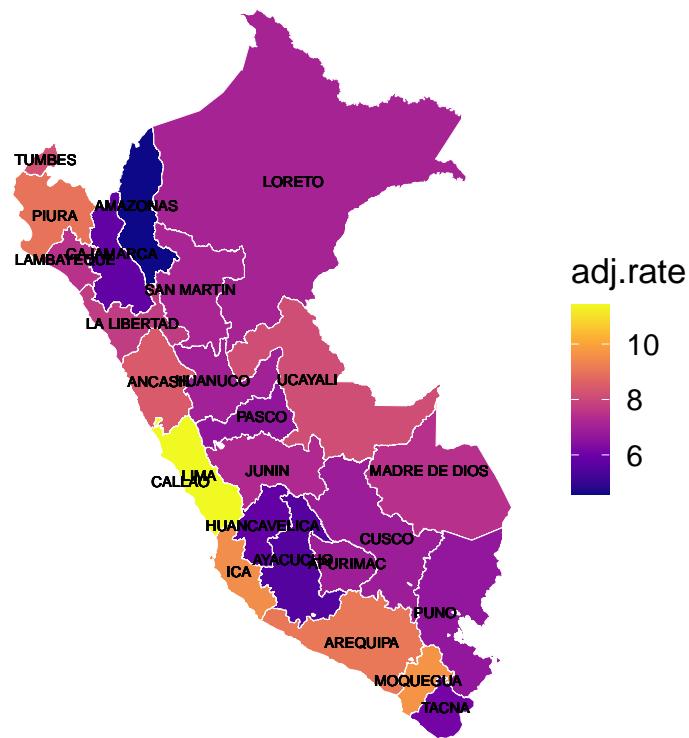


Figure 3: Mortality rates

deaths occurred among people aged 65 or more.⁴¹ This reflects a higher proportion of population aged 60 or more in Europe (25.5% in 2020) than in Peru.

5 Limitations

Our analysis assumes that comparison between years is not invalidated by specific time-bound mortality events such as additional disease outbreaks or other major external shocks. Our estimates of registration completeness assume no variation across age groups, which may not be the case. We do not take into account changes in registration over time as our estimates are based on provisional data, which are incomplete and there are still fatalities due to COVID-19. Finally, we present a conservative scenario, allowing for the existence of negative as well as positive excess deaths.

6 Conclusion

There is an evident need for robust estimates of the direct and indirect mortality effects of the COVID-19 pandemic. To date, much of the data for LMICs in the public domain rely on officially-registered deaths. Inaccurate attribution of cause of death can, to some degree, be resolved by generating excess mortality estimates based on temporal comparisons of all-cause mortality. Also, these approaches do not include deaths that are not officially registered. This paper develops and applies a method to obtain robust estimates of excess mortality for both registered and unregistered deaths.

Our study indicates that official data under-represent the overall mortality impact of the COVID-19 pandemic in Peru. This gap is considerably greater than those reported for high-income countries. It is plausible that under-estimation of excess mortality in other countries with low quality mortality data will be comparable to Peru. In that case, LMICs would account for a much larger share of the global distribution of excess mortality associated with the COVID-19 pandemic than indicated by official data sources.

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8 Appendices

8.1 Appendix 1: Model fit registration completeness

```
##          R2m   R2c
## [1,] 0.743 0.87
## [1] 0.0527
## [1] 0.239
```

8.1.1 Quasi-Poisson models

Table 4: Model fit Quasi-Poisson (continued below)

Departamento	range	null.deviance	df.null	logLik	AIC	BIC
AMAZONAS	a0.9	70.49	136	NA	NA	NA
AMAZONAS	a10.19	24.47	74	NA	NA	NA
AMAZONAS	a20.29	22.62	106	NA	NA	NA
AMAZONAS	a30.39	41.52	119	NA	NA	NA
AMAZONAS	a40.49	77.28	159	NA	NA	NA
AMAZONAS	a50.59	125.9	172	NA	NA	NA
AMAZONAS	a60.69	263.5	179	NA	NA	NA
AMAZONAS	a70.79	243.1	200	NA	NA	NA
AMAZONAS	a80	279.1	209	NA	NA	NA
ANCASH	a0.9	234.9	209	NA	NA	NA
ANCASH	a10.19	109	183	NA	NA	NA
ANCASH	a20.29	171.9	200	NA	NA	NA
ANCASH	a30.39	196.3	201	NA	NA	NA
ANCASH	a40.49	353.7	207	NA	NA	NA
ANCASH	a50.59	791.8	209	NA	NA	NA
ANCASH	a60.69	1186	211	NA	NA	NA
ANCASH	a70.79	1201	211	NA	NA	NA
ANCASH	a80	934.5	211	NA	NA	NA
APURIMAC	a0.9	105.1	169	NA	NA	NA
APURIMAC	a10.19	21.71	104	NA	NA	NA
APURIMAC	a20.29	41.26	135	NA	NA	NA
APURIMAC	a30.39	68.23	148	NA	NA	NA
APURIMAC	a40.49	120.5	162	NA	NA	NA
APURIMAC	a50.59	182	194	NA	NA	NA
APURIMAC	a60.69	284.2	202	NA	NA	NA
APURIMAC	a70.79	275.9	210	NA	NA	NA
APURIMAC	a80	398	210	NA	NA	NA
AREQUIPA	a0.9	239.6	207	NA	NA	NA
AREQUIPA	a10.19	104.9	172	NA	NA	NA
AREQUIPA	a20.29	188.4	201	NA	NA	NA
AREQUIPA	a30.39	239.2	209	NA	NA	NA
AREQUIPA	a40.49	614.1	207	NA	NA	NA
AREQUIPA	a50.59	1524	210	NA	NA	NA
AREQUIPA	a60.69	2509	211	NA	NA	NA
AREQUIPA	a70.79	2570	211	NA	NA	NA
AREQUIPA	a80	2815	211	NA	NA	NA
AYACUCHO	a0.9	157.4	173	NA	NA	NA
AYACUCHO	a10.19	52.87	122	NA	NA	NA
AYACUCHO	a20.29	104.5	167	NA	NA	NA

Departamento	range	null.deviance	df.null	logLik	AIC	BIC
AYACUCHO	a30.39	96.41	161	NA	NA	NA
AYACUCHO	a40.49	171.5	172	NA	NA	NA
AYACUCHO	a50.59	215.1	191	NA	NA	NA
AYACUCHO	a60.69	346	205	NA	NA	NA
AYACUCHO	a70.79	499.8	208	NA	NA	NA
AYACUCHO	a80	485.4	209	NA	NA	NA
CAJAMARCA	a0.9	261.1	206	NA	NA	NA
CAJAMARCA	a10.19	53.35	143	NA	NA	NA
CAJAMARCA	a20.29	129.5	182	NA	NA	NA
CAJAMARCA	a30.39	183.2	193	NA	NA	NA
CAJAMARCA	a40.49	253.6	201	NA	NA	NA
CAJAMARCA	a50.59	553.2	209	NA	NA	NA
CAJAMARCA	a60.69	873.1	210	NA	NA	NA
CAJAMARCA	a70.79	754.7	211	NA	NA	NA
CAJAMARCA	a80	640.8	211	NA	NA	NA
CALLAO	a0.9	226.1	198	NA	NA	NA
CALLAO	a10.19	54.22	134	NA	NA	NA
CALLAO	a20.29	188.2	201	NA	NA	NA
CALLAO	a30.39	287.1	205	NA	NA	NA
CALLAO	a40.49	740.9	208	NA	NA	NA
CALLAO	a50.59	1510	211	NA	NA	NA
CALLAO	a60.69	2428	210	NA	NA	NA
CALLAO	a70.79	2093	211	NA	NA	NA
CALLAO	a80	1616	211	NA	NA	NA
CUSCO	a0.9	311	211	NA	NA	NA
CUSCO	a10.19	162.5	196	NA	NA	NA
CUSCO	a20.29	237.3	205	NA	NA	NA
CUSCO	a30.39	204.6	206	NA	NA	NA
CUSCO	a40.49	374.3	210	NA	NA	NA
CUSCO	a50.59	496.7	210	NA	NA	NA
CUSCO	a60.69	863.2	211	NA	NA	NA
CUSCO	a70.79	745.5	211	NA	NA	NA
CUSCO	a80	968.5	211	NA	NA	NA
HUANCAVELICA	a0.9	180.1	200	NA	NA	NA
HUANCAVELICA	a10.19	79.24	135	NA	NA	NA
HUANCAVELICA	a20.29	68.08	146	NA	NA	NA
HUANCAVELICA	a30.39	79.13	159	NA	NA	NA
HUANCAVELICA	a40.49	158	176	NA	NA	NA
HUANCAVELICA	a50.59	210.3	199	NA	NA	NA
HUANCAVELICA	a60.69	289.6	206	NA	NA	NA
HUANCAVELICA	a70.79	383.5	210	NA	NA	NA
HUANCAVELICA	a80	283	209	NA	NA	NA
HUANUCO	a0.9	197.9	207	NA	NA	NA
HUANUCO	a10.19	98.72	161	NA	NA	NA
HUANUCO	a20.29	118.3	190	NA	NA	NA
HUANUCO	a30.39	177.5	194	NA	NA	NA
HUANUCO	a40.49	248.7	204	NA	NA	NA
HUANUCO	a50.59	335.8	210	NA	NA	NA
HUANUCO	a60.69	450.2	210	NA	NA	NA
HUANUCO	a70.79	423	211	NA	NA	NA
HUANUCO	a80	399.5	210	NA	NA	NA
ICA	a0.9	176.6	204	NA	NA	NA

