

On Exit Strategy from Covid-19 Lockdown

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Three main models for handling the Covid-19 outbreak:

1 Full lockdown

2 Precision quarantine- Isolate positive cases and their immediate contacts using “contact tracing” technologies, until a vaccine/cure is available.

3 Herd-immunity: No need to wait for a vaccine; but, high mortality and health system might not contain the outburst

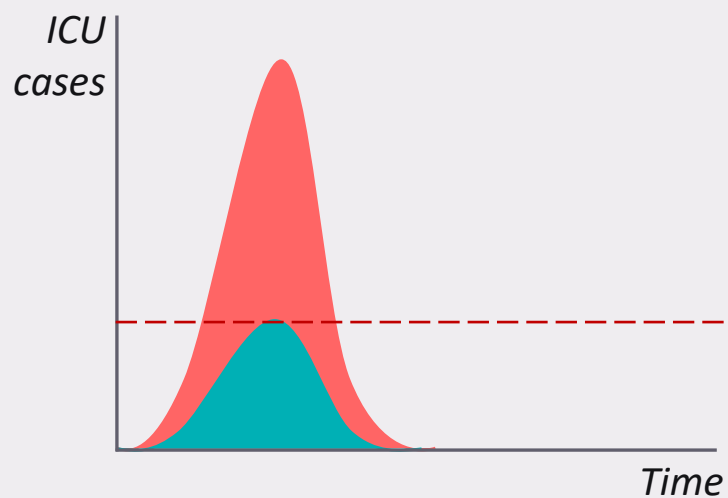
We propose a combination of both models:

Managed herd-immunity with the RISK-BASED MODEL (1+2+3)

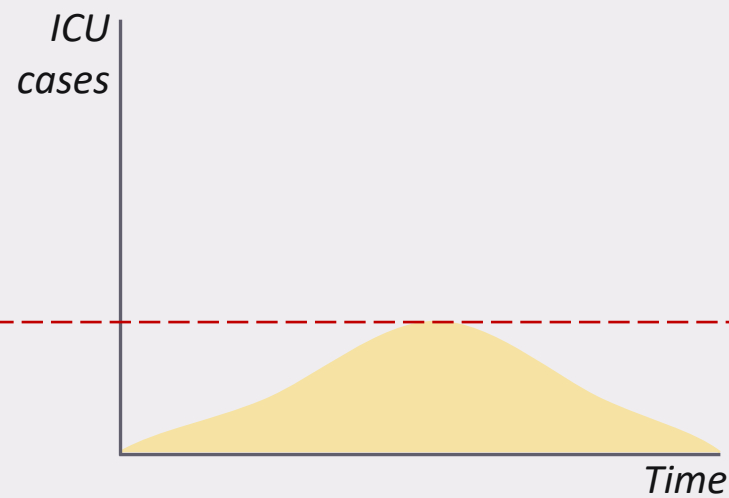
- High-risk group → **1**
 - Low-risk group → **3**
- Second phase → High-risk group is gradually released **3+2**
-
- The diagram illustrates a risk-based exit strategy. It shows two rows of text: 'High-risk group' and 'Low-risk group'. An arrow points from 'High-risk group' to a large teal number '1'. Another arrow points from 'Low-risk group' to a large teal number '3'. A bracket connects the '1' and '3' to a horizontal line. From the end of this horizontal line, an arrow points to the right, labeled 'Second phase'. To the right of this arrow, the text reads 'High-risk group is gradually released' followed by a large teal '3+2'.

The Risk-based Model

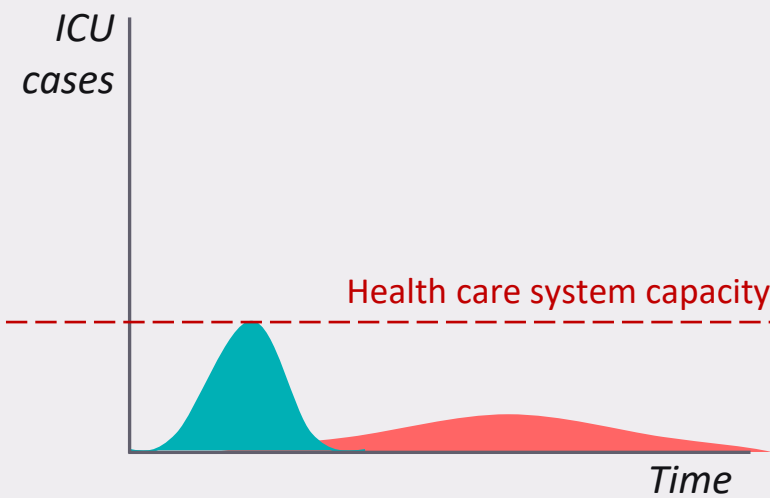
Without any safety measures



Flattening the curve for the entire population



Risk-based model



■ Entire population

■ Low-risk population

■ High-risk population

Questions of Interest

- **Define “high-risk” group: age and pre-existing conditions cut-off**
 - + How to determine without overfitting
- **Can the health system capacity (say, ICU beds) contain the number of severe cases among the low-risk group?**
- **Beyond obvious benefits to the economy, is the risk-based model safer in terms of overall mortality?**

