

# The Impact of Covid-19 on Higher Age Mortality

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# Agenda

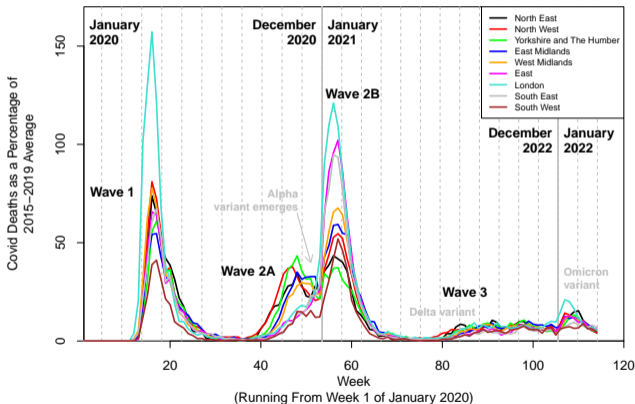
- Objectives
- Demographics of the Covid-19 victims
  - What is the relationship between Covid mortality and all-cause mortality?
  - What do we know about infection rates?
  - What has the impact of vaccination and new variants been?
- Demographics of the surviving population
  - The Accelerated Deaths Model
  - Adjusted (Post-Pandemic) Life Expectancy

Focus on English data.

But many conclusions will apply to other countries.

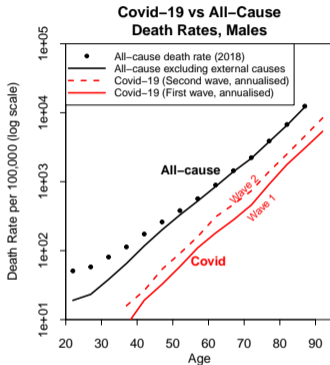
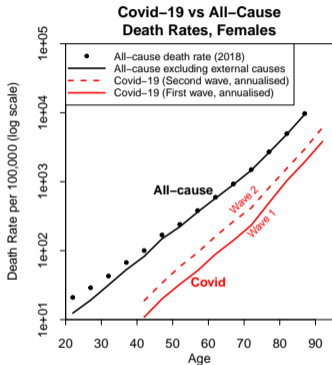
# Weekly Covid-19 Death Rates: 2020/21 by English Region

Weekly Deaths Involving Covid-19 By Region  
As a Percentage of All-Cause Deaths By Week  
(2015-2019, 5-Year Average)



- Considerable variation between the 9 English regions
- Wave 1:
  - London leads, but similar timing
  - Very different magnitudes
- Wave 2:
  - Wave 2A more focused in the northern regions
  - Wave 2B stronger in the south
- **London** Covid death rates 170% higher than the **South West**
- Urban/rural variation is even wider

# Covid-19 Death Rates By Age, Waves 1 and 2 (up to January 2021)



(Adapted from a David Spiegelhalter blog)

- Death rates are on a logarithmic scale
- All cause: with and without external causes
- Waves 1, 2 and 2018-all-cause are almost parallel!
- Waves 1 & 2: very similar age profile
- Conclusion: Covid death rates by age are approximately proportional to all-cause mortality (excluding external causes).

## Provisional Takeaway: The Proportionality Hypothesis

The comparison with all-cause death rates suggests the following way to look at Covid-19 mortality for age  $x$ :

$$\text{Covid Mortality Rate}(x) = \text{all-cause mortality rate}(x) \times \text{infection rate}(x) \times \text{relative frailty}(x)$$

- “Relative Frailty” measures the probability of death from Covid-19 (if infected) *relative to* the annual probability of death from all causes.
- The graphic suggests that  $\text{infection rate}(x) \times \text{relative frailty}(x)$  changes only slowly with age

## Generalising this concept by group

Group  $i$

$$\text{Covid Mortality Rate}(i, x) = \text{All-cause mortality rate}(i, x) \times \text{infection rate}(i, x) \times \text{relative frailty}(i, x)$$

where group  $i$  might be characterised by e.g.  
region; urban/rural; neighbourhood deprivation; ethnic group; ...