Departamento	range	null.deviance	df.null	logLik	AIC	BIC
ICA	a10.19	55.59	133	NA	NA	NA
ICA	a20.29	157.8	191	NA	NA	NA
ICA	a30.39	225.2	199	NA	NA	NA
ICA	a40.49	547.1	206	NA	NA	NA
ICA	a50.59	1065	211	NA	NA	NA
ICA	a60.69	1750	211	NA	NA	NA
ICA	a70.79	1184	209	NA	NA	NA
ICA	a80	1048	210	NA	NA	NA
JUNIN	a0.9	250.7	211	NA	NA	NA
JUNIN	a10.19	136.6	192	NA	NA	NA
JUNIN	a20.29	203.9	208	NA	NA	NA
JUNIN	a30.39	255.6	209	NA	NA	NA
JUNIN	a40.49	446.4	210	NA	NA	NA
JUNIN	a50.59	824.6	211	NA	NA	NA
JUNIN	a60.69	1085	211	NA	NA	NA
JUNIN	a70.79	907.6	211	NA	NA	NA
JUNIN	a80	929	211	NA	NA	NA
LA LIBERTAD	a0.9	250.4	209	NA	NA	NA
LA LIBERTAD	a10.19	181	190	NA	NA	NA
LA LIBERTAD	a20.29	244.2	208	NA	NA	NA
LA LIBERTAD	a30.39	270.6	210	NA	NA	NA
LA LIBERTAD	a40.49	567.4	210	NA	NA	NA
LA LIBERTAD	a50.59	1518	211	NA	NA	NA
LA LIBERTAD	a60.69	2366	211	NA	NA	NA
LA LIBERTAD	a70.79	2050	211	NA	NA	NA
LA LIBERTAD	a80	1492	211	NA	NA	NA
LAMBAYEQUE	a0.9	350.2	182	NA	NA	NA
LAMBAYEQUE	a10.19	60.08	114	NA	NA	NA
LAMBAYEQUE	a20.29	150.1	163	NA	NA	NA
LAMBAYEQUE	a30.39	203	176	NA	NA	NA
LAMBAYEQUE	a40.49	571.5	182	NA	NA	NA
LAMBAYEQUE	a50.59	1090	186	NA	NA	NA
LAMBAYEQUE	a60.69	2064	198	NA	NA	NA
LAMBAYEQUE	a70.79	2191	203	NA	NA	NA
LAMBAYEQUE	a80	3716	205	NA	NA	NA
LIMA	a0.9	463.5	211	NA	NA	NA
LIMA	a10.19	282.5	211	NA	NA	NA
LIMA	a20.29	497.2	211	NA	NA	NA
LIMA	a30.39	1413	211	NA	NA	NA
LIMA	a40.49	5115	211	NA	NA	NA
LIMA	a50.59	12694	211	NA	NA	NA
LIMA	a60.69	20998	211	NA	NA	NA
LIMA	a70.79	18227	211	NA	NA	NA
LIMA	a80	15668	211	NA	NA	NA
LORETO	a0.9	249	209	NA	NA	NA
LORETO	a10.19	79.99	150	NA	NA	NA
LORETO	a20.29	143.8	191	NA	NA	NA
LORETO	a30.39	218.4	199	NA	NA	NA
LORETO	a40.49	447.8	206	NA	NA	NA
LORETO	a50.59	857.3	208	NA	NA	NA
LORETO	a60.69	1598	210	NA	NA	NA
LORETO	a70.79	1387	210	NA	NA	NA

Departamento	range	null.deviance	df.null	logLik	AIC	BIC
LORETO	a80	866.2	210	NA	NA	NA
MADRE DE DIOS	a0.9	74.38	132	NA	NA	NA
MADRE DE DIOS	a10.19	20.02	80	NA	NA	NA
MADRE DE DIOS	a20.29	72.1	153	NA	NA	NA
MADRE DE DIOS	a30.39	69.1	140	NA	NA	NA
MADRE DE DIOS	a40.49	93.59	149	NA	NA	NA
MADRE DE DIOS	a50.59	217.3	165	NA	NA	NA
MADRE DE DIOS	a60.69	262.2	176	NA	NA	NA
MADRE DE DIOS	a70.79	202.9	172	NA	NA	NA
MADRE DE DIOS	a80	174	176	NA	NA	NA
MOQUEGUA	a0.9	19.28	77	NA	NA	NA
MOQUEGUA	a10.19	4.309	39	NA	NA	NA
MOQUEGUA	a20.29	17.29	82	NA	NA	NA
MOQUEGUA	a30.39	16.74	79	NA	NA	NA
MOQUEGUA	a40.49	92.24	123	NA	NA	NA
MOQUEGUA	a50.59	253.6	157	NA	NA	NA
MOQUEGUA	a60.69	556	186	NA	NA	NA
MOQUEGUA	a70.79	718.1	203	NA	NA	NA
MOQUEGUA	a80	530.2	209	NA	NA	NA
PASCO	a0.9	69.04	140	NA	NA	NA
PASCO	a10.19	14.4	60	NA	NA	NA
PASCO	a20.29	31.61	101	NA	NA	NA
PASCO	a30.39	49.05	105	NA	NA	NA
PASCO	a40.49	58.92	131	NA	NA	NA
PASCO	a50.59	138.3	163	NA	NA	NA
PASCO	a60.69	211.9	169	NA	NA	NA
PASCO	a70.79	243.2	192	NA	NA	NA
PASCO	a80	291.5	196	NA	NA	NA
PIURA	a0.9	270.7	209	NA	NA	NA
PIURA	a10.19	111	169	NA	NA	NA
PIURA	a20.29	226.6	200	NA	NA	NA
PIURA	a30.39	359.6	206	NA	NA	NA
PIURA	a40.49	935.7	207	NA	NA	NA
PIURA	a50.59	1661	211	NA	NA	NA
PIURA	a60.69	3278	211	NA	NA	NA
PIURA	a70.79	2489	211	NA	NA	NA
PIURA	a80	1948	211	NA	NA	NA
PUNO	a0.9	265.3	209	NA	NA	NA
PUNO	a10.19	135.9	185	NA	NA	NA
PUNO	a20.29	217.2	206	NA	NA	NA
PUNO	a30.39	278.8	207	NA	NA	NA
PUNO	a40.49	403	210	NA	NA	NA
PUNO	a50.59	752.1	210	NA	NA	NA
PUNO	a60.69	1015	211	NA	NA	NA
PUNO	a70.79	926.8	211	NA	NA	NA
PUNO	a80	857.7	211	NA	NA	NA
SAN MARTIN	a0.9	213.3	206	NA	NA	NA
SAN MARTIN	a10.19	95.47	152	NA	NA	NA
SAN MARTIN	a20.29	120	188	NA	NA	NA
SAN MARTIN	a30.39	204.7	205	NA	NA	NA
SAN MARTIN	a40.49	327.1	208	NA	NA	NA
SAN MARTIN	a50.59	502	207	NA	NA	NA

Departamento	range	null.deviance	df.null	logLik	AIC	BIC
SAN MARTIN	a60.69	815.9	211	NA	NA	NA
SAN MARTIN	a70.79	605.2	210	NA	NA	NA
SAN MARTIN	a80	815.5	211	NA	NA	NA
TACNA	a0.9	54.88	128	NA	NA	NA
TACNA	a10.19	4.114	66	NA	NA	NA
TACNA	a20.29	47.15	125	NA	NA	NA
TACNA	a30.39	91.16	155	NA	NA	NA
TACNA	a40.49	161	174	NA	NA	NA
TACNA	a50.59	347.6	191	NA	NA	NA
TACNA	a60.69	602.6	201	NA	NA	NA
TACNA	a70.79	453.1	209	NA	NA	NA
TACNA	a80	360.2	209	NA	NA	NA
TUMBES	a0.9	57.83	131	NA	NA	NA
TUMBES	a10.19	5.142	58	NA	NA	NA
TUMBES	a20.29	29.23	109	NA	NA	NA
TUMBES	a30.39	53.22	130	NA	NA	NA
TUMBES	a40.49	98.7	152	NA	NA	NA
TUMBES	a50.59	274.3	179	NA	NA	NA
TUMBES	a60.69	472.5	191	NA	NA	NA
TUMBES	a70.79	375.1	197	NA	NA	NA
TUMBES	a80	398.4	210	NA	NA	NA
UCAYALI	a0.9	189.3	193	NA	NA	NA
UCAYALI	a10.19	43.62	126	NA	NA	NA
UCAYALI	a20.29	141	174	NA	NA	NA
UCAYALI	a30.39	170.5	177	NA	NA	NA
UCAYALI	a40.49	255.5	202	NA	NA	NA
UCAYALI	a50.59	518	205	NA	NA	NA
UCAYALI	a60.69	747.7	210	NA	NA	NA
UCAYALI	a70.79	899.3	206	NA	NA	NA
UCAYALI	a80	630	209	NA	NA	NA

deviance	df.residual	nobs	fit	dif.dev	df	p
63.16	128	137	0.1041	7.336	8	0.5008
19.4	66	75	0.2071	5.069	8	0.7502
20.89	98	107	0.07651	1.73	8	0.9882
34.49	111	120	0.1693	7.03	8	0.5334
74.37	151	160	0.0376	2.906	8	0.9401
113.7	164	173	0.09711	12.23	8	0.1414
175.1	171	180	0.3353	88.33	8	1.015e-15
230.5	192	201	0.05163	12.55	8	0.1283
246.4	201	210	0.1173	32.73	8	0.00006879
216.5	201	210	0.07825	18.38	8	0.01853
106.7	175	184	0.02046	2.23	8	0.9732
161.2	192	201	0.06239	10.72	8	0.2178
187.1	193	202	0.04697	9.22	8	0.3241
251.1	199	208	0.29	102.6	8	1.275e-18
328.5	201	210	0.5852	463.3	8	5.146e-95
336.3	203	212	0.7165	849.9	8	3.644e-178
400.4	203	212	0.6666	800.6	8	1.565e-167
454.5	203	212	0.5136	479.9	8	1.439e-98
95.77	161	170	0.08903	9.36	8	0.3129

