

# Suppressing Covid-19: Public Health Policy and Effective Mass-Testing Rates

Division F Conference  
Department of Engineering, U of Cambridge

September 17, 2020

**Ioannis Kontoyiannis**  
*Statistical Laboratory, U of Cambridge*



## Joint work with



Jussi Taipale  
Biochemistry  
U of Cambridge



Sten Linnarsson  
Medical Biochemistry & Biophysics  
Karolinska Institutet

# Outline

## △ Covid-19 tests

- ▷ Antibody tests
- ▷ PCR tests
- ▷ Antigen tests

~→ Rapid, cheap, at-home tests

## △ Epidemiological models

- ▷ SIR models on random population networks
  - ▶ Erdős-Rényi graphs
  - ▶ Random graphs with given degree distribution
- ▷ SIR epidemics with mass testing

## △ Necessary testing rates for suppression

- ▷ Rigorous results for a broad class of models
- ▷ Explicit numerical examples

# Outline

## △ Covid-19 tests

- ▷ Antibody tests

- ▷ PCR tests

- ▷ Antigen tests

  - ~> Rapid, cheap, at-home tests

## △ Epidemiological models

- ▷ SIR models on random population networks

  - ▶ Erdős-Rényi graphs

  - ▶ Random graphs with given degree distribution

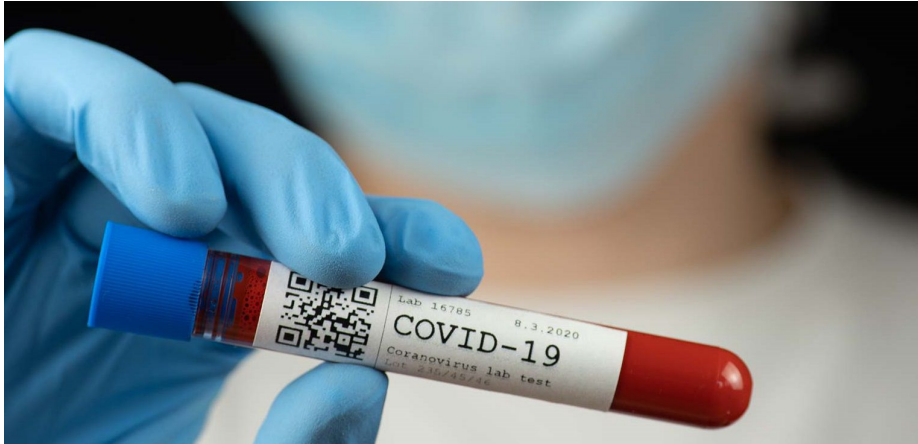
- ▷ SIR epidemics with mass testing

## △ Necessary testing rates for suppression

- ▷ Rigorous results for a broad class of models

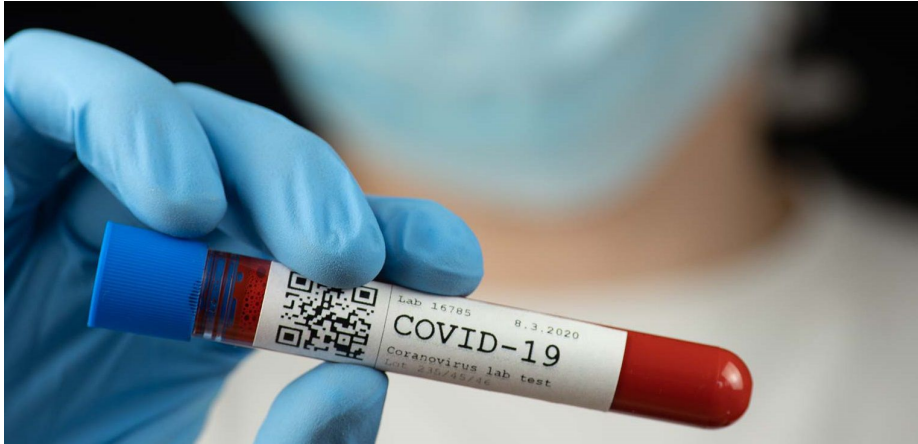
- ▷ Explicit numerical examples

## Covid-19 testing: Why?



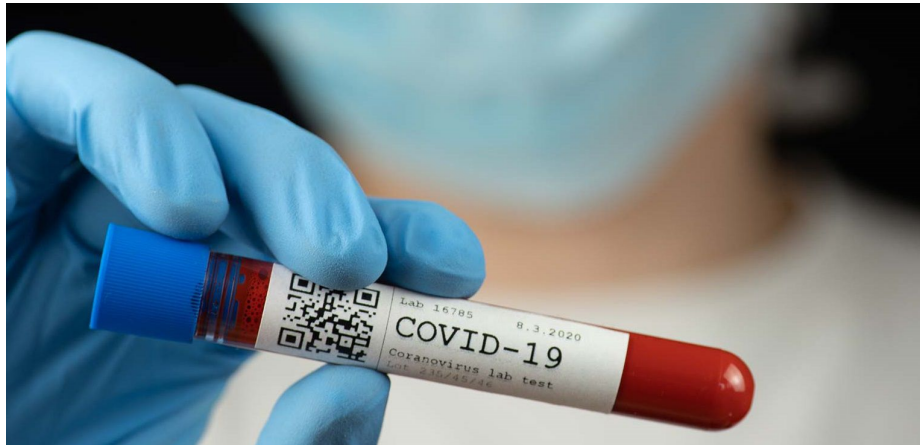
- Most basic task in infectious **disease control**: Identify and isolate infected individuals

## Covid-19 testing: Why?



- ▶ Most basic task in infectious **disease control**: Identify and isolate infected individuals
- ▶ Large-scale testing **assuages the fear** that threatens the economy and public life

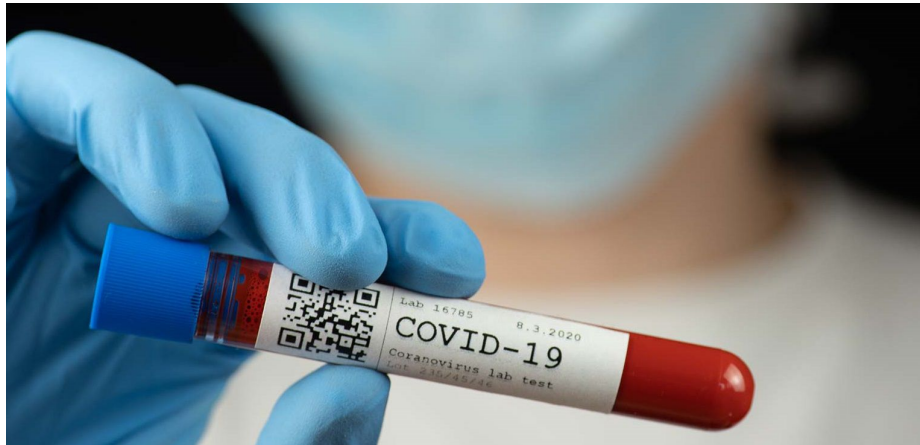
## Covid-19 testing: Why?



- ▶ Most basic task in infectious **disease control**: Identify and isolate infected individuals
- ▶ Large-scale testing **assuages the fear** that threatens the economy and public life
- ▶ Accurate and timely test results inform our **models, predictions, public health policy**, and our **understanding of the disease**



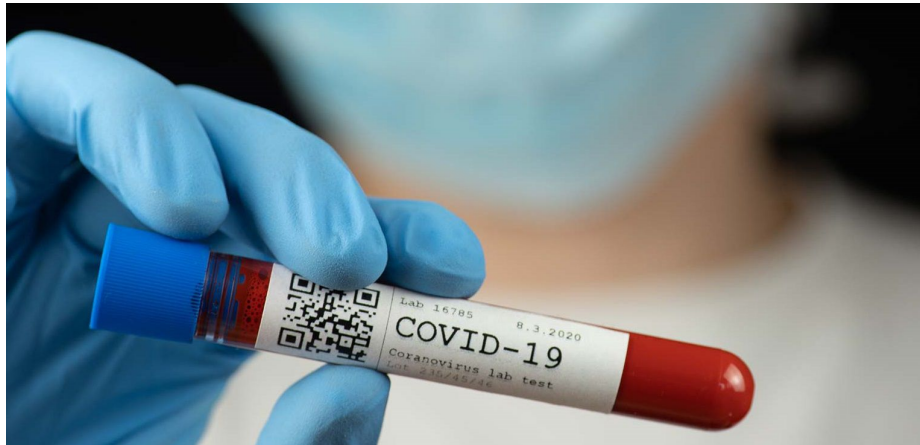
# Covid-19 testing: Why?