Worst-case Analysis

- **The problem:** Covid-19 dynamics (spread and duration) is too complicated at this stage
- **The solution:** adopting worst-case analysis under reasonable assumptions

Worst-case analysis under reasonable assumptions

- Assumptions:
 - + Upper-bound on the time from infection → ICU care- 1 weeks*
 - + The probability of a low-risk person to be infected and need a critical-bed care is fixed.
 - Why: low-risk group release can be done under social distancing restrictions (to avoid viral load)
- Worst-case:
 - + Infection rate among the low-risk group will be 100%
 - + All severe cases among the low-risk group will need ICU **simultaneously**

*research shows the average time is 5-7 days

- The analysis below focuses on the low-risk population.
- m = size of the low-risk population
- b = budget of ICUs

$$\mathbb{P}[\text{severe}] = \mathbb{P}[\text{severe} \wedge \text{infected}] = \underbrace{\mathbb{P}[\text{infected}]}_{\leq 1} \underbrace{\mathbb{P}[\text{severe} \mid \text{infected}]}_{:=\nu}$$

Therefore, we need:

$$b \geq m \nu .$$

- Our goal is to derive an **upper bound** on ν

Upper bounding ν

$$\nu := \mathbb{P}[\text{severe} \mid \text{infected}] \approx \frac{\# \text{ severe today}}{\# \text{ infected today}}$$

- **We know** $k := \# \text{ severe today}$
(and to be on the safe size, “today” should be in the time interval $[0, \text{one week}]$)
- **We don't know** $\# \text{ infected today}$
- Let $p^* = \text{probability to be infected today}$, then

$$\nu \approx \frac{k}{\# \text{ infected today}} = \frac{k}{p^* m}$$

- So, to prove $b \geq m \nu$ we need

$$b \geq m \nu \approx \frac{k}{p^*}$$

- We now need a **lower bound** on p^*

Lower bounding p^*

- By sampling n i.i.d. persons from the low-risk population and finding S_n cases, we have $p^* \approx \frac{S_n}{n}$

- Overall:

$$b \geq m \nu \approx \frac{k}{p^*} \approx \frac{k}{S_n/n}$$

- Example: in Israel, we have $k = 15$ and we estimate $p^* = 0.02$ so we need $b \geq 750$
- But, all of the above involved approximations. We need **bounds** that hold with sufficient probability!

Tail bounds

We derive concrete bounds based on the following techniques (and some tricks):

- **Bernstein's inequality:** If $S_n \sim \text{Binom}(p, n)$ then

$$\mathbb{P}[S_n - np > t] \leq e^{-\frac{t^2/2}{np(1-p)+t/3}}$$

- **Zubkov and Serov 2013:**

$$\mathbb{P}[S_n \leq k] \leq \Phi\left(-\sqrt{2nD_{KL}(p, (k+1)/n)}\right)$$

where Φ is the cumulative distribution function of a standard normal variable and D_{KL} is the KL-divergence