deviance	df.residual	nobs	fit	dif.dev	df	p
20.37	96	105	0.06181	1.342	8	0.995
36.57	127	136	0.1138	4.696	8	0.7895
65.64	140	149	0.03788	2.585	8	0.9577
101.4	154	163	0.1582	19.06	8	0.01454
146.3	186	195	0.1958	35.63	8	0.00002053
236.7	194	203	0.1671	47.5	8	1.231e-07
206.1	202	211	0.2531	69.84	8	5.289e-12
242.8	202	211	0.3899	155.2	8	1.642e-29
222.2	199	208	0.07273	17.43	8	0.02597
93.53	164	173	0.1079	11.32	8	0.1844
180.8	193	202	0.04036	7.605	8	0.4729
205.3	201	210	0.1419	33.94	8	0.00004166
304.6	199	208	0.504	309.5	8	3.886e-62
339.5	202	211	0.7772	1184	8	2.68e-250
401.3	203	212	0.8401	2108	8	0
441.4	203	212	0.8282	2128	8	0
530.2	203	212	0.8116	2284	8	0
123.3	165	174	0.2162	34.03	8	0.00004016
49.06	114	123	0.07201	3.807	8	0.8741
90.34	159	168	0.1353	14.14	8	0.07819
85.03	153	162	0.118	11.38	8	0.1812
148.4	164	173	0.1348	23.12	8	0.003213
183.8	183	192	0.1455	31.29	8	0.0001246
216.2	197	206	0.3752	129.8	8	3.102e-24
289.3	200	209	0.4211	210.4	8	4.009e-41
257.6	201	210	0.4694	227.9	8	8.331e-45
222.6	198	207	0.1475	38.5	8	6.082e-06
51.37	135	144	0.03702	1.975	8	0.9818
121.5	174	183	0.06211	8.044	8	0.4292
158.6	185	194	0.1344	24.63	8	0.001796
197.2	193	202	0.2223	56.36	8	2.404e-09
272.4	201	210	0.5075	280.7	8	5.211e-56
330.7	202	211	0.6212	542.4	8	5.614e-112
323	203	212	0.572	431.7	8	3.088e-88
361.7	203	212	0.4356	279.1	8	1.132e-55
214.1	190	199	0.05334	12.06	8	0.1485
51.73	126	135	0.04593	2.49	8	0.9622
179.1	193	202	0.04879	9.185	8	0.3269
226.4	197	206	0.2115	60.72	8	3.374e-10
308	200	209	0.5843	432.9	8	1.71e-88
355	203	212	0.7649	1155	8	4.307e-244
433.4	202	211	0.8215	1995	8	0
410	203	212	0.8041	1683	8	0
491	203	212	0.6961	1125	8	1.792e-237
247.6	203	212	0.2039	63.4	8	1e-10
153.7	188	197	0.05443	8.846	8	0.3554
218.6	197	206	0.07858	18.64	8	0.01689
173.8	198	207	0.1507	30.84	8	0.0001497
305.8	202	211	0.1829	68.44	8	1.004e-11
337.3	202	211	0.3208	159.3	8	2.212e-30
403.5	203	212	0.5325	459.7	8	3.067e-94
361.5	203	212	0.5151	384	8	4.894e-78

deviance	df.residual	nobs	fit	dif.dev	df	p
448.6	203	212	0.5368	519.9	8	3.806e-107
168	192	201	0.06735	12.13	8	0.1454
71.26	127	136	0.1006	7.975	8	0.4359
60.85	138	147	0.1063	7.236	8	0.5114
76.66	151	160	0.03121	2.469	8	0.9631
137.4	168	177	0.1304	20.6	8	0.008303
178.4	191	200	0.1516	31.87	8	0.00009811
224.2	198	207	0.2259	65.42	8	3.977e-11
281.8	202	211	0.2652	101.7	8	1.915e-18
239.2	201	210	0.1546	43.76	8	6.321e-07
176.9	199	208	0.1063	21.04	8	0.007054
89.81	153	162	0.09024	8.908	8	0.3501
113.3	182	191	0.04221	4.991	8	0.7585
162.5	186	195	0.08425	14.95	8	0.06005
222.6	196	205	0.1052	26.16	8	0.0009844
276.6	202	211	0.1763	59.2	8	6.684e-10
258.2	202	211	0.4264	191.9	8	3.173e-37
290.1	203	212	0.314	132.8	8	7.306e-25
290.7	202	211	0.2723	108.8	8	6.753e-20
171	196	205	0.03186	5.628	8	0.6888
49.36	125	134	0.1121	6.23	8	0.6214
146.5	183	192	0.07158	11.29	8	0.1857
177.9	191	200	0.2099	47.27	8	1.36e-07
235.2	198	207	0.5701	311.9	8	1.217e-62
298.5	203	212	0.7198	766.7	8	3.103e-160
353.2	203	212	0.7981	1396	8	3.343e-296
399.9	201	210	0.6623	784.1	8	5.619e-164
450.3	202	211	0.5705	598.1	8	5.904e-124
241.7	203	212	0.03601	9.028	8	0.34
129.3	184	193	0.0532	7.267	8	0.5081
190.2	200	209	0.06737	13.74	8	0.08888
241.1	201	210	0.05696	14.56	8	0.06827
299.3	202	211	0.3295	147.1	8	7.976e-28
298.9	203	212	0.6376	525.7	8	2.137e-108
366.7	203	212	0.6621	718.3	8	8.164e-150
366.9	203	212	0.5957	540.7	8	1.305e-111
449.6	203	212	0.516	479.4	8	1.882e-98
231.4	201	210	0.07591	19.01	8	0.01482
170.4	182	191	0.05862	10.61	8	0.2247
231.1	200	209	0.05354	13.07	8	0.1093
235.4	202	211	0.1299	35.15	8	0.00002506
336.1	202	211	0.4076	231.2	8	1.619e-45
432.4	203	212	0.7151	1085	8	5.239e-229
466.8	203	212	0.8027	1900	8	0
455.5	203	212	0.7779	1595	8	0
515	203	212	0.6549	977.4	8	1.108e-205
163.9	174	183	0.5319	186.3	8	5.002e-36
49.37	106	115	0.1784	10.72	8	0.2182
106.7	155	164	0.2894	43.45	8	7.213e-07
135.4	168	177	0.3331	67.61	8	1.464e-11
276	174	183	0.5171	295.5	8	3.688e-59
288	178	187	0.7357	801.7	8	8.861e-168

deviance	df.residual	nobs	fit	dif.dev	df	p
383.1	190	199	0.8144	1681	8	0
412.4	195	204	0.8118	1778	8	0
525.3	197	206	0.8587	3191	8	0
378.2	203	212	0.184	85.27	8	4.235e-15
248.1	203	212	0.1219	34.45	8	0.00003372
394.7	203	212	0.2063	102.6	8	1.263e-18
412.8	203	212	0.7079	1001	8	1.15e-210
485.2	203	212	0.9051	4630	8	0
831.7	203	212	0.9345	11862	8	0
1068	203	212	0.9491	19930	8	0
1182	203	212	0.9351	17044	8	0
1476	203	212	0.9058	14192	8	0
227.6	201	210	0.08598	21.41	8	0.006144
74.43	142	151	0.06951	5.56	8	0.6964
136.3	183	192	0.05269	7.578	8	0.4757
172.2	191	200	0.2114	46.17	8	2.211e-07
242.2	198	207	0.4591	205.6	8	4.283e-40
327.1	200	209	0.6184	530.2	8	2.33e-109
393.3	202	211	0.7539	1205	8	8.391e-255
390.3	202	211	0.7186	996.6	8	8.132e-210
379	202	211	0.5624	487.1	8	4.022e-100
64.85	124	133	0.1281	9.526	8	0.2999
17.98	72	81	0.1018	2.037	8	0.9799
66.87	145	154	0.07247	5.225	8	0.7333
64.3	132	141	0.06952	4.804	8	0.7783
79.85	141	150	0.1468	13.74	8	0.08874
126.2	157	166	0.4192	91.1	8	2.778e-16
138.8	168	177	0.4707	123.4	8	6.534e-23
123.6	164	173	0.3905	79.22	8	7.021e-14
117.6	168	177	0.3243	56.41	8	2.344e-09
17.27	69	78	0.1039	2.004	8	0.9809
3.486	31	40	0.1911	0.8234	8	0.9991
15.12	74	83	0.1251	2.163	8	0.9756
14.04	71	80	0.1611	2.697	8	0.9519
60.24	115	124	0.3468	31.99	8	0.00009352
108.9	149	158	0.5708	144.8	8	2.406e-27
212.4	178	187	0.618	343.6	8	2.092e-69
258.7	195	204	0.6397	459.4	8	3.592e-94
254.3	201	210	0.5204	275.9	8	5.473e-55
58.75	132	141	0.149	10.29	8	0.2453
11.93	52	61	0.1715	2.47	8	0.9631
28.6	93	102	0.09542	3.017	8	0.9333
45.02	97	106	0.08213	4.028	8	0.8546
48.35	123	132	0.1794	10.57	8	0.2274
91.88	155	164	0.3357	46.44	8	1.961e-07
133.3	161	170	0.371	78.62	8	9.269e-14
179.8	184	193	0.2608	63.44	8	9.821e-11
229.4	188	197	0.2129	62.07	8	1.829e-10
239.1	201	210	0.1165	31.53	8	0.0001131
102.7	161	170	0.07488	8.312	8	0.4036
186.3	192	201	0.1777	40.27	8	2.857e-06
222.8	198	207	0.3803	136.7	8	1.128e-25