- ▶ Most basic task in infectious **disease control**: Identify and isolate infected individuals
- ▶ Large-scale testing **assuages the fear** that threatens the economy and public life
- ▶ Accurate and timely test results inform our **models, predictions, public health policy**, and our **understanding of the disease**
- ▶ **Asymptomatic spreaders** can remain infectious for weeks and they account for 20-70% of all infections

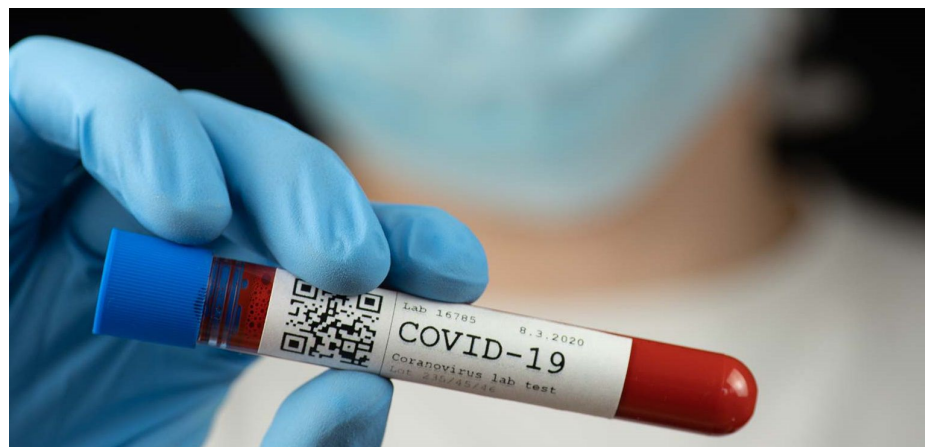


# Covid-19 testing: Why?



- ▶ Most basic task in infectious **disease control**: Identify and isolate infected individuals
- ▶ Large-scale testing **assuages the fear** that threatens the economy and public life
- ▶ Accurate and timely test results inform our **models, predictions, public health policy**, and our **understanding of the disease**
- ▶ **Asymptomatic spreaders** can remain infectious for weeks and they account for 20-70% of all infections
  - ↪ We need a **very large number of tests** in order to control the pandemic before a vaccine becomes widely available

# Covid-19 testing: Why?



- ▶ Most basic task in infectious **disease control**: Identify and isolate infected individuals
- ▶ Large-scale testing **assuages the fear** that threatens the economy and public life
- ▶ Accurate and timely test results inform our **models, predictions, public health policy**, and our **understanding of the disease**
- ▶ **Asymptomatic spreaders** can remain infectious for weeks and they account for 20-70% of all infections
  - ↪ We need a **very large number of tests** in order to control the pandemic before a vaccine becomes widely available

**Currently:**  $\approx$  200 Covid-19 FDA-approved tests in the US (+many more in development) using different technologies . . .

# Antibody tests

## Purpose

Detect previously infected  
and currently immune individuals  
⇒ Not relevant for our purposes



# Antibody tests

## Purpose

Detect previously infected  
and currently immune individuals  
⇒ Not relevant for our purposes



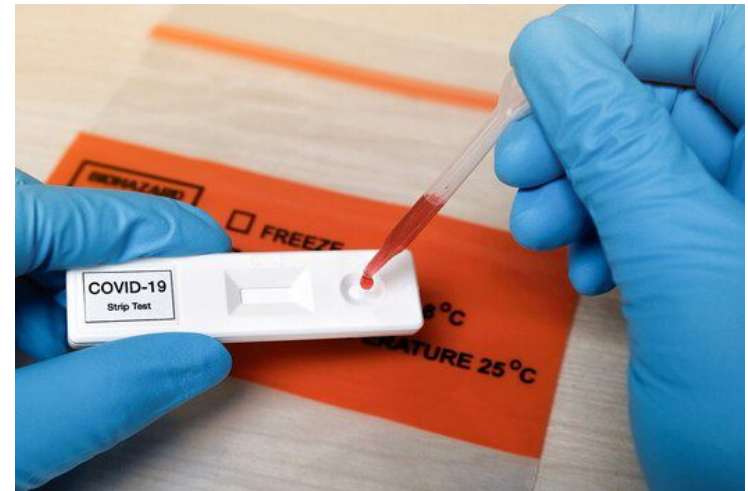
## Problems

- ▷ A lot of unknowns
  - ~> Do antibodies confer immunity? For how long?
  - ~> Relationship between antibody level (titer) and degree and persistence of immunity?
  - ~> Antibody test accuracy?
- ▷ More than 90 tests on the US market without FDA review
- ▷ Inappropriate use of results as “immunity passports”

# Antibody tests

## Purpose

Detect previously infected  
and currently immune individuals  
⇒ Not relevant for our purposes



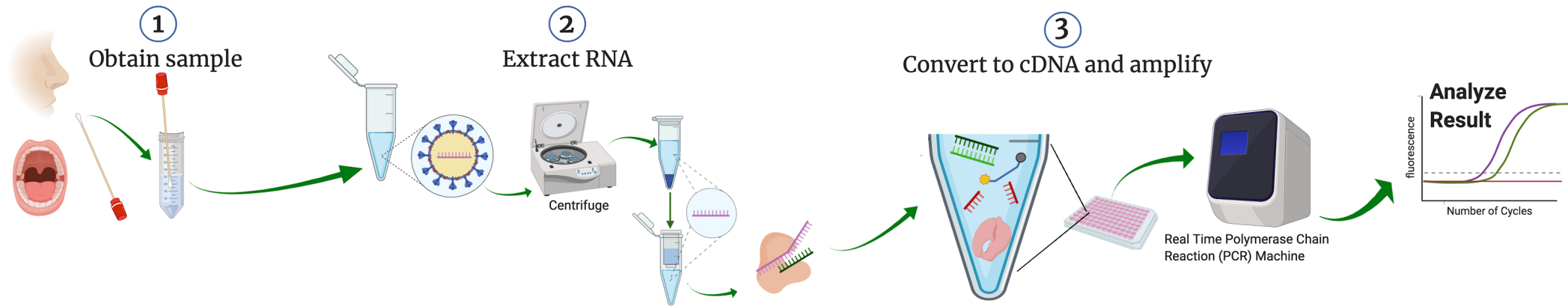
## Problems

- ▷ A lot of unknowns
  - ~> Do antibodies confer immunity? For how long?
  - ~> Relationship between antibody level (titer) and degree and persistence of immunity?
  - ~> Antibody test accuracy?
- ▷ More than 90 tests on the US market without FDA review
- ▷ Inappropriate use of results as “immunity passports”

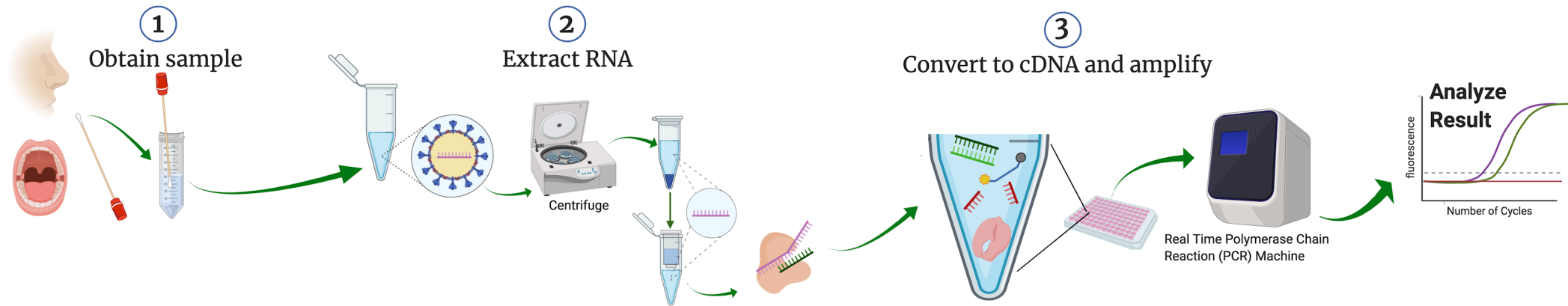
## But

- ▷ Important and possibly becoming much more common soon

# PCR tests [reverse-transcription polymerase chain reaction tests]

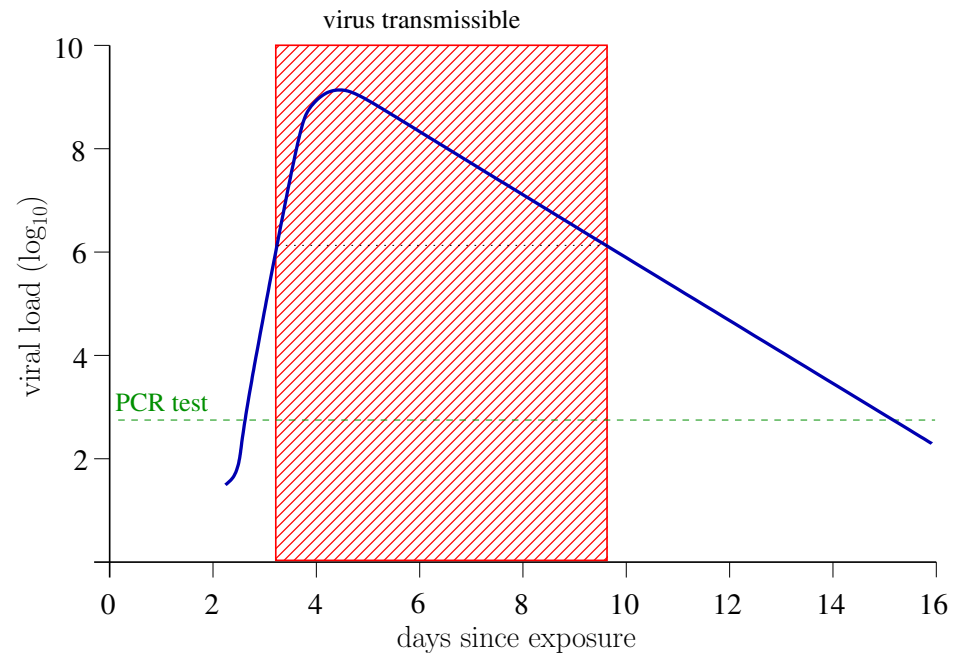


# PCR tests [reverse-transcription polymerase chain reaction tests]



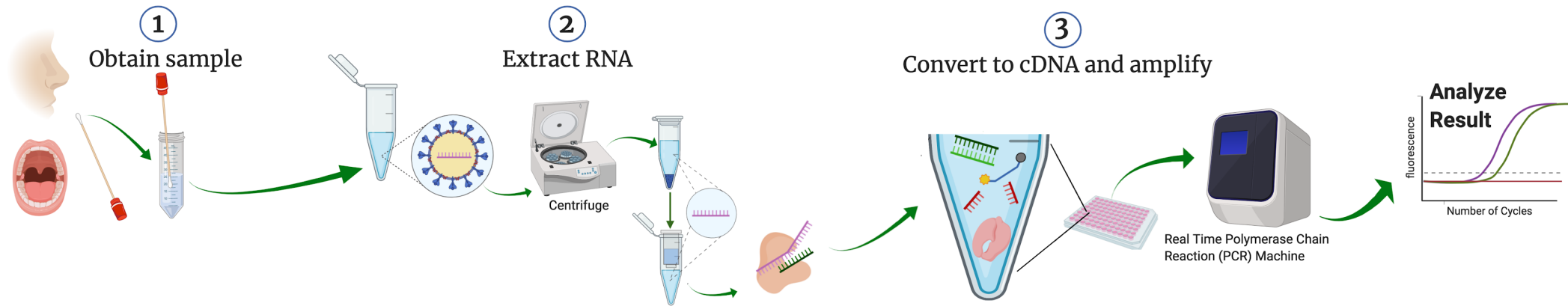
## The gold standard

- ▷ They take only a few hours
- ▷ Detect as few as 100-1000 copies of viral RNA in 1ml of sample
- ▷ Sensitivity close to 100%