Recall, we need

$$b \geq m \nu$$

For sufficiently large sample size n , with high probability we have

$$m \nu \geq 1.92 \frac{k}{S_n/n} .$$

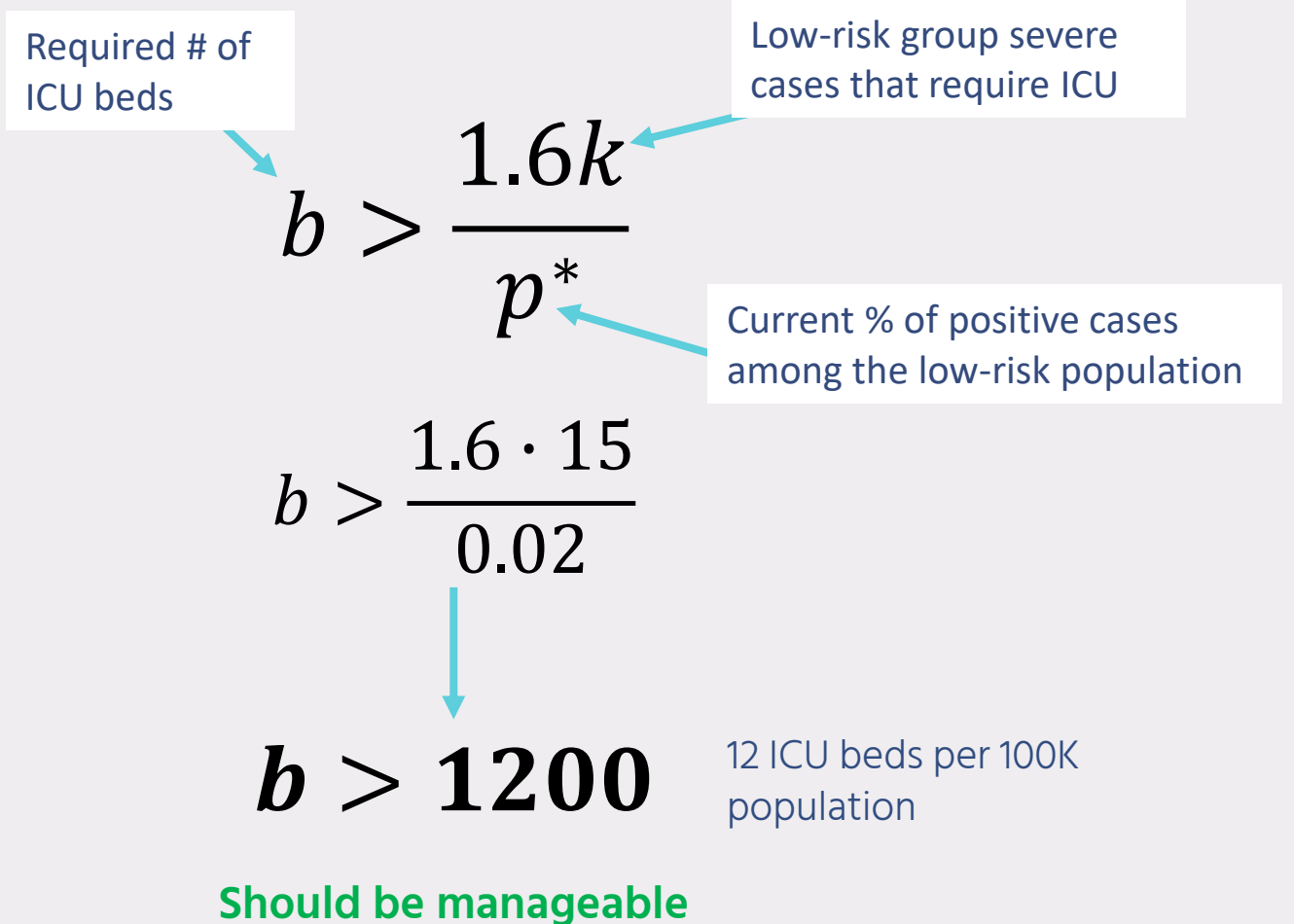
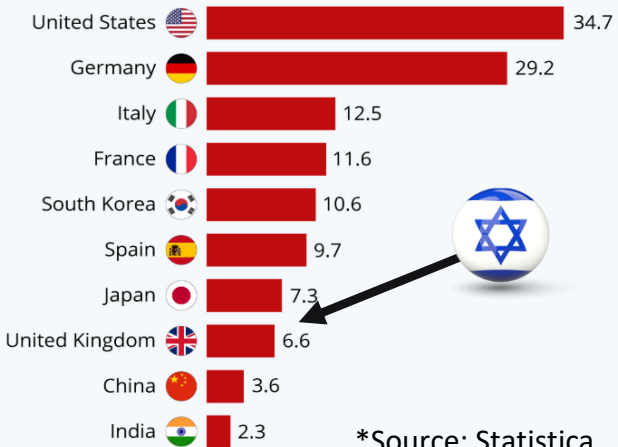
Israel as a Case Study

As of 3/29/20:

- $k \leq 15$
- p^* indications
 - 1%- pre-outburst situation in Iceland
 - 1.8%- medical crew sampling in Jerusalem
 - 8%- Partners Health, Boston
 - We assume 2%
- p^* can be estimated with a relatively small i.i.d. sampling

The Countries With The Most Critical Care Beds Per Capita

Total number of critical care beds per 100,000 inhabitants in selected countries*



Summary

Risk-based model

- High infection rate among low-risk group → **low mortality**
- Low infection rate among the high-risk group thanks to herd-immunity of the rest

- Minor negative effect on the economy

- Clear exit point- result in economic stability and higher civilian cooperation

- Short duration and controlled effect on the health care system

Precision quarantine

- Unified infection rate among the entire population → **high mortality** among high-risk group

- Extensive negative effect on the economy

- Lack of visibility- result in economic instability and civilian despair

- Devastating and prolonged effect on the health care system