deviance	df.residual	nobs	fit	dif.dev	df	p
351.5	199	208	0.6243	584.2	8	5.897e-121
356.9	203	212	0.7851	1304	8	3.315e-276
462.9	203	212	0.8588	2815	8	0
438.6	203	212	0.8238	2050	8	0
574.6	203	212	0.705	1373	8	3.799e-291
211	201	210	0.2046	54.27	8	6.114e-09
125.3	177	186	0.07841	10.66	8	0.2219
204.5	198	207	0.05853	12.71	8	0.1222
272.2	199	208	0.0238	6.635	8	0.5765
296.3	202	211	0.2647	106.7	8	1.852e-19
342.8	202	211	0.5443	409.3	8	1.88e-83
337.8	203	212	0.6671	676.9	8	6.763e-141
353.2	203	212	0.6189	573.6	8	1.106e-118
453.4	203	212	0.4714	404.3	8	2.256e-82
187	198	207	0.1234	26.31	8	0.0009299
84.75	144	153	0.1122	10.71	8	0.2184
110.1	180	189	0.08208	9.847	8	0.276
174	197	206	0.1503	30.76	8	0.0001547
238.1	200	209	0.2719	88.93	8	7.666e-16
287.2	199	208	0.4279	214.8	8	4.756e-42
333.7	203	212	0.5909	482.2	8	4.741e-99
301.8	202	211	0.5013	303.4	8	7.942e-61
379.5	203	212	0.5346	436	8	3.736e-89
48.73	120	129	0.1122	6.156	8	0.6298
3.41	58	67	0.171	0.7036	8	0.9995
42.47	117	126	0.0993	4.682	8	0.791
88.65	147	156	0.02755	2.511	8	0.9612
126.7	166	175	0.213	34.29	8	0.00003604
211.7	183	192	0.3909	135.9	8	1.709e-25
290.9	193	202	0.5173	311.7	8	1.304e-62
272.7	201	210	0.3982	180.4	8	8.331e-35
308.6	201	210	0.1433	51.63	8	1.985e-08
55.09	123	132	0.04737	2.739	8	0.9496
3.969	50	59	0.2282	1.173	8	0.9969
27.64	101	110	0.05437	1.589	8	0.9911
47.87	122	131	0.1005	5.349	8	0.7197
75.36	144	153	0.2365	23.34	8	0.00295
146.2	171	180	0.4668	128	8	7.229e-24
209.8	183	192	0.556	262.7	8	3.448e-52
214.6	189	198	0.4279	160.5	8	1.264e-30
248.4	202	211	0.3766	150.1	8	1.909e-28
154.3	185	194	0.1848	34.97	8	0.00002711
40.99	118	127	0.0603	2.63	8	0.9554
127	166	175	0.09946	14.02	8	0.08117
142.8	169	178	0.1622	27.65	8	0.0005449
203.5	194	203	0.2036	52.01	8	1.673e-08
277.6	197	206	0.464	240.3	8	1.927e-47
331.1	202	211	0.5572	416.6	8	5.261e-85
297.1	198	207	0.6696	602.2	8	8.045e-125
334.6	201	210	0.469	295.5	8	3.82e-59

8.1.2 Under-registration rates

```
## # A tibble: 25 x 2
## # Groups: Departamento [25]
##   Departamento sub.mean
##   <fct>          <dbl>
## 1 ICA            99.9
## 2 MADRE DE DIOS 99.2
## 3 AREQUIPA       94.8
## 4 ANCASH         85.1
## 5 TUMBES         85.0
## 6 JUNIN          82.3
## 7 MOQUEGUA      82.1
## 8 CUSCO          80.8
## 9 LA LIBERTAD    80.4
## 10 HUANCAVELICA 77.7
## 11 CALLAO        77.5
## 12 TACNA         74.4
## 13 PUNO          62.3
## 14 UCAYALI       61.8
## 15 LIMA           58.1
## 16 APURIMAC      57.0
## 17 SAN MARTIN    51.5
## 18 AYACUCHO       50.5
## 19 PIURA          49.5
## 20 HUANUCO        46.6
## 21 PASCO          40.6
## 22 LORETO         32.1
## 23 CAJAMARCA     28.5
## 24 AMAZONAS       28.0
## 25 LAMBAYEQUE    11.9
```

8.1.3 mortality rates

```
## # A tibble: 25 x 8
##   Departamento adj.rate   lci   uci crude.rate `2010` `2015` Difference
##   <chr>          <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 AMAZONAS        4.59  4.39  4.80  4.60  6.05  6.19  -1.59
## 2 ANCASH          8.43  8.27  8.60  8.41  6.09  6.15   2.26
## 3 APURIMAC        6.90  6.66  7.14  7.49  6.76  6.61   0.882
## 4 AREQUIPA        9.09  8.93  9.25  8.24  5.53  5.8    2.44
## 5 AYACUCHO         5.56  5.39  5.74  6.04  6.15  5.91   0.131
## 6 CAJAMARCA        5.80  5.68  5.93  6.17  5.39  5.5    0.666
## 7 CALLAO           11.0  10.8  11.2  10.6  4.91  5.27   5.30
## 8 CUSCO            6.86  6.72  7.00  6.84  6.88  6.97  -0.133
## 9 HUANCAVELICA    5.84  5.63  6.05  8.18  5.83  5.54   2.64
## 10 HUANUCO          7.01  6.84  7.19  8.20  5.94  5.98   2.22
## # ... with 15 more rows
```

8.1.4 Counterfacutual and total deaths 2020

```
##   Departamento range sinadef excess.T excess.l excess.u excess.reg
##   <chr>          <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 AMAZONAS        a0.9   60    0.000  0.000 0.00e+00  0.0
## 2 AMAZONAS        a10.19  22    2.797  2.797 2.80e+00  0.0
## 3 AMAZONAS        a20.29  36    0.000  0.000 0.00e+00  0.0
```

##	AMAZONAS	a30.39	60	6.929	6.929	6.93e+00	0.0
##	AMAZONAS	a40.49	83	17.415	17.415	1.74e+01	0.0
##	AMAZONAS	a50.59	123	9.589	9.589	9.59e+00	0.0
##	AMAZONAS	a60.69	199	165.887	125.102	1.91e+02	96.5
##	AMAZONAS	a70.79	199	0.000	0.000	0.00e+00	0.0
##	AMAZONAS	a80	337	171.535	80.850	2.28e+02	99.7
##	ANCASH	a0.9	243	42.116	42.116	4.21e+01	0.0
##	ANCASH	a10.19	101	0.000	0.000	0.00e+00	0.0
##	ANCASH	a20.29	163	0.000	0.000	0.00e+00	0.0
##	ANCASH	a30.39	224	27.659	27.659	2.77e+01	0.0
##	ANCASH	a40.49	471	258.581	217.321	2.88e+02	225.1
##	ANCASH	a50.59	892	653.851	624.299	6.78e+02	569.1
##	ANCASH	a60.69	1391	1094.688	1068.452	1.12e+03	952.8
##	ANCASH	a70.79	1960	1310.347	1254.640	1.36e+03	1140.6
##	ANCASH	a80	3133	1206.919	1017.742	1.36e+03	1050.5
##	APURIMAC	a0.9	100	-34.010	1649.604	8.77e+01	-83.8
##	APURIMAC	a10.19	51	0.974	0.974	9.74e-01	0.0
##	APURIMAC	a20.29	64	2.862	2.862	2.86e+00	0.0
##	APURIMAC	a30.39	73	0.000	0.000	0.00e+00	0.0
##	APURIMAC	a40.49	119	0.000	0.000	0.00e+00	0.0
##	APURIMAC	a50.59	200	0.000	0.000	0.00e+00	0.0
##	APURIMAC	a60.69	275	0.000	0.000	0.00e+00	0.0
##	APURIMAC	a70.79	449	0.000	0.000	0.00e+00	0.0
##	APURIMAC	a80	863	0.000	0.000	0.00e+00	0.0
##	AREQUIPA	a0.9	216	-0.706	-558.878	1.03e+02	-90.2
##	AREQUIPA	a10.19	98	5.150	-1272.499	6.66e+01	-54.6
##	AREQUIPA	a20.29	232	53.939	53.939	5.39e+01	0.0
##	AREQUIPA	a30.39	329	0.000	0.000	0.00e+00	0.0
##	AREQUIPA	a40.49	662	323.316	266.442	3.64e+02	307.3
##	AREQUIPA	a50.59	1213	756.726	715.655	7.91e+02	719.2
##	AREQUIPA	a60.69	1990	1354.212	1307.535	1.39e+03	1287.0
##	AREQUIPA	a70.79	2480	1415.300	1324.757	1.49e+03	1345.1
##	AREQUIPA	a80	4307	2103.229	1947.117	2.24e+03	1998.9
##	AYACUCHO	a0.9	122	-55.780	1820.458	1.20e+02	-114.8
##	AYACUCHO	a10.19	62	1.767	1.767	1.77e+00	0.0
##	AYACUCHO	a20.29	95	25.573	25.573	2.56e+01	0.0
##	AYACUCHO	a30.39	100	19.597	19.597	1.96e+01	0.0
##	AYACUCHO	a40.49	161	35.609	35.609	3.56e+01	0.0
##	AYACUCHO	a50.59	214	0.000	0.000	0.00e+00	0.0
##	AYACUCHO	a60.69	373	65.908	65.908	6.59e+01	0.0
##	AYACUCHO	a70.79	554	326.403	232.379	3.90e+02	218.4
##	AYACUCHO	a80	843	0.000	0.000	0.00e+00	0.0
##	CAJAMARCA	a0.9	232	-116.622	-3985.197	1.40e+02	-165.8
##	CAJAMARCA	a10.19	75	21.831	21.831	2.18e+01	0.0
##	CAJAMARCA	a20.29	128	41.127	41.127	4.11e+01	0.0
##	CAJAMARCA	a30.39	207	0.000	0.000	0.00e+00	0.0
##	CAJAMARCA	a40.49	294	160.024	65.197	2.16e+02	93.3
##	CAJAMARCA	a50.59	539	348.689	229.545	4.27e+02	203.3
##	CAJAMARCA	a60.69	793	579.875	429.501	6.82e+02	338.0
##	CAJAMARCA	a70.79	1071	766.382	639.497	8.63e+02	446.8
##	CAJAMARCA	a80	1752	881.212	625.159	1.08e+03	513.7
##	CALLAO	a0.9	156	0.000	0.000	0.00e+00	0.0
##	CALLAO	a10.19	49	11.571	11.571	1.16e+01	0.0
##	CALLAO	a20.29	185	41.380	41.380	4.14e+01	0.0