Hypothesis: **relative frailty( $i, x$ ) does not vary much by age or sub-group**

*i.e. differences in Covid-19 mortality between groups are largely due to differences in all-cause mortality and in infection rates*

How to verify if this is true?

$$\text{Covid Mortality Rate}(i, x) = \text{All-cause mortality rate}(i, x) \times \text{infection rate}(i, x) \\ \times \text{relative frailty}(i, x)$$

**Infection rates:** antibody prevalence data following the **first wave**  $\Rightarrow$

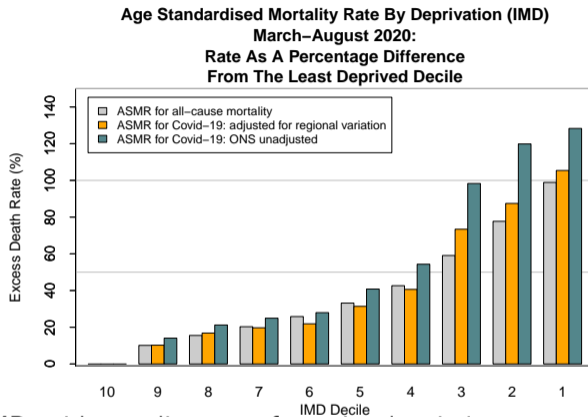
- About 6% of adults were infected
- Relatively little variation by age and sex
- Relatively little variation by deprivation (Index of Multiple Deprivation, IMD)
- Significant variation by region (e.g. London  $\gg$  South West)

$$\text{Covid Mortality Rate}(i, x) = \text{All-cause mortality rate}(i, x) \times \text{infection rate}(i, x) \\ \times \text{relative frailty}(i, x)$$

- Data: Age Standardised Mortality Rates (ASMR) by deprivation decile ( $i$ )
- Data: Age Standardised Mortality Rates by region
- But: ASMR by deprivation decile is distorted by regional effects (e.g. London more deprived and higher infection rates)
- ASMR by region allows us to adjust the ASMR by deprivation decile



# ASMRs by deprivation: Wave 1, Adjusted for Regional Variation



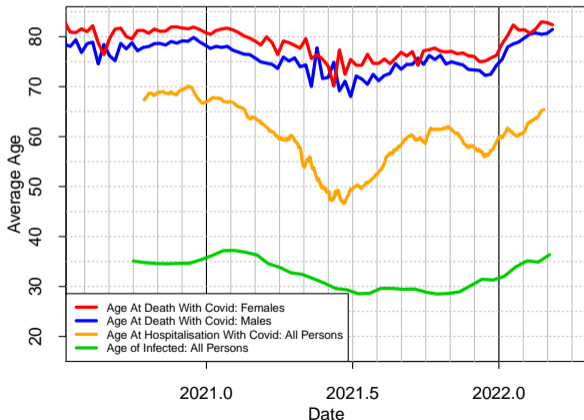
- **Blue bars:** ASMRs with no adjustment for regional variation
- **Orange bars:** ASMRs with the effect of regional variation filtered out
- Covid-19 ASMRs by decile are now approximately proportional to all-cause ASMRs
- Conclusion:  $\text{Relative Frailty}(i, x)$  varies very little across deprivation deciles  $i = 1, \dots, 10$

## 2021/22: The impact of vaccination and new variants

- Data for 2021: much more complex
- Vaccination: oldest ages in December 2020 → teenagers in October/November 2021
- New variants in the UK:
  - Alpha: December 2020
  - Delta: May/June 2021 (more infectious than Alpha)
  - Omicron BA.1: December 2021 (more infectious than Delta, less severe)
  - Omicron BA.2: February 2022 (more infectious than Omicron BA.1)
- What have been the impacts at different ages on:
  - Infection rates
  - Hospital admissions
  - Deaths

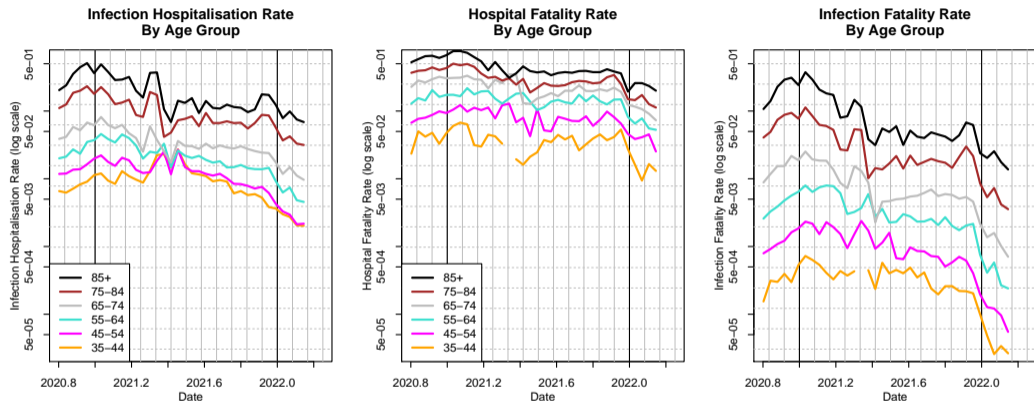
# Average Age At Death With Covid and Average Age At Hospitalisation With Covid