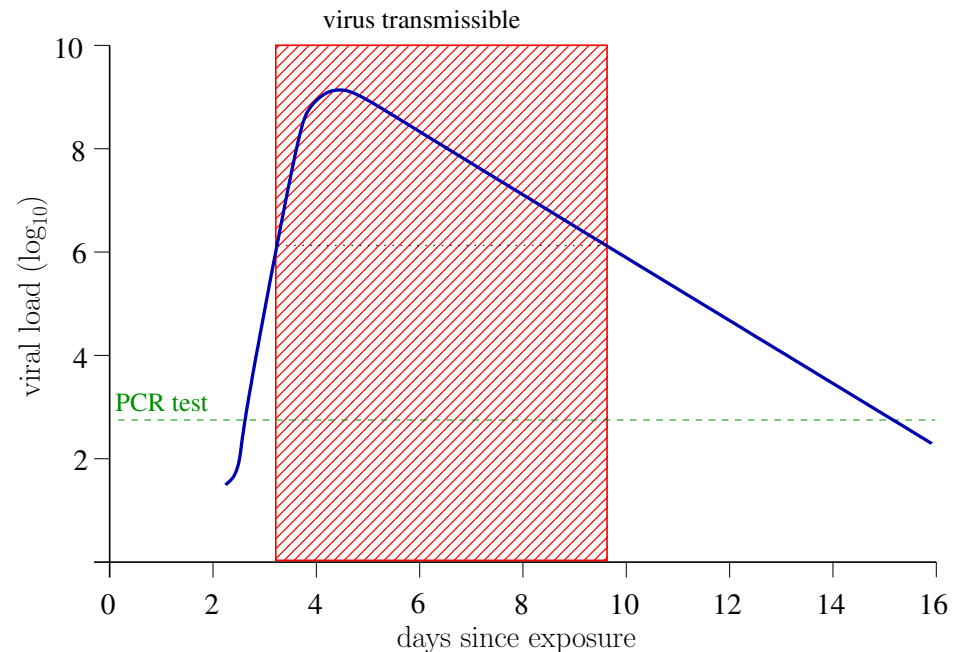


# PCR tests [reverse-transcription polymerase chain reaction tests]



## The gold standard

- ▷ They take only a few hours
- ▷ Detect as few as 100-1000 copies of viral RNA in 1ml of sample
- ▷ Sensitivity close to 100%



- ▷ Indeed, **the** FDA test standard
- ▷ **But:** Not designed for an out-of-control pandemic

# PCR tests

## Problems

- ▷ Very **slow** turnaround times (3-14 days in most places in the US)  
    ~> they miss the most infectious period
- ▷ **Not enough** of them: Bottlenecks:  
    Chemical reagents, lab supplies, PCR machines
- ▷ **Expensive** [\$35-200], very tightly regulated, require specialized personnel and equipment [⇒ social inequality issues]
  - ▷ [Ct values not reported]

# PCR tests

## Problems

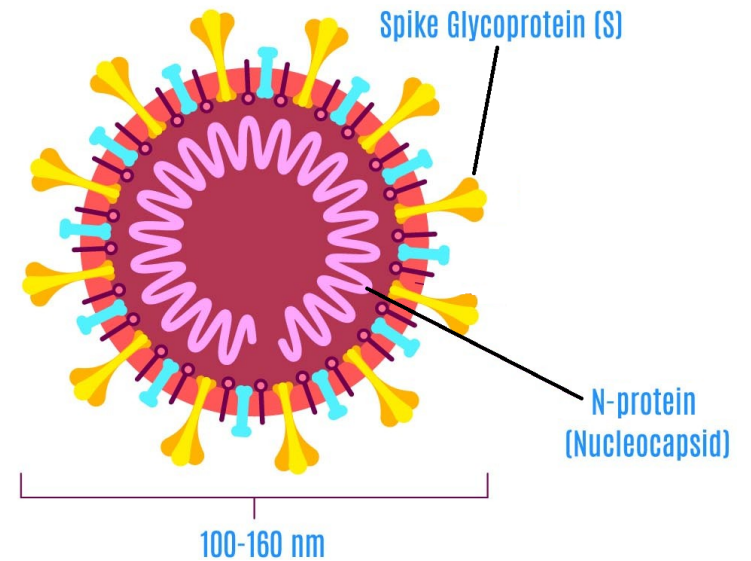
- ▷ Very **slow** turnaround times (3-14 days in most places in the US)  
    ~> they miss the most infectious period
- ▷ **Not enough** of them: Bottlenecks:  
    Chemical reagents, lab supplies, PCR machines
- ▷ **Expensive** [\$35-200], very tightly regulated, require specialized personnel and equipment [⇒ social inequality issues]
  - ▷ [Ct values not reported]

## Partial fixes

- ▷ Pooling or “**group testing**”: Still expensive and not fast enough
- ▷ Saliva-based tests with Ginkgo Bioworks’s **Illumina sequencing machines** instead of PCR. Factor of 6 faster, still slow:  
    Samples must be shipped centralized locations
- ▷ Saliva-based modified-PCR laboratory tests: UIUC’s **I-COVID**, Yale’s **SalivaDirect**. Results in 2-6 hours, cost \$10-20.  
    UIUC story highlights the need for even more, cheaper tests

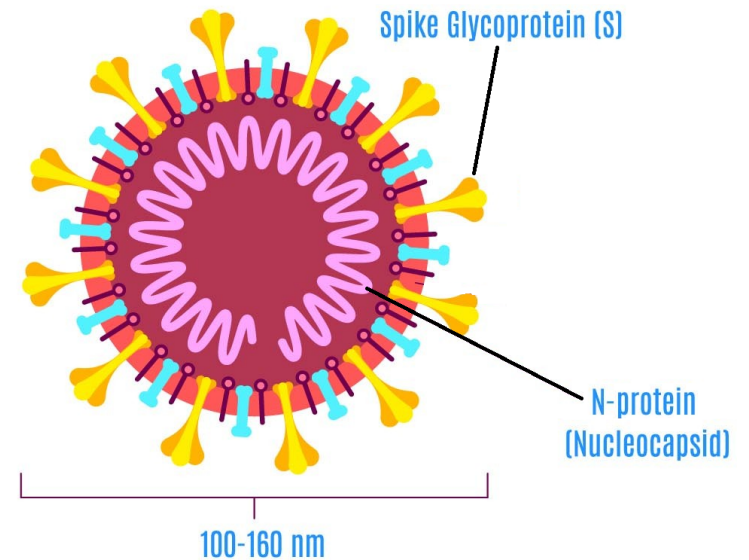
## Antigen tests: I lab-based

They do not identify virus RNA  
but an **antigen**, typically a protein



## Antigen tests: I lab-based

They do not identify virus RNA  
but an **antigen**, typically a protein



- E.g., tests made by *Quidel* and *Becton-Dickinson* (US) detect the **nucleocapsid** (N) protein in nasal/throat swab samples

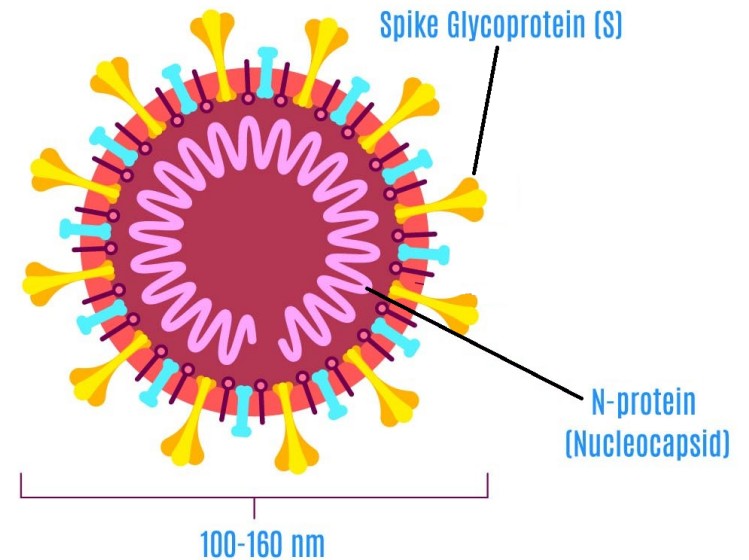
Tests cost is  $\approx$  half of PCR, give results in 15 mins

Can be administered at a point-of-care location

Will be making 14 million tests/month by end of September

## Antigen tests: I lab-based

They do not identify virus RNA  
but an **antigen**, typically a protein



- ▶ E.g., tests made by *Quidel* and *Becton-Dickinson* (US) detect the **nucleocapsid** (N) protein in nasal/throat swab samples