##	CALLAO	a30.39	319	106.690	27.916	1.52e+02	87.1
##	CALLAO	a40.49	641	559.143	542.794	5.73e+02	456.5
##	CALLAO	a50.59	1192	1146.817	1136.497	1.16e+03	936.2
##	CALLAO	a60.69	1958	1782.091	1757.898	1.80e+03	1454.9
##	CALLAO	a70.79	2222	1994.467	1969.137	2.02e+03	1628.3
##	CALLAO	a80	2982	2196.083	2129.402	2.26e+03	1792.9
##	CUSCO	a0.9	354	66.995	66.995	6.70e+01	0.0
##	CUSCO	a10.19	178	4.158	4.158	4.16e+00	0.0
##	CUSCO	a20.29	246	76.133	76.133	7.61e+01	0.0
##	CUSCO	a30.39	319	0.000	0.000	0.00e+00	0.0
##	CUSCO	a40.49	519	0.000	0.000	0.00e+00	0.0
##	CUSCO	a50.59	789	0.000	0.000	0.00e+00	0.0
##	CUSCO	a60.69	1193	421.071	231.231	5.50e+02	353.2
##	CUSCO	a70.79	1610	0.000	0.000	0.00e+00	0.0
##	CUSCO	a80	2451	762.746	496.928	9.65e+02	639.7
##	HUANCABELICA	a0.9	141	64.947	33.753	8.23e+01	53.1
##	HUANCABELICA	a10.19	60	4.598	4.598	4.60e+00	0.0
##	HUANCABELICA	a20.29	76	0.000	0.000	0.00e+00	0.0
##	HUANCABELICA	a30.39	79	18.865	18.865	1.89e+01	0.0
##	HUANCABELICA	a40.49	152	0.000	0.000	0.00e+00	0.0
##	HUANCABELICA	a50.59	232	100.080	54.233	1.28e+02	81.8
##	HUANCABELICA	a60.69	355	217.753	183.090	2.42e+02	178.1
##	HUANCABELICA	a70.79	548	308.899	255.047	3.48e+02	252.6
##	HUANCABELICA	a80	782	0.786	0.786	7.86e-01	0.0
##	HUANUCO	a0.9	187	10.678	10.678	1.07e+01	0.0
##	HUANUCO	a10.19	71	1.796	1.796	1.80e+00	0.0
##	HUANUCO	a20.29	125	0.000	0.000	0.00e+00	0.0
##	HUANUCO	a30.39	166	16.623	16.623	1.66e+01	0.0
##	HUANUCO	a40.49	267	7.246	7.246	7.25e+00	0.0
##	HUANUCO	a50.59	411	240.295	162.454	2.91e+02	156.6
##	HUANUCO	a60.69	670	400.108	300.117	4.71e+02	260.8
##	HUANUCO	a70.79	908	425.091	283.569	5.27e+02	277.1
##	HUANUCO	a80	1233	490.408	295.907	6.33e+02	319.7
##	ICA	a0.9	170	50.749	50.749	5.07e+01	0.0
##	ICA	a10.19	52	9.276	9.276	9.28e+00	0.0
##	ICA	a20.29	163	33.301	33.301	3.33e+01	0.0
##	ICA	a30.39	256	95.823	63.728	1.16e+02	95.7
##	ICA	a40.49	527	298.629	274.786	3.17e+02	298.2
##	ICA	a50.59	967	633.300	612.556	6.51e+02	632.4
##	ICA	a60.69	1468	1109.486	1093.909	1.12e+03	1107.9
##	ICA	a70.79	1638	817.414	742.587	8.78e+02	816.2
##	ICA	a80	2521	1148.712	1037.511	1.24e+03	1147.0
##	JUNIN	a0.9	372	62.256	62.256	6.23e+01	0.0
##	JUNIN	a10.19	151	2.607	2.607	2.61e+00	0.0
##	JUNIN	a20.29	219	69.883	69.883	6.99e+01	0.0
##	JUNIN	a30.39	322	36.910	36.910	3.69e+01	0.0
##	JUNIN	a40.49	553	285.279	217.193	3.32e+02	242.3
##	JUNIN	a50.59	947	545.545	479.919	5.96e+02	463.4
##	JUNIN	a60.69	1394	755.344	658.311	8.31e+02	641.6
##	JUNIN	a70.79	1806	876.830	755.274	9.74e+02	744.8
##	JUNIN	a80	2746	923.230	692.372	1.11e+03	784.2
##	LA LIBERTAD	a0.9	334	47.575	47.575	4.76e+01	0.0
##	LA LIBERTAD	a10.19	142	24.410	24.410	2.44e+01	0.0
##	LA LIBERTAD	a20.29	297	67.088	67.088	6.71e+01	0.0

##	LA LIBERTAD	a30.39	409	154.407	83.940	2.00e+02	129.1
##	LA LIBERTAD	a40.49	743	407.672	333.163	4.61e+02	340.9
##	LA LIBERTAD	a50.59	1443	1135.426	1097.601	1.17e+03	949.4
##	LA LIBERTAD	a60.69	2323	1734.597	1673.595	1.79e+03	1450.4
##	LA LIBERTAD	a70.79	2871	2030.325	1951.264	2.10e+03	1697.7
##	LA LIBERTAD	a80	4055	1915.679	1703.278	2.09e+03	1601.8
##	LAMBAYEQUE	a0.9	81	26.550	26.550	2.66e+01	0.0
##	LAMBAYEQUE	a10.19	29	0.000	0.000	0.00e+00	0.0
##	LAMBAYEQUE	a20.29	90	0.000	0.000	0.00e+00	0.0
##	LAMBAYEQUE	a30.39	175	204.133	190.371	2.14e+02	108.5
##	LAMBAYEQUE	a40.49	368	588.050	582.468	5.93e+02	312.6
##	LAMBAYEQUE	a50.59	734	1258.995	1256.009	1.26e+03	669.2
##	LAMBAYEQUE	a60.69	1231	2067.809	2062.608	2.07e+03	1099.1
##	LAMBAYEQUE	a70.79	1194	2004.897	1998.611	2.01e+03	1065.6
##	LAMBAYEQUE	a80	1298	1950.664	1929.131	1.97e+03	1036.8
##	LIMA	a0.9	1374	63.727	63.727	6.37e+01	0.0
##	LIMA	a10.19	483	0.000	0.000	0.00e+00	0.0
##	LIMA	a20.29	1297	412.120	190.794	5.69e+02	290.4
##	LIMA	a30.39	2631	1869.322	1756.338	1.97e+03	1317.1
##	LIMA	a40.49	5419	5542.869	5489.824	5.59e+03	3905.5
##	LIMA	a50.59	10548	11383.436	11312.181	1.15e+04	8020.8
##	LIMA	a60.69	17129	18620.741	18519.995	1.87e+04	13120.2
##	LIMA	a70.79	19325	19153.432	18971.491	1.93e+04	13495.5
##	LIMA	a80	27134	21319.870	20811.895	2.18e+04	15022.0
##	LORETO	a0.9	275	-98.112	-1473.659	1.53e+02	-153.4
##	LORETO	a10.19	85	39.533	39.533	3.95e+01	0.0
##	LORETO	a20.29	152	29.907	29.907	2.99e+01	0.0
##	LORETO	a30.39	255	119.193	39.243	1.67e+02	71.0
##	LORETO	a40.49	389	406.040	381.931	4.25e+02	241.9
##	LORETO	a50.59	615	787.209	775.699	7.97e+02	468.9
##	LORETO	a60.69	958	1279.430	1267.497	1.29e+03	762.1
##	LORETO	a70.79	933	1214.444	1200.033	1.23e+03	723.4
##	LORETO	a80	785	876.719	846.403	9.02e+02	522.2
##	MADRE DE DIOS	a0.9	92	24.801	24.801	2.48e+01	0.0
##	MADRE DE DIOS	a10.19	21	0.000	0.000	0.00e+00	0.0
##	MADRE DE DIOS	a20.29	79	27.295	27.295	2.73e+01	0.0
##	MADRE DE DIOS	a30.39	92	0.000	0.000	0.00e+00	0.0
##	MADRE DE DIOS	a40.49	117	0.000	0.000	0.00e+00	0.0
##	MADRE DE DIOS	a50.59	172	91.194	77.583	1.00e+02	90.4
##	MADRE DE DIOS	a60.69	191	102.006	87.849	1.12e+02	101.2
##	MADRE DE DIOS	a70.79	168	72.214	47.276	8.72e+01	71.6
##	MADRE DE DIOS	a80	146	86.401	78.982	9.18e+01	85.7
##	MOQUEGUA	a0.9	21	15.708	15.708	1.57e+01	0.0
##	MOQUEGUA	a10.19	10	0.000	0.000	0.00e+00	0.0
##	MOQUEGUA	a20.29	39	10.259	10.259	1.03e+01	0.0
##	MOQUEGUA	a30.39	33	3.288	3.288	3.29e+00	0.0
##	MOQUEGUA	a40.49	90	58.702	48.197	6.52e+01	49.8
##	MOQUEGUA	a50.59	173	137.157	124.482	1.46e+02	116.4
##	MOQUEGUA	a60.69	277	229.320	215.414	2.39e+02	194.6
##	MOQUEGUA	a70.79	380	279.254	251.401	2.99e+02	236.9
##	MOQUEGUA	a80	506	242.534	172.625	2.89e+02	205.8
##	PASCO	a0.9	91	26.258	26.258	2.63e+01	0.0
##	PASCO	a10.19	28	0.000	0.000	0.00e+00	0.0
##	PASCO	a20.29	45	19.109	19.109	1.91e+01	0.0