Average Age of Covid-19 Victims



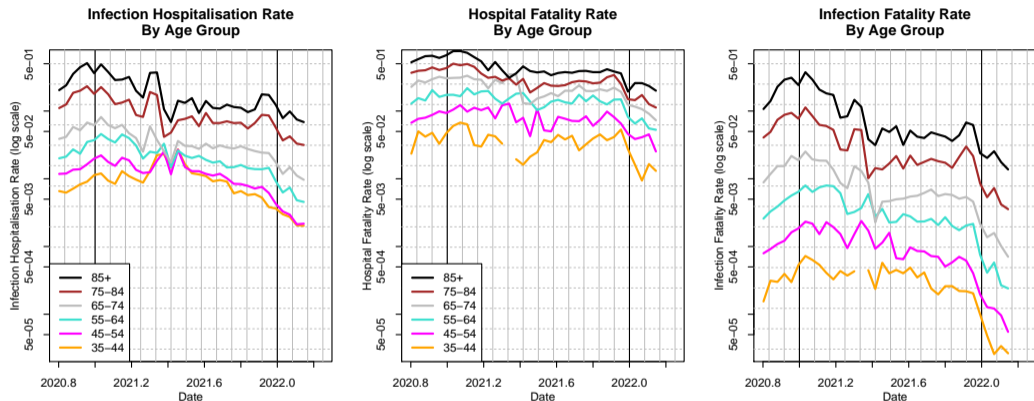
- Pre-covid seasonal variation: around 1 year higher in winter
- End 2020 to mid 2021:
  - Avg Age at Death drops by 7 to 9 years
  - Avg Age at Hospitalisation drops by 20 years
- Due to:
  - vaccination by age group
  - age-related behaviour

# Estimated infection fatality rates and related quantities



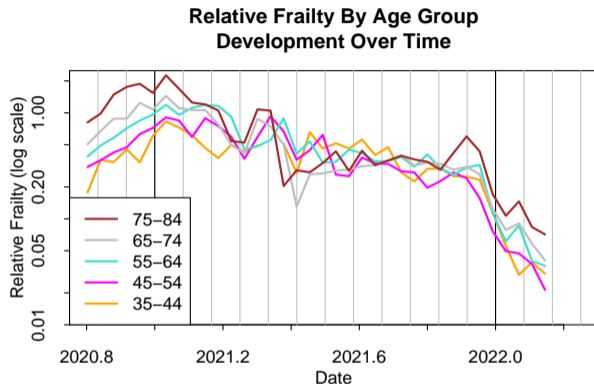
- Infection Hospitalisation Rate = Proportion of newly infected who are admitted to hospital
- Hospital Fatality Rate = Proportion of newly hospitalised who die from Covid
- Infection Fatality Rate = Proportion of newly infected who die from Covid

# Estimated infection fatality rates and related quantities



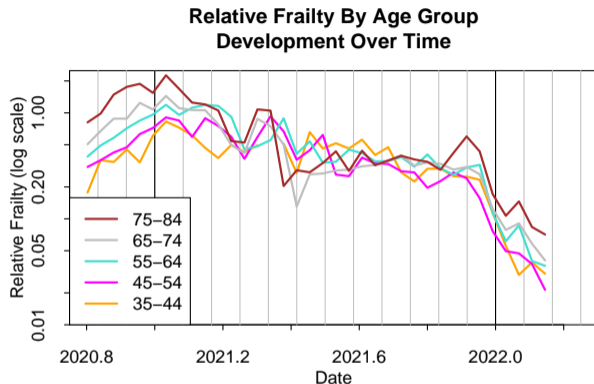
- Infection → Hospitalisation: strong benefit from vaccination
- Hospital → Dead: some vaccination effect, but weaker
- Infection Fatality Rate: declines sooner for the older groups due to vaccination
- Infection Fatality Rate: booster + Omicron ⇒ big decline