Tests cost is  $\approx$  half of PCR, give results in 15 mins

Can be administered at a point-of-care location

Will be making 14 million tests/month by end of September

### Drawbacks

- ▷ They only work with a proprietary reader  
but companies cannot produce it at same scale
- ▷ Since nucleocapsid is *inside* the virus  
they need reagents to break down its outer membrane

## Antigen tests: II rapid tests

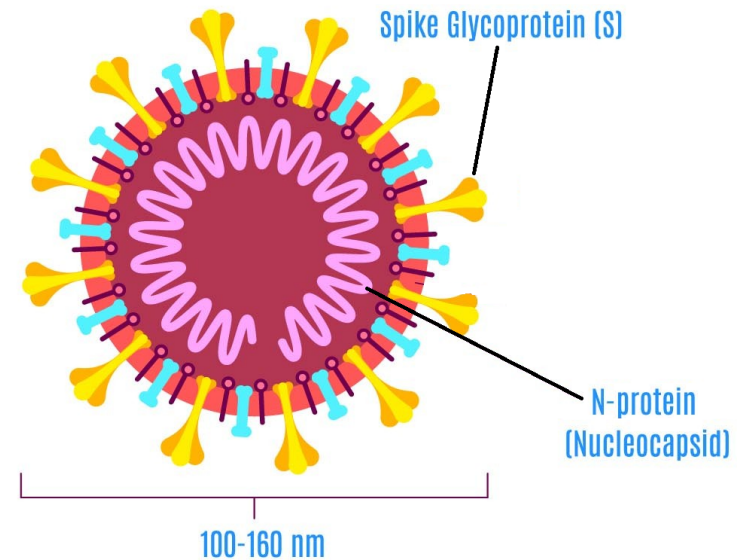
Made by US-based companies including e25 Bio, Sona Nanotech, Icen Diagnostics, OraSure

### Characteristics

Cheap: \$1-2

Results in 15 minutes

Home tests, no equipment: Saliva + saline solution + small cup





## Antigen tests: II rapid tests

Made by US-based companies including e25 Bio, Sona Nanotech, Icen Diagnostics, OraSure

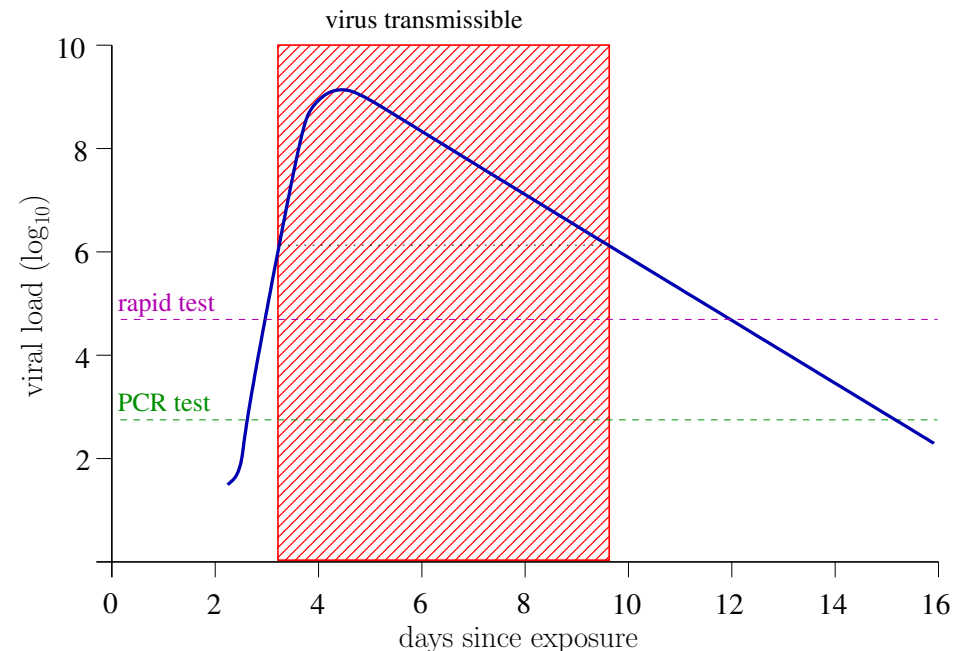
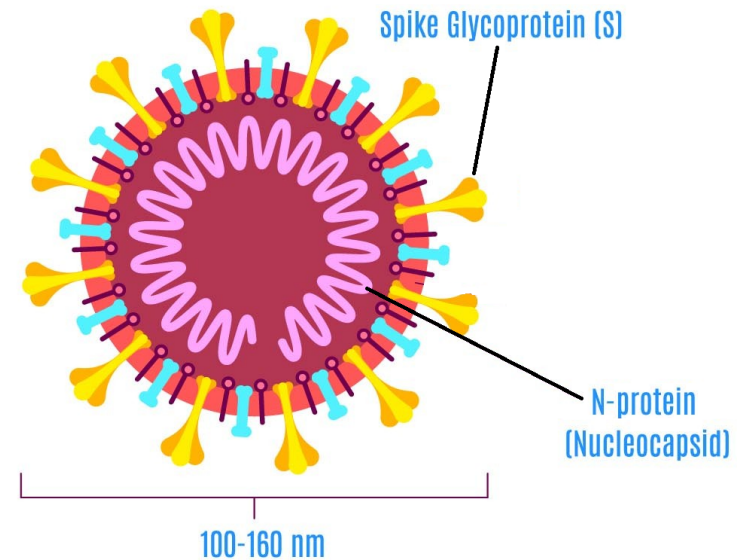
### Characteristics

Cheap: \$1-2

Results in 15 minutes

Home tests, no equipment: Saliva + saline solution + small cup

Less sensitive: Need equivalent of 100,000 viral strands/ml



## Antigen tests: II rapid tests

Made by US-based companies including e25 Bio, Sona Nanotech, Icen Diagnostics, OraSure