##	PASCO	a30.39	58	15.138	15.138	1.51e+01	0.0
##	PASCO	a40.49	83	-24.767	-11571.736	7.18e+01	-60.2
##	PASCO	a50.59	152	115.403	85.517	1.34e+02	72.4
##	PASCO	a60.69	219	0.000	0.000	0.00e+00	0.0
##	PASCO	a70.79	245	19.225	19.225	1.92e+01	0.0
##	PASCO	a80	294	2.808	2.808	2.81e+00	0.0
##	PIURA	a0.9	302	126.576	126.576	1.27e+02	0.0
##	PIURA	a10.19	91	0.000	0.000	0.00e+00	0.0
##	PIURA	a20.29	219	24.479	24.479	2.45e+01	0.0
##	PIURA	a30.39	362	308.641	277.136	3.32e+02	205.0
##	PIURA	a40.49	743	845.867	828.623	8.61e+02	561.9
##	PIURA	a50.59	1387	1676.237	1662.437	1.69e+03	1113.5
##	PIURA	a60.69	2446	2962.756	2938.073	2.99e+03	1968.2
##	PIURA	a70.79	2514	2740.362	2698.620	2.78e+03	1820.5
##	PIURA	a80	3251	2827.183	2706.771	2.93e+03	1878.1
##	PUNO	a0.9	291	-60.413	-1075.595	1.51e+02	-170.7
##	PUNO	a10.19	129	13.174	13.174	1.32e+01	0.0
##	PUNO	a20.29	230	33.647	33.647	3.36e+01	0.0
##	PUNO	a30.39	282	9.589	9.589	9.59e+00	0.0
##	PUNO	a40.49	486	0.000	0.000	0.00e+00	0.0
##	PUNO	a50.59	788	296.233	95.268	4.20e+02	215.1
##	PUNO	a60.69	1128	606.650	446.629	7.20e+02	440.4
##	PUNO	a70.79	1551	710.265	514.467	8.56e+02	515.7
##	PUNO	a80	2134	932.142	672.852	1.13e+03	676.8
##	SAN MARTIN	a0.9	191	-23.652	-553.404	9.34e+01	-69.5
##	SAN MARTIN	a10.19	69	0.000	0.000	0.00e+00	0.0
##	SAN MARTIN	a20.29	153	48.117	48.117	4.81e+01	0.0
##	SAN MARTIN	a30.39	208	19.630	19.630	1.96e+01	0.0
##	SAN MARTIN	a40.49	322	246.122	205.499	2.74e+02	165.8
##	SAN MARTIN	a50.59	522	376.699	314.843	4.21e+02	253.8
##	SAN MARTIN	a60.69	744	617.194	560.598	6.60e+02	415.8
##	SAN MARTIN	a70.79	796	484.915	380.845	5.60e+02	326.6
##	SAN MARTIN	a80	1207	678.233	525.374	7.91e+02	456.9
##	TACNA	a0.9	39	7.379	7.379	7.38e+00	0.0
##	TACNA	a10.19	18	5.526	5.526	5.53e+00	0.0
##	TACNA	a20.29	49	0.000	0.000	0.00e+00	0.0
##	TACNA	a30.39	77	0.000	0.000	0.00e+00	0.0
##	TACNA	a40.49	131	0.000	0.000	0.00e+00	0.0
##	TACNA	a50.59	225	0.000	0.000	0.00e+00	0.0
##	TACNA	a60.69	375	226.448	157.509	2.67e+02	180.3
##	TACNA	a70.79	396	197.119	127.779	2.40e+02	157.0
##	TACNA	a80	444	162.710	66.081	2.21e+02	129.6
##	TUMBES	a0.9	58	12.295	12.295	1.23e+01	0.0
##	TUMBES	a10.19	20	3.396	3.396	3.40e+00	0.0
##	TUMBES	a20.29	47	7.290	7.290	7.29e+00	0.0
##	TUMBES	a30.39	80	0.000	0.000	0.00e+00	0.0
##	TUMBES	a40.49	116	57.964	40.370	6.88e+01	50.4
##	TUMBES	a50.59	232	191.674	185.160	1.97e+02	166.7
##	TUMBES	a60.69	366	299.283	291.012	3.06e+02	260.3
##	TUMBES	a70.79	354	238.163	218.495	2.53e+02	207.1
##	TUMBES	a80	505	229.170	182.225	2.63e+02	199.3
##	UCAYALI	a0.9	137	64.582	64.582	6.46e+01	0.0
##	UCAYALI	a10.19	57	2.327	2.327	2.33e+00	0.0
##	UCAYALI	a20.29	135	0.000	0.000	0.00e+00	0.0

##	UCAYALI	a30.39	170	96.321	64.443	1.16e+02	69.7
##	UCAYALI	a40.49	213	128.925	96.269	1.50e+02	93.3
##	UCAYALI	a50.59	422	311.732	283.988	3.33e+02	225.5
##	UCAYALI	a60.69	596	551.689	533.127	5.67e+02	399.1
##	UCAYALI	a70.79	655	586.209	565.647	6.03e+02	424.1
##	UCAYALI	a80	649	476.759	434.215	5.10e+02	344.9
##	Total	-	211669	153832.058	129069.633	1.61e+05	110113.1
##	excess.reg.l	excess.reg.u		count	count.l	count.u	total
##	0.0	0.000		60.0	60.0	60.0	103.2
##	0.0	0.000		22.0	22.0	22.0	40.6
##	0.0	0.000		36.0	36.0	36.0	61.9
##	0.0	0.000		60.0	60.0	60.0	110.1
##	0.0	0.000		83.0	83.0	83.0	160.2
##	0.0	0.000		123.0	123.0	123.0	221.1
##	72.7	111.024		102.5	88.0	126.3	342.2
##	0.0	0.000		199.0	199.0	199.0	342.2
##	47.0	132.518		237.3	204.5	290.0	579.6
##	0.0	0.000		243.0	243.0	243.0	321.3
##	0.0	0.000		101.0	101.0	101.0	116.0
##	0.0	0.000		163.0	163.0	163.0	187.3
##	0.0	0.000		224.0	224.0	224.0	285.0
##	189.2	250.982		245.9	220.0	281.8	541.1
##	543.4	590.281		322.9	301.7	348.6	1024.8
##	930.0	972.795		438.2	418.2	461.0	1598.1
##	1092.1	1182.194		819.4	777.8	867.9	2251.8
##	885.9	1186.202		2082.5	1946.8	2247.1	3599.4
##	1093.6	1.362		183.8	98.6	-993.6	228.7
##	0.0	0.000		51.0	51.0	51.0	73.9
##	0.0	0.000		64.0	64.0	64.0	94.4
##	0.0	0.000		73.0	73.0	73.0	104.4
##	0.0	0.000		119.0	119.0	119.0	170.2
##	0.0	0.000		200.0	200.0	200.0	286.0
##	0.0	0.000		275.0	275.0	275.0	393.2
##	0.0	0.000		449.0	449.0	449.0	642.0
##	0.0	0.000		863.0	863.0	863.0	1234.0
##	-620.7	8.744		306.2	207.3	836.7	321.4
##	-1268.9	3.748		152.6	94.3	1366.9	165.7
##	0.0	0.000		232.0	232.0	232.0	298.0
##	0.0	0.000		329.0	329.0	329.0	346.2
##	253.2	346.109		354.7	315.9	408.8	696.5
##	680.2	751.346		493.8	461.7	532.8	1276.3
##	1242.7	1325.072		703.0	664.9	747.3	2093.9
##	1259.0	1416.885		1134.9	1063.1	1221.0	2609.4
##	1850.5	2125.511		2308.1	2181.5	2456.5	4531.8
##	1140.6	2.645		236.8	119.4	-1018.6	298.1
##	0.0	0.000		62.0	62.0	62.0	94.4
##	0.0	0.000		95.0	95.0	95.0	167.6
##	0.0	0.000		100.0	100.0	100.0	169.1
##	0.0	0.000		161.0	161.0	161.0	276.2
##	0.0	0.000		214.0	214.0	214.0	319.8
##	0.0	0.000		373.0	373.0	373.0	623.4
##	155.5	260.926		335.6	293.1	398.5	828.0
##	0.0	0.000		843.0	843.0	843.0	1259.9
##	-2420.9	-16.052		397.8	248.1	2652.9	565.8
						-3559.7	4691.2