# Estimated relatively frailty



- Relative Frailty  $(t, x) := \text{Infection Fatality Rate}(t, x) \div \text{All-Cause Death Rate}(2018, x)$
- All-cause Death Rate excludes external causes
- Steady decline as vaccines take effect and new variants replace older variants

## Estimated relatively frailty: further remarks



- Pre-vaccination (end of 2020): relative frailty has some variation with age (increases by 3-4% per year of age)
- Vaccination causes the age gradient to disappear in mid 2021
- Age gradient re-emerges in 2022 after booster roll out to all age groups

# The Impact of Covid-19 on Future Mortality

Preceding discussion:

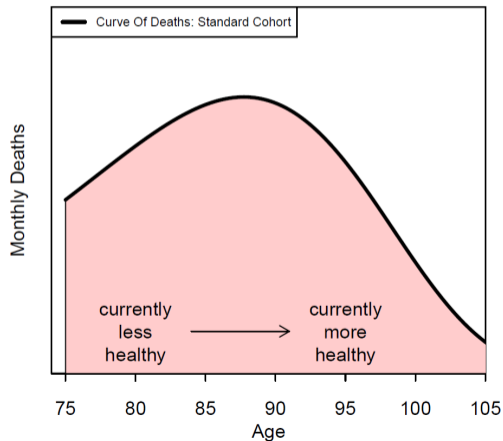
People of the same age who are more “frail” are more likely to die if they become infected with Covid-19.

What is the impact on the surviving population?



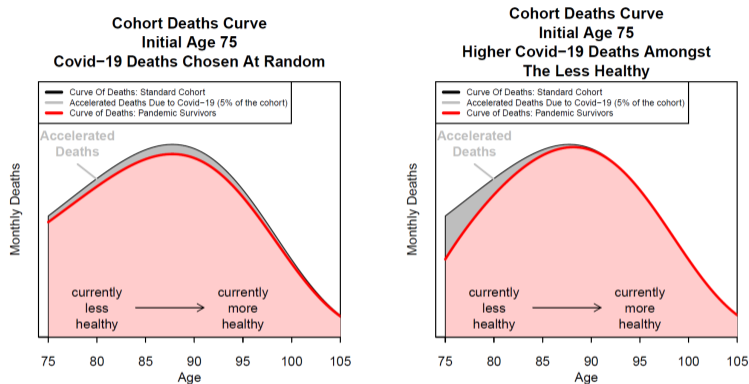
# Pre-Covid: Cohort Curve of Deaths

## Cohort Deaths Curve Initial Age 75 Before Covid-19



- For a cohort currently aged 75: what will be the ages at death?
- Less healthy now  $\Rightarrow$  more likely to die earlier

# Impact of Covid-19 on the Curve of Deaths



- A (left): Covid victims randomly chosen from the cohort
- B (right): Covid deaths more prevalent amongst the less healthy

Scenario B is consistent with the empirical evidence that those with co-morbidities are more likely to die if they get infected

# The Accelerated Deaths Model

Example: Consider a cohort currently aged  $x$  (e.g. 75)

- Initial cohort size: 100,000
- $d(t, x) =$  pre-Covid curve of deaths,  $t = 0, 1, 2, \dots$
- Out of the original  $d(t, x)$  “scheduled” to die at  $t$   
a proportion  $\pi(t, x)$  die in the short term from Covid

## The Accelerated Deaths Model (cont.)

- Simple starting point:

$$\pi(t, x) = \alpha(x)R(x)\exp[-t/\rho(x)]$$

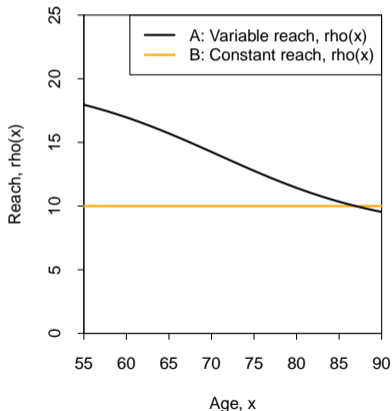
- $\alpha(x)$  = “amplitude”  $\Rightarrow$   
this determines the proportion of the entire cohort who die from Covid
- $\rho(x)$  = “reach”  $\Rightarrow$   
links to the years-of-life-lost (YLL) by those who die from Covid
- $R(x)$  = normalising const. depending on  $\rho(x)$  and the shape of  $d(t, x)$