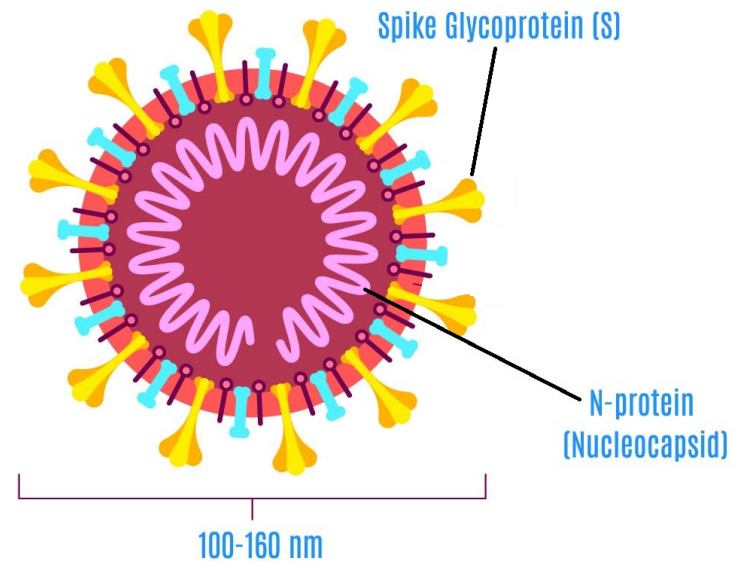
### Characteristics

Cheap: \$1-2

Results in 15 minutes

Home tests, no equipment: Saliva + saline solution + small cup

Less sensitive: Need equivalent of 100,000 viral strands/ml

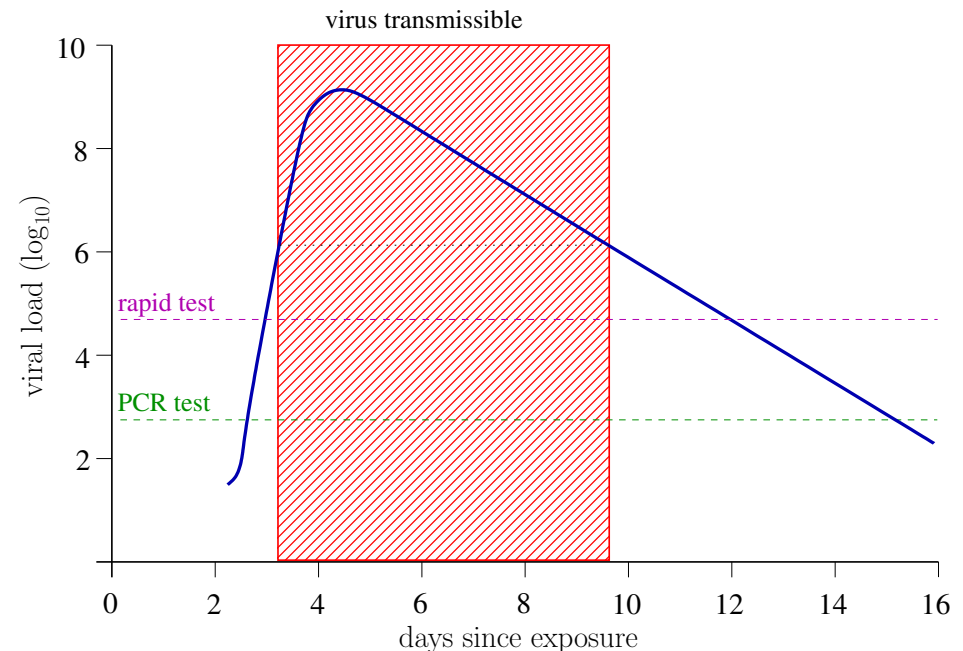


### e25 Bio test

Paper strip no larger than 1x5 in

Looks for the **spike** (S) protein  
on the *outside* of the virus

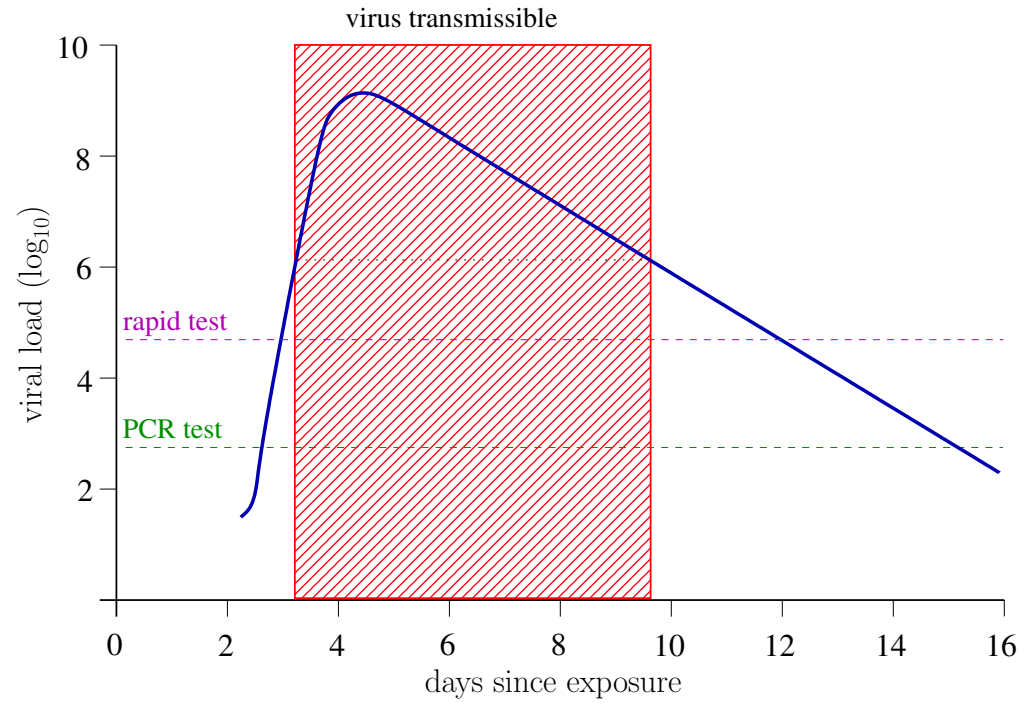
Sensitivity: 60-80%



## PCR vs rapid tests

For  $\approx 24$  hours in the beginning the tests give different results,  
rapid tests give false negatives

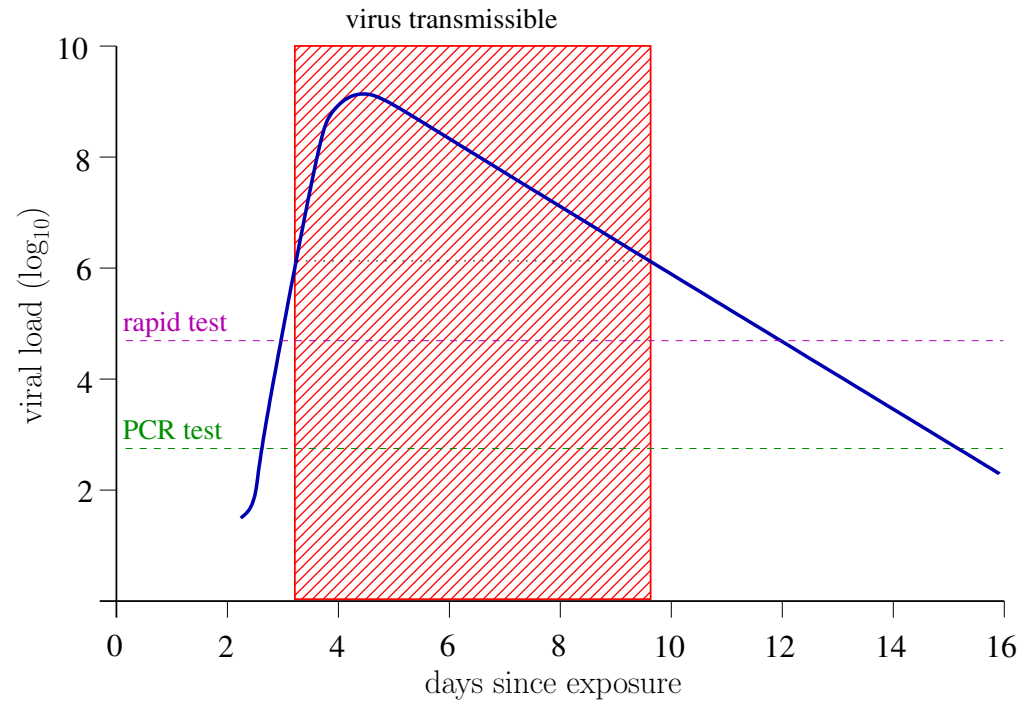
BUT: For a long time in the end  
PCR detects mostly dead virus



## PCR vs rapid tests

For  $\approx 24$  hours in the beginning the tests give different results,  
rapid tests give false negatives

BUT: For a long time in the end  
PCR detects mostly dead virus



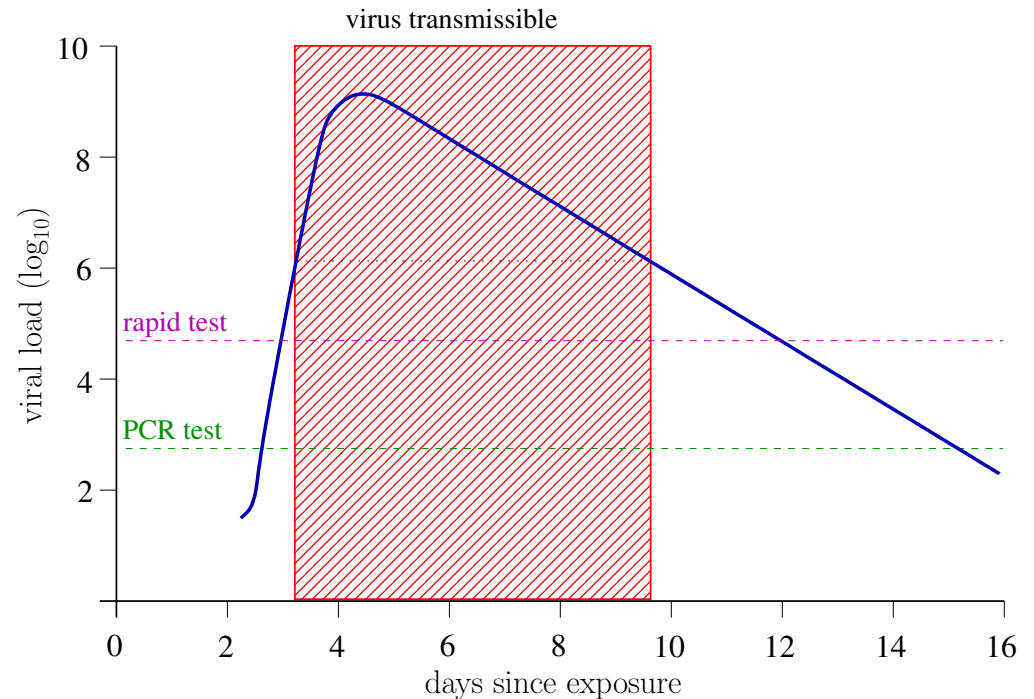
~> PRC tests are **clinical diagnostic tests**

~> Rapid tests are **contagiousness tests**

## PCR vs rapid tests

For  $\approx 24$  hours in the beginning the tests give different results,  
rapid tests give false negatives

BUT: For a long time in the end  
PCR detects mostly dead virus



~> PCR tests are **clinical diagnostic tests**

~> Rapid tests are **contagiousness tests**

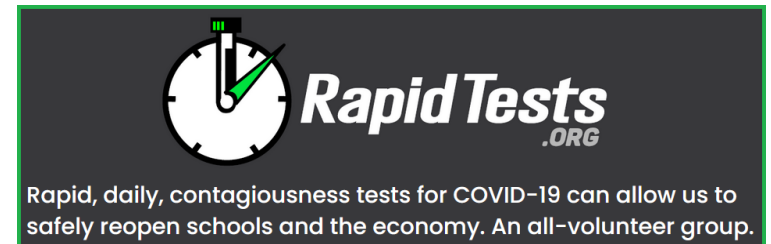
- ▶ FDA will not change its regulations to approve antigen tests but could re-frame them as *transmission-detecting tests* or *surveillance tools*
- ▶ The FDA recently stated they would consider less sensitive tests as part of a high-frequency testing plan

# Population-scale daily testing

△ Growing movement advocating 10s-100s of millions of rapid tests/day

Idea: **Test every individual before every major social contact:**

Work, school, cinema, shopping, etc



# Population-scale daily testing

△ Growing movement advocating 10s-100s of millions of rapid tests/day

Idea: **Test every individual before every major social contact:**

Work, school, cinema, shopping, etc

Epidemiologists at Harvard and Yale claim

- ▶ it will stop the virus in three weeks
- ▶ normal life will resume completely
- ▶ only the government can do it!