##	0.0	0.000	75.0	75.0	75.0	150.5	150.5	150.5
##	0.0	0.000	128.0	128.0	128.0	260.7	260.7	260.7
##	0.0	0.000	207.0	207.0	207.0	355.1	355.1	355.1
##	38.0	125.951	200.7	168.0	256.0	504.3	353.5	655.2
##	133.8	248.876	335.7	290.1	405.2	924.6	727.2	1122.0
##	250.4	397.661	455.0	395.3	542.6	1360.3	1107.7	1613.0
##	372.8	503.105	624.2	567.9	698.2	1837.2	1613.7	2060.8
##	364.4	627.418	1238.3	1124.6	1387.6	3005.5	2554.3	3456.6
##	0.0	0.000	156.0	156.0	156.0	191.1	191.1	191.1
##	0.0	0.000	49.0	49.0	49.0	71.6	71.6	71.6
##	0.0	0.000	185.0	185.0	185.0	268.0	268.0	268.0
##	22.8	124.336	231.9	194.7	296.2	390.7	266.4	515.1
##	443.1	467.614	184.5	173.4	197.9	785.2	755.2	815.2
##	927.8	943.902	255.8	248.1	264.2	1460.1	1440.4	1479.8
##	1435.1	1472.639	503.1	485.4	522.9	2398.4	2352.4	2444.3
##	1607.6	1647.044	593.7	575.0	614.4	2721.7	2673.4	2770.1
##	1738.4	1840.972	1189.1	1141.0	1243.6	3652.7	3527.1	3778.3
##	0.0	0.000	354.0	354.0	354.0	489.1	489.1	489.1
##	0.0	0.000	178.0	178.0	178.0	216.4	216.4	216.4
##	0.0	0.000	246.0	246.0	246.0	369.4	369.4	369.4
##	0.0	0.000	319.0	319.0	319.0	380.3	380.3	380.3
##	0.0	0.000	519.0	519.0	519.0	618.8	618.8	618.8
##	0.0	0.000	789.0	789.0	789.0	940.7	940.7	940.7
##	193.9	461.140	839.8	731.9	999.1	1422.4	1103.8	1740.9
##	0.0	0.000	1610.0	1610.0	1610.0	1919.5	1919.5	1919.5
##	416.8	809.310	1811.3	1641.7	2034.2	2922.2	2454.3	3390.2
##	27.6	67.294	87.9	73.7	113.4	172.4	123.9	221.0
##	0.0	0.000	60.0	60.0	60.0	78.0	78.0	78.0
##	0.0	0.000	76.0	76.0	76.0	92.9	92.9	92.9
##	0.0	0.000	79.0	79.0	79.0	115.5	115.5	115.5
##	0.0	0.000	152.0	152.0	152.0	185.9	185.9	185.9
##	44.3	104.392	150.2	127.6	187.7	283.7	210.3	357.1
##	149.7	197.986	176.9	157.0	205.3	434.1	375.1	493.1
##	208.6	284.231	295.4	263.8	339.4	670.1	577.6	762.6
##	0.0	0.000	782.0	782.0	782.0	957.1	957.1	957.1
##	0.0	0.000	187.0	187.0	187.0	297.5	297.5	297.5
##	0.0	0.000	71.0	71.0	71.0	110.7	110.7	110.7
##	0.0	0.000	125.0	125.0	125.0	191.8	191.8	191.8
##	0.0	0.000	166.0	166.0	166.0	271.3	271.3	271.3
##	0.0	0.000	267.0	267.0	267.0	416.8	416.8	416.8
##	105.9	189.799	254.4	221.2	305.1	630.5	501.8	759.2
##	195.6	307.139	409.2	362.9	474.4	1027.8	856.8	1198.9
##	184.9	343.700	630.9	564.3	723.1	1392.9	1149.2	1636.6
##	192.9	412.907	913.3	820.1	1040.1	1891.5	1554.0	2229.0
##	0.0	0.000	170.0	170.0	170.0	221.0	221.0	221.0
##	0.0	0.000	52.0	52.0	52.0	61.4	61.4	61.4
##	0.0	0.000	163.0	163.0	163.0	196.5	196.5	196.5
##	63.6	116.218	160.3	139.8	192.4	256.4	203.7	309.0
##	274.4	316.715	228.8	210.3	252.6	527.8	485.4	570.2
##	611.7	650.016	334.6	317.0	355.3	968.4	930.0	1006.8
##	1092.3	1121.751	360.1	346.2	375.7	1470.1	1440.7	1499.6
##	741.5	876.445	821.8	761.6	896.5	1640.4	1505.3	1775.5
##	1036.0	1238.615	1374.0	1282.4	1485.0	2524.7	2321.8	2727.6
##	0.0	0.000	372.0	372.0	372.0	500.2	500.2	500.2

##	0.0	0.000	151.0	151.0	151.0	180.4	180.4	180.4
##	0.0	0.000	219.0	219.0	219.0	327.7	327.7	327.7
##	0.0	0.000	322.0	322.0	322.0	416.0	416.0	416.0
##	184.5	281.751	310.7	271.2	368.5	651.0	536.5	765.5
##	407.7	506.288	483.6	440.7	539.3	1114.9	998.7	1231.0
##	559.2	705.787	752.4	688.2	834.8	1641.1	1468.5	1813.7
##	641.6	827.036	1061.2	979.0	1164.4	2126.1	1907.8	2344.5
##	588.1	939.354	1961.8	1806.6	2157.9	3232.7	2819.2	3646.2
##	0.0	0.000	334.0	334.0	334.0	447.0	447.0	447.0
##	0.0	0.000	142.0	142.0	142.0	194.2	194.2	194.2
##	0.0	0.000	297.0	297.0	297.0	422.3	422.3	422.3
##	70.2	167.099	279.9	241.9	338.8	489.1	373.2	605.0
##	278.6	385.484	402.1	357.5	464.4	888.6	760.7	1016.4
##	917.8	976.255	493.6	466.7	525.2	1725.8	1655.8	1795.7
##	1399.4	1494.374	872.6	828.6	923.6	2778.2	2664.6	2891.8
##	1631.6	1755.039	1173.3	1116.0	1239.4	3433.6	3285.9	3581.2
##	1424.2	1750.644	2453.2	2304.4	2630.8	4849.6	4459.2	5240.0
##	0.0	0.000	81.0	81.0	81.0	178.9	178.9	178.9
##	0.0	0.000	29.0	29.0	29.0	54.6	54.6	54.6
##	0.0	0.000	90.0	90.0	90.0	169.3	169.3	169.3
##	101.2	114.005	66.5	61.0	73.8	329.2	305.1	353.4
##	309.6	315.128	55.4	52.9	58.4	692.4	681.9	702.8
##	667.6	670.657	64.8	63.3	66.4	1381.0	1375.2	1386.7
##	1096.3	1101.673	131.9	129.3	134.7	2316.0	2305.9	2326.1
##	1062.3	1068.745	128.4	125.3	131.7	2246.4	2234.3	2258.5
##	1025.4	1046.972	261.2	251.0	272.6	2442.1	2401.4	2482.7
##	0.0	0.000	1374.0	1374.0	1374.0	2013.8	2013.8	2013.8
##	0.0	0.000	483.0	483.0	483.0	685.5	685.5	685.5
##	134.4	400.721	1006.6	896.3	1162.6	1840.8	1462.8	2218.7
##	1237.5	1385.328	1313.9	1245.7	1393.5	3734.0	3524.2	3943.8
##	3868.1	3940.394	1513.5	1478.6	1550.9	7690.9	7588.3	7793.4
##	7970.6	8068.236	2527.2	2479.8	2577.4	14970.2	14831.6	15108.8
##	13049.2	13187.726	4008.8	3941.3	4079.8	24310.2	24113.6	24506.8
##	13367.3	13616.197	5829.5	5708.8	5957.7	27426.9	27073.7	27780.1
##	14664.1	15352.883	12112.0	11781.1	12469.9	38509.7	37532.2	39487.3
##	-972.8	-4.001	428.4	279.0	1247.8	621.0	-1005.3	2247.4
##	0.0	0.000	85.0	85.0	85.0	182.2	182.2	182.2
##	0.0	0.000	152.0	152.0	152.0	285.1	285.1	285.1
##	23.4	99.343	184.0	155.7	231.6	428.1	300.6	555.6
##	227.5	253.102	147.1	135.9	161.5	653.0	610.1	696.0
##	462.1	474.899	146.1	140.1	152.9	1032.4	1010.9	1054.0
##	755.0	768.510	195.9	189.5	203.0	1608.3	1585.6	1630.9
##	714.8	731.037	209.6	202.0	218.2	1566.3	1539.1	1593.5
##	504.2	537.266	262.8	247.7	280.8	1317.8	1262.3	1373.4
##	0.0	0.000	92.0	92.0	92.0	117.6	117.6	117.6
##	0.0	0.000	21.0	21.0	21.0	21.2	21.2	21.2
##	0.0	0.000	79.0	79.0	79.0	106.9	106.9	106.9
##	0.0	0.000	92.0	92.0	92.0	92.8	92.8	92.8
##	0.0	0.000	117.0	117.0	117.0	118.0	118.0	118.0
##	76.9	99.585	81.6	72.4	95.1	173.4	150.6	196.2
##	87.1	110.827	89.8	80.2	103.9	192.6	168.7	216.5
##	46.9	86.478	96.4	81.5	121.1	169.4	129.5	209.3
##	78.3	91.049	60.3	55.0	67.7	147.2	134.4	160.0
##	0.0	0.000	21.0	21.0	21.0	40.5	40.5	40.5