$$R(x) = d(0, x) / \int_0^{\infty} d(t, x) \exp[-t/\rho(x)] dt$$

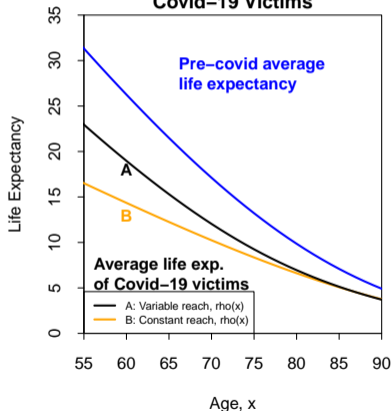
- $R(x)$  definition  $\Rightarrow$   
 $\alpha(x)$  = infection rate  $\times$  relative frailty  
covid mortality(x) = all cause mortality(x)  $\times$  infection rate (x)  $\times$  relative frailty(x)

# Calibrating the reach parameter, $\rho(x)$

Reach,  $\rho(x)$ , As A Function of Age

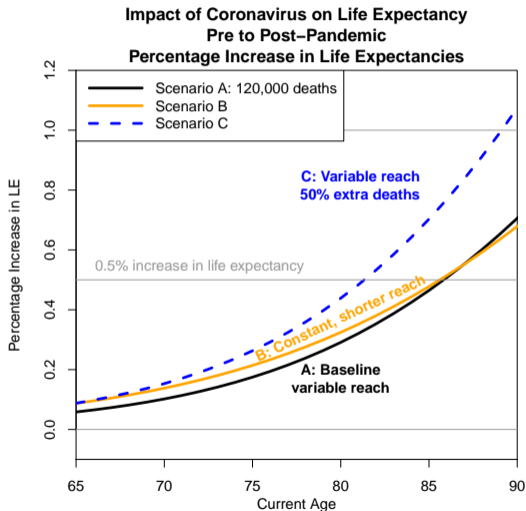


Avg. Pre-Covid Life Expectancy versus Avg. Life Expectancy of Covid-19 Victims



- The shape of  $\rho(x)$  depends on variability in underlying frailty
- Scenario A: (experimental) reach:  $\sim 18$  (young) to  $\sim 10$  (old)
- Scenario B: (extreme) reach = 10 constant
- B is simple but not very plausible

# Adjusted (Post-Pandemic) Life Expectancy (APPLE)



- More realistic scenarios in terms of total Covid-19 deaths
- $LE(\text{pre-covid}) \rightarrow LE(\text{survivors})$
- What is the percentage increase?
- Scenarios:
  - A: 120,000 deaths + variable reach
  - B: 120,000 deaths + constant reach
  - C: 180,000 deaths + variable reach
- Age 65: APPLE of healthier survivors is less than 0.1% higher than pre-Covid cohort life expectancy
- Impact assumes no secondary effects e.g. no long-term impairments  $\Rightarrow$  further data and modelling

## What are the other secondary effects beyond this model?

- Non-Covid illnesses (e.g. late cancer diagnosis or delayed treatment)
- More extreme forms of “Long Covid”  
Covid survivors might have long-term health impairments
- Lasting impact of innovation during the pandemic
- Behavioural changes (positive and negative)
- Impact of increased long-term unemployment
- Economic impact on future health spending and research

Some secondary effects might be observable in 2020/21 cause of death data

- Higher cancer death rates in 2021, 2022, ...
- Potentially lower death rates in 2021 from e.g. respiratory diseases  
(due to accelerated death from Covid-19 in 2020)

# Conclusions and Lessons Learned

- 1 Strong relationship between covid mortality( $i, x$ ) and all-cause mortality( $i, x$ )
  - contrasts with Spanish Flu: younger affected much more; some prior immunity
  - Covid-19: novel  $\Rightarrow$  no prior immunity
- 2 Significant variation by region and urban-rural  $\Rightarrow$  much more than a normal year
- 3 In the absence of “secondary effects”, the impact of the pandemic on the life expectancy of survivors is likely to be small
- 4 We will need time to understand the nature and magnitude of secondary effects

Thank you

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