**Cost:** Even 500 million tests/day, total cost  $< 5\%$  of the \$3 trillion Congress already spent on Covid-related support for the economy





# Population-scale daily testing

△ Growing movement advocating 10s-100s of millions of rapid tests/day

Idea: **Test every individual before every major social contact:**

Work, school, cinema, shopping, etc

Epidemiologists at Harvard and Yale claim

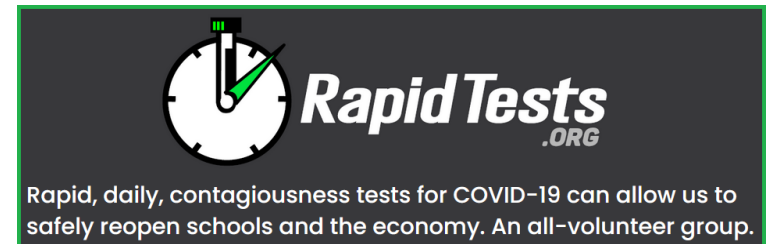
- ▶ it will stop the virus in three weeks
- ▶ normal life will resume completely
- ▶ only the government can do it!

**Cost:** Even 500 million tests/day, total cost < 5% of the \$3 trillion

Congress already spent on Covid-related support for the economy

**Concerns:**

- ▷ massive production capacity
- ▷ test and isolation compliance
- ▷ loss of public health surveillance data
- ▷ PCR for the rich/rapid tests for the poor?



# Population-scale daily testing

△ Growing movement advocating 10s-100s of millions of rapid tests/day

Idea: **Test every individual before every major social contact:**

Work, school, cinema, shopping, etc

Epidemiologists at Harvard and Yale claim

- ▶ it will stop the virus in three weeks
- ▶ normal life will resume completely
- ▶ only the government can do it!

**Cost:** Even 500 million tests/day, total cost  $< 5\%$  of the \$3 trillion

Congress already spent on Covid-related support for the economy

- Concerns:**
- ▷ massive production capacity
  - ▷ test and isolation compliance
  - ▷ loss of public health surveillance data
  - ▷ PCR for the rich/rapid tests for the poor?



~> **How much testing** is needed for this to work?

# Outline

## △ Covid-19 tests

- ▷ Antibody tests
- ▷ PCR tests
- ▷ Antigen tests
  - ~> Rapid, cheap, at-home tests

## △ Epidemiological models

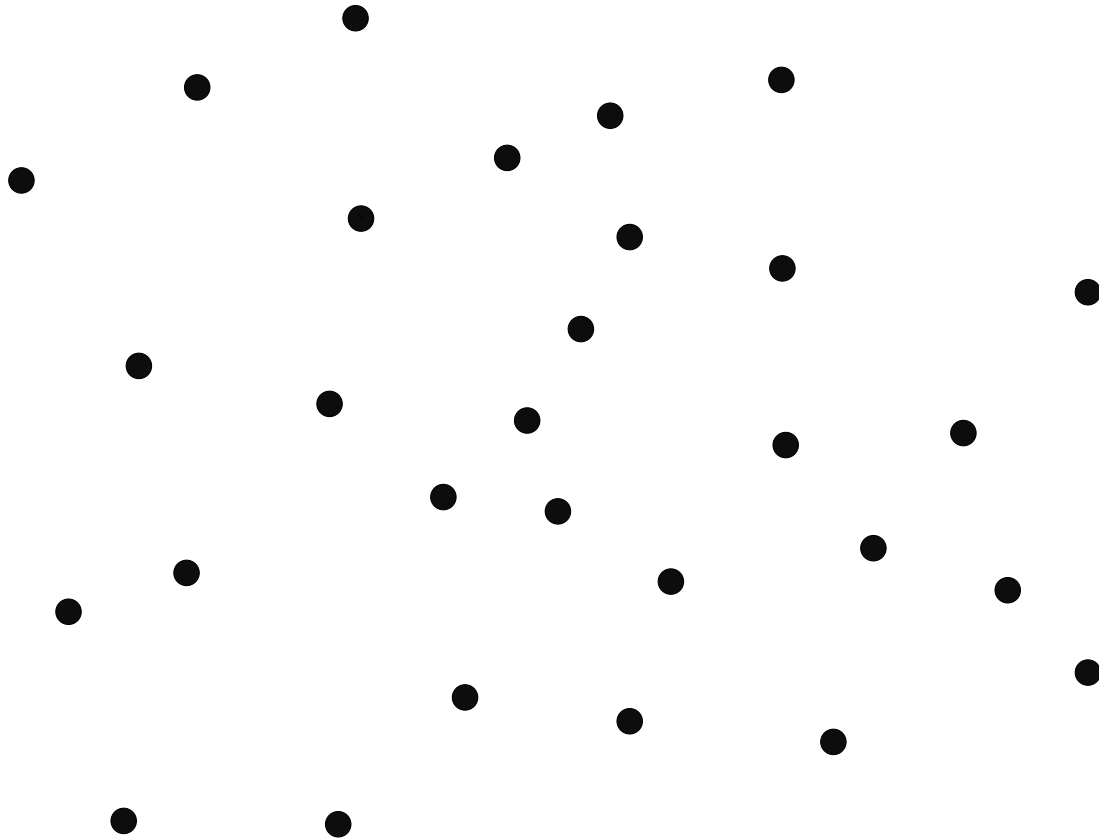
- ▷ SIR models on random population networks
  - ▶ Erdős-Rényi graphs
  - ▶ Random graphs with given degree distribution
- ▷ SIR epidemics with mass testing

## △ Necessary testing rates for suppression

- ▷ Rigorous results for a broad class of models
- ▷ Explicit numerical examples

# SIR epidemic on an Erdős-Rényi graph

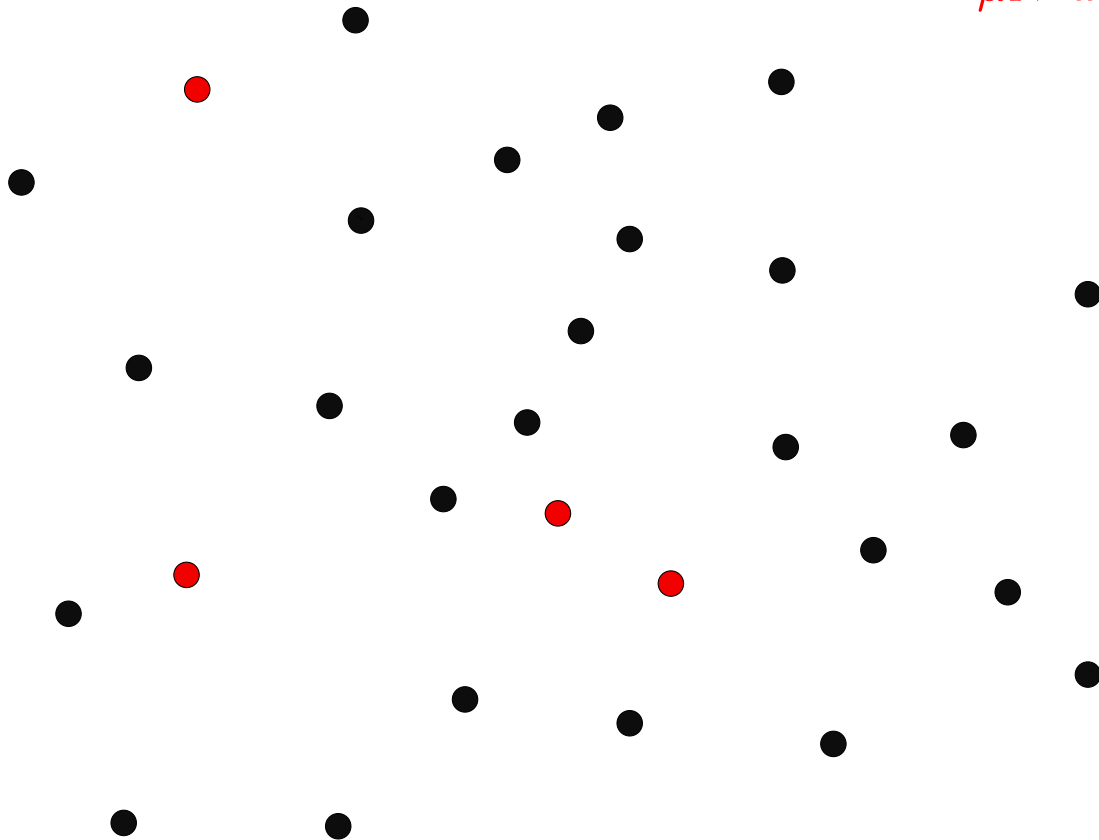
$N$  individuals



# SIR epidemic on an Erdős-Rényi graph

$N$  individuals

$\mu N$  infected

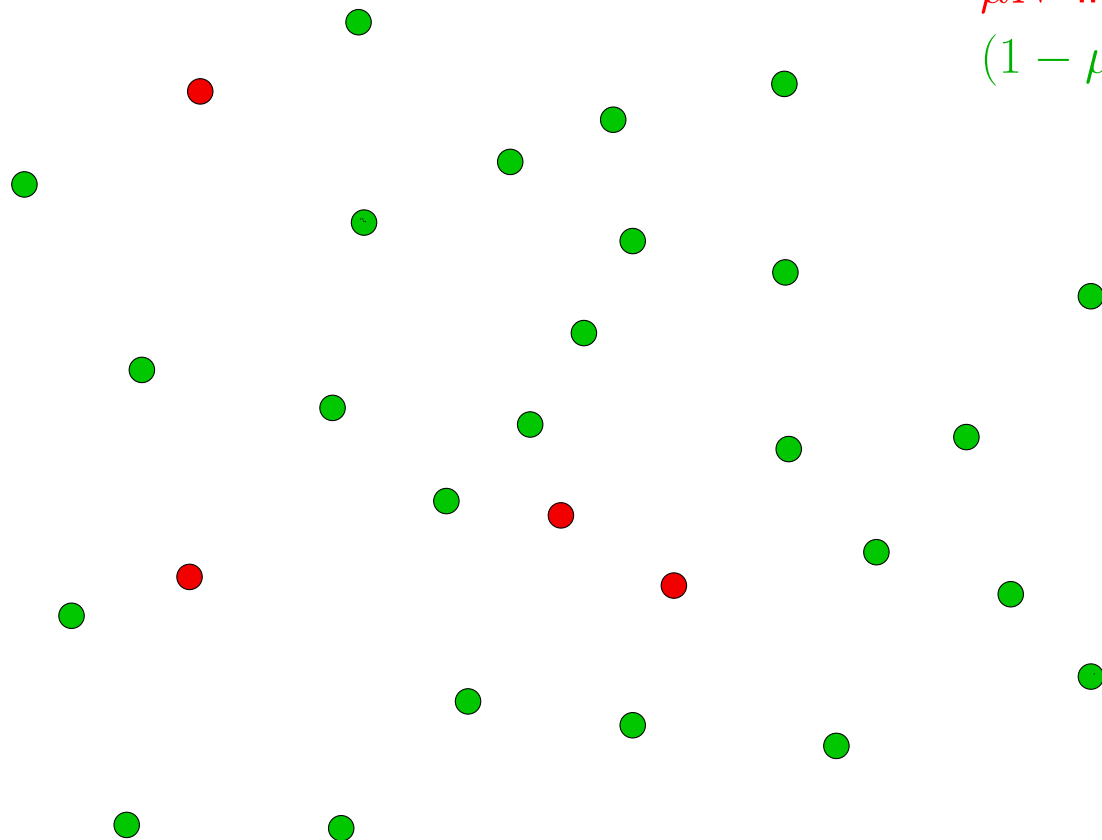


# SIR epidemic on an Erdős-Rényi graph

$N$  individuals

$\mu N$  infected

$(1 - \mu)N$  susceptible



# SIR epidemic on an Erdős-Rényi graph

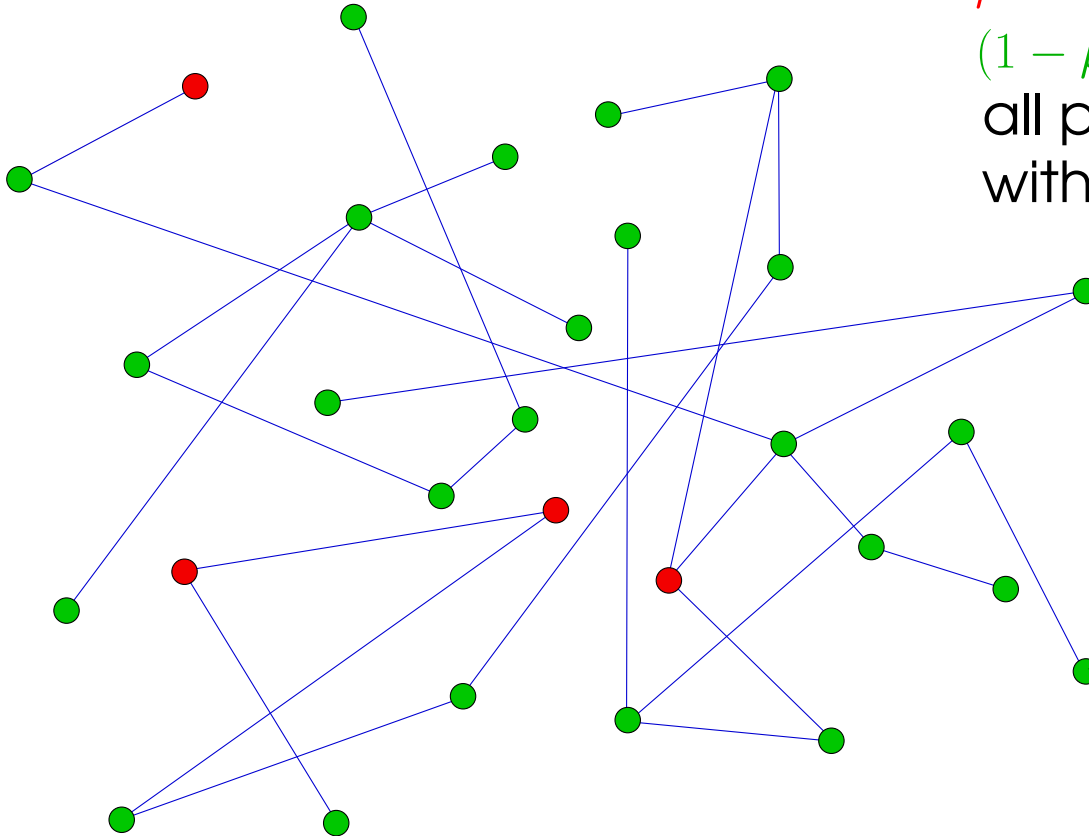
$N$  individuals

$\mu N$  infected

$(1 - \mu)N$  susceptible

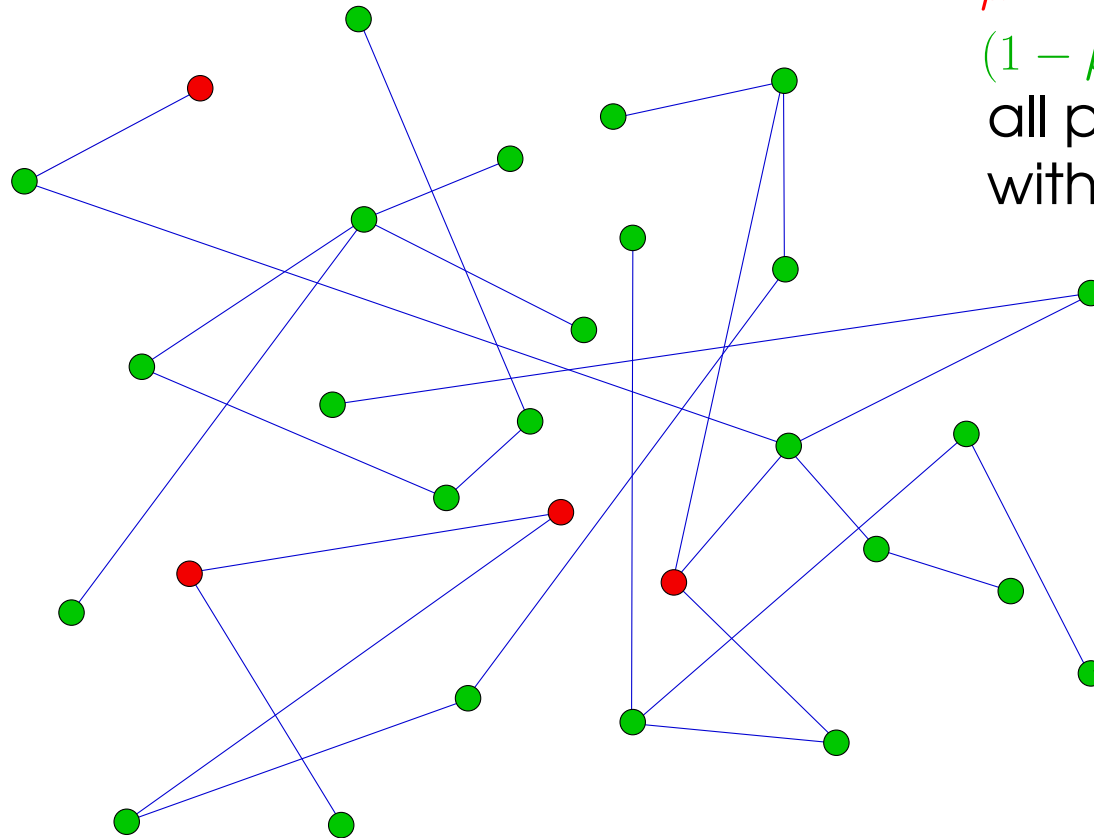
all pairs connected IID

with prob  $p = \alpha/N$





# SIR epidemic on an Erdős-Rényi graph



$N$  individuals

$\mu N$  infected

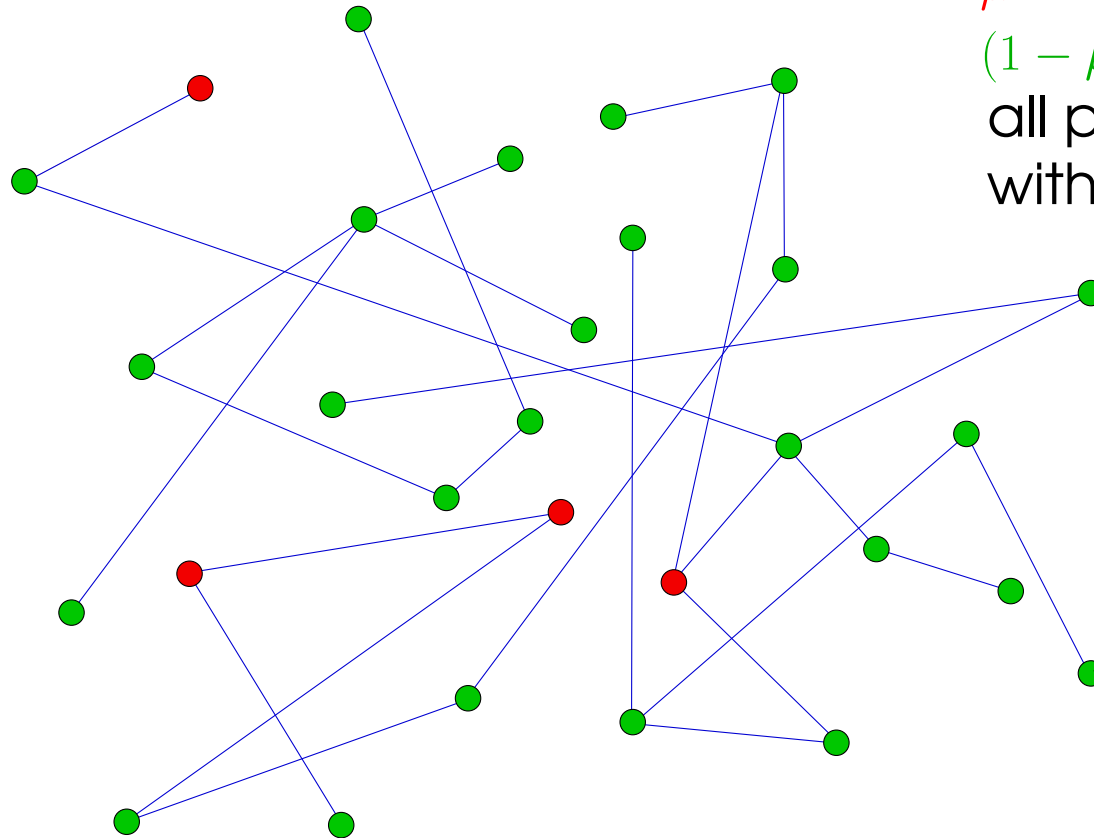
$(1 - \mu)N$  susceptible

all pairs connected IID

with prob  $p = \alpha/N$

▷ For large  $N$ , each individual has  $\approx \text{Poisson}(\alpha)$  acquaintances

# SIR epidemic on an Erdős-Rényi graph



$N$  individuals

$\mu N$  infected

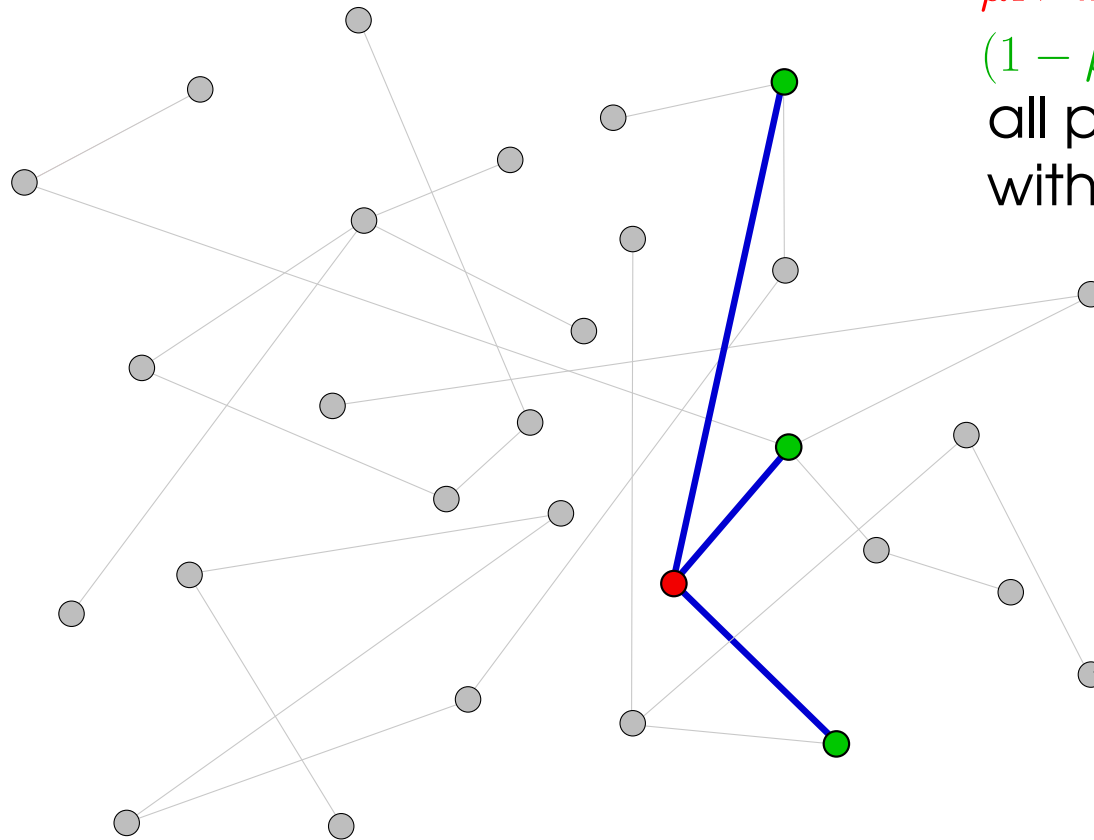
$(1 - \mu)N$  susceptible

all pairs connected IID

with prob  $p = \alpha/N$

- ▷ For large  $N$ , each individual has  $\approx \text{Poisson}(\alpha)$  acquaintances
- ▷ Each individual's infection has duration  $\text{Exp}(\gamma)$  days

# SIR epidemic on an Erdős-Rényi graph



$N$  individuals

$\mu N$  infected

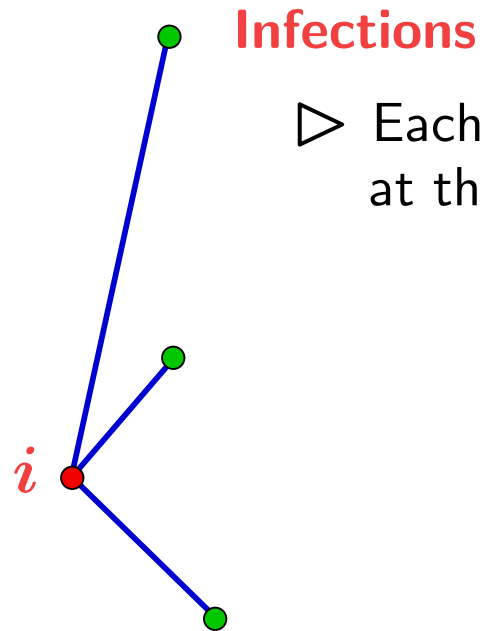
$(1 - \mu)N$  susceptible

all pairs connected IID

with prob  $p = \alpha/N$

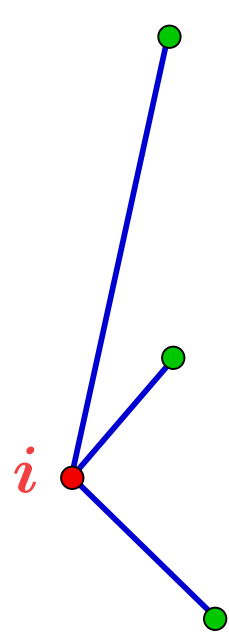
- ▷ For large  $N$ , each individual has  $\approx \text{Poisson}(\alpha)$  acquaintances
- ▷ Each individual's infection has duration  $\text{Exp}(\gamma)$  days
- ▷ Each infected individual makes infectious contact with their acquaintances at random times

# SIR epidemic + random testing



- ▷ Each infected individual becomes **recovered** at the end of their infection period

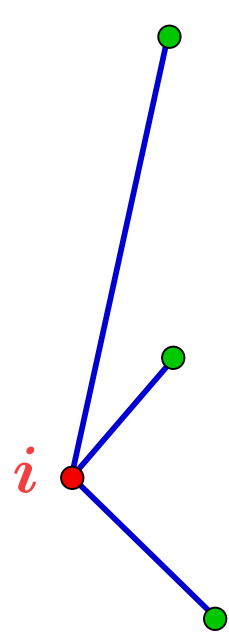
# SIR epidemic + random testing



## Infections

- ▷ Each infected individual becomes **recovered** at the end of their infection period
- ▷ Let  $d_i = \#$  of acquaintances of an infected individual  $i$
- ▷ Her **infectious contacts** are at the event times of a Poisson process with rate  $\beta d_i$  (while  $i$  remains infected)
- ▷ At each such time,  $i$  uniformly chooses an acquaintance and infects them

# SIR epidemic + random testing



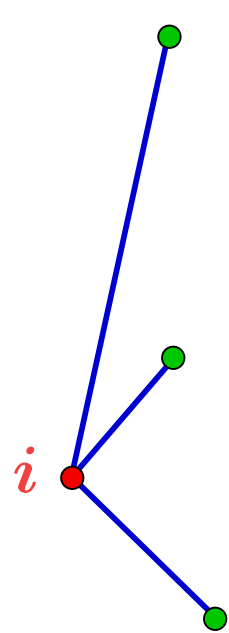
## Infections

- ▷ Each infected individual becomes **recovered** at the end of their infection period
- ▷ Let  $d_i = \#$  of acquaintances of an infected individual  $i$
- ▷ Her **infectious contacts** are at the event times of a Poisson process with rate  $\beta d_i$  (while  $i$  remains infected)
- ▷ At each such time,  $i$  uniformly chooses an acquaintance and infects them

## A parallel random testing process

- ▷ At the event times of a Poisson process with rate  $\theta N$  an individual is selected uniformly and tested, so that on average a **proportion  $\theta$  of the population is tested daily**

# SIR epidemic + random testing



## Infections

- ▷ Each infected individual becomes **recovered** at the end of their infection period
- ▷ Let  $d_i = \#$  of acquaintances of an infected individual  $i$
- ▷ Her **infectious contacts** are at the event times of a Poisson process with rate  $\beta d_i$  (while  $i$  remains infected)
- ▷ At each such time,  $i$  uniformly chooses an acquaintance and infects them

## A parallel random testing process

- ▷ At the event times of a Poisson process with rate  $\theta N$  an individual is selected uniformly and tested, so that on average a **proportion  $\theta$  of the population is tested daily**
- ▷ If found **+ve**, individual is quarantined until she becomes **recovered**

# SIR epidemic + random testing: Process evolution

## Testing parameters

*Sensitivity*:  $1 - \delta$ , with  $\delta$  = probability of false negative

*Specificity*: assume no false positives

*Compliance*:  $q$  = probability of quarantine compliance



# SIR epidemic + random testing: Process evolution

## Testing parameters