##	0.0	0.000	10.0	10.0	10.0	11.8	11.8	11.8
##	0.0	0.000	39.0	39.0	39.0	56.2	56.2	56.2
##	0.0	0.000	33.0	33.0	33.0	42.2	42.2	42.2
##	40.9	55.338	40.2	34.7	49.1	106.1	89.1	123.1
##	105.6	123.726	56.6	49.3	67.4	203.9	182.6	225.3
##	182.8	203.159	82.4	73.8	94.2	326.5	302.4	350.5
##	213.3	253.590	143.1	126.4	166.7	447.9	400.4	495.4
##	146.5	245.311	300.2	260.7	359.5	596.4	479.9	712.9
##	0.0	0.000	91.0	91.0	91.0	171.3	171.3	171.3
##	0.0	0.000	28.0	28.0	28.0	44.6	44.6	44.6
##	0.0	0.000	45.0	45.0	45.0	90.8	90.8	90.8
##	0.0	0.000	58.0	58.0	58.0	107.6	107.6	107.6
##	-7303.8	0.386	143.2	82.6	7386.8	203.5	-11440.0	11847.1
##	53.6	84.258	79.6	67.7	98.4	242.3	193.5	291.1
##	0.0	0.000	219.0	219.0	219.0	349.1	349.1	349.1
##	0.0	0.000	245.0	245.0	245.0	409.8	409.8	409.8
##	0.0	0.000	294.0	294.0	294.0	471.5	471.5	471.5
##	0.0	0.000	302.0	302.0	302.0	581.2	581.2	581.2
##	0.0	0.000	91.0	91.0	91.0	137.0	137.0	137.0
##	0.0	0.000	219.0	219.0	219.0	354.1	354.1	354.1
##	184.1	220.483	157.0	141.5	177.9	544.9	490.2	599.7
##	550.5	571.684	181.1	171.3	192.5	1118.4	1086.5	1150.4
##	1104.4	1121.911	273.5	265.1	282.6	2087.9	2061.5	2114.3
##	1951.8	1983.203	477.8	462.8	494.2	3682.0	3634.7	3729.3
##	1792.7	1845.455	693.5	668.5	721.3	3784.4	3705.0	3863.7
##	1798.1	1947.209	1372.9	1303.8	1452.9	4893.8	4669.4	5118.2
##	-907.8	-17.529	461.7	308.5	1198.8	575.5	-650.6	1801.7
##	0.0	0.000	129.0	129.0	129.0	190.9	190.9	190.9
##	0.0	0.000	230.0	230.0	230.0	350.4	350.4	350.4
##	0.0	0.000	282.0	282.0	282.0	398.0	398.0	398.0
##	0.0	0.000	486.0	486.0	486.0	669.4	669.4	669.4
##	69.2	304.790	572.9	483.2	718.8	1085.4	760.8	1409.9
##	324.3	522.581	687.6	605.4	803.7	1553.7	1280.5	1826.8
##	373.5	621.274	1035.3	929.7	1177.5	2136.3	1795.0	2477.5
##	488.5	819.103	1457.2	1314.9	1645.5	2939.3	2483.9	3394.6
##	-426.3	9.403	260.5	181.6	617.3	363.0	-283.8	1009.8
##	0.0	0.000	69.0	69.0	69.0	102.4	102.4	102.4
##	0.0	0.000	153.0	153.0	153.0	275.2	275.2	275.2
##	0.0	0.000	208.0	208.0	208.0	328.4	328.4	328.4
##	138.4	184.546	156.2	137.5	183.6	478.0	409.6	546.5
##	212.1	283.385	268.2	238.6	309.9	774.9	669.1	880.8
##	377.6	444.918	328.2	299.1	366.4	1104.5	1004.6	1204.4
##	256.5	377.192	469.4	418.8	539.5	1181.7	1002.6	1360.8
##	353.9	532.758	750.1	674.2	853.1	1791.8	1526.3	2057.3
##	0.0	0.000	39.0	39.0	39.0	56.4	56.4	56.4
##	0.0	0.000	18.0	18.0	18.0	28.1	28.1	28.1
##	0.0	0.000	49.0	49.0	49.0	61.5	61.5	61.5
##	0.0	0.000	77.0	77.0	77.0	96.7	96.7	96.7
##	0.0	0.000	131.0	131.0	131.0	164.5	164.5	164.5
##	0.0	0.000	225.0	225.0	225.0	282.5	282.5	282.5
##	125.4	212.870	194.7	162.1	249.6	470.9	361.1	580.7
##	101.8	191.492	239.0	204.5	294.2	497.2	384.6	609.9
##	52.6	175.877	314.4	268.1	391.4	557.5	402.8	712.3
##	0.0	0.000	58.0	58.0	58.0	79.0	79.0	79.0

```

##      0.0      0.000    20.0    20.0    20.0    26.4    26.4    26.4
##      0.0      0.000    47.0    47.0    47.0    61.3    61.3    61.3
##      0.0      0.000    80.0    80.0    80.0    92.0    92.0    92.0
##     35.1    59.877    65.6    56.1    80.9   133.4   104.9   161.8
##    161.0   171.262    65.3    60.7    71.0   266.7   255.0   278.5
##   253.1   266.295   105.7    99.7   112.9   420.8   405.6   436.0
##   190.0   220.073   146.9   133.9   164.0   407.0   372.5   441.5
##   158.5   228.940   305.7   276.1   346.5   580.6   499.6   661.6
##      0.0      0.000   137.0   137.0   137.0   254.0   254.0   254.0
##      0.0      0.000    57.0    57.0    57.0    81.1    81.1    81.1
##      0.0      0.000   135.0   135.0   135.0   186.6   186.6   186.6
##     46.6    83.764   100.3    86.2   123.4   235.0   183.6   286.3
##     69.6   108.437   119.7   104.6   143.4   294.4   240.8   348.0
##    205.5   240.815   196.5   181.2   216.5   583.3   534.4   632.2
##   385.7   410.246   196.9   185.8   210.3   823.8   789.9   857.8
##   409.2   436.504   230.9   218.5   245.8   905.4   867.7   943.1
##   314.1   368.714   304.1   280.3   334.9   897.1   821.6   972.5
##  93247.9  115768.635 101555.9  95900.4  118421.1  289949.3  257564.7  322333.9

```

8.1.5 Proportion total excess deaths per region

```

## # A tibble: 25 x 3
## # Groups: Departamento [25]
##   Departamento     suma     prop
##   <chr>        <dbl>    <dbl>
## 1 AMAZONAS      374.    0.366
## 2 ANCASH        4594.   4.50
## 3 APURIMAC     -30.2  -0.0296
## 4 AREQUIPA      6011.   5.89
## 5 AYACUCHO       419.   0.410
## 6 CAJAMARCA     2683.   2.63
## 7 CALLAO         7838.   7.68
## 8 CUSCO          1331.   1.30
## 9 HUANCAVELICA  716.   0.701
## 10 HUANUCO       1592.   1.56
## 11 ICA           4197.   4.11
## 12 JUNIN          3558.   3.48
## 13 LA LIBERTAD   7517.   7.36
## 14 LAMBAYEQUE    8101.   7.93
## 15 LIMA          78366.  76.7
## 16 LORETO        4654.   4.56
## 17 MADRE DE DIOS 404.   0.396
## 18 MOQUEGUA      976.   0.956
## 19 PASCO          173.   0.170
## 20 PIURA         11512.  11.3
## 21 PUNO           2541.   2.49
## 22 SAN MARTIN    2447.   2.40
## 23 TACNA          599.   0.587
## 24 TUMBES         1039.   1.02
## 25 UCAYALI        2219.   2.17

```

8.1.6 Population 2020 per region

```

## # A tibble: 25 x 2
##   Departamento pop.INEI

```

```

##      <chr>      <dbl>
## 1 MADRE DE DIOS    153164
## 2 MOQUEGUA       189701
## 3 TUMBES        251363
## 4 PASCO         314677
## 5 TACNA         362331
## 6 AMAZONAS      427202
## 7 APURIMAC      467707
## 8 HUANCAVELICA  511794
## 9 UCAYALI       523086
## 10 AYACUCHO      725649
## # ... with 15 more rows

```

8.2 Appendix 2

A simple comparison of registered deaths in Peru for months April to June of 2020 and corresponding periods in 2017-19 estimates 36,322 excess deaths.⁴⁶ It does not include non-registered deaths or take improvements in data registration into account. Applying our approach to this time period gives an estimate of 35,461 (CI 95% 32,425 – 37,803) registered deaths which climbs to 49,648 (CI 95% 48,037- 51,034). A similar comparison between 1 January to 12 July 2020 and the corresponding periods in 2017-19 shows excess 46,863 deaths, compared to 50,534 (CI 95% 44,448- 55,582), applying our approach to the same period.⁴⁷ This second study includes 2,000 excess deaths before March 2020, which cannot be attributed to the COVID-19 pandemic. A study of Lima metropolitan region finds an excess of 20,093 non-violent deaths for the first 24 weeks of 2020.⁴⁸ This is above to our own estimate based on registered deaths in Lima over the same period: 14659 (CI 95% 13579 – 15609). Adding unregistered deaths increases this to 20149 (CI 95% 18839-21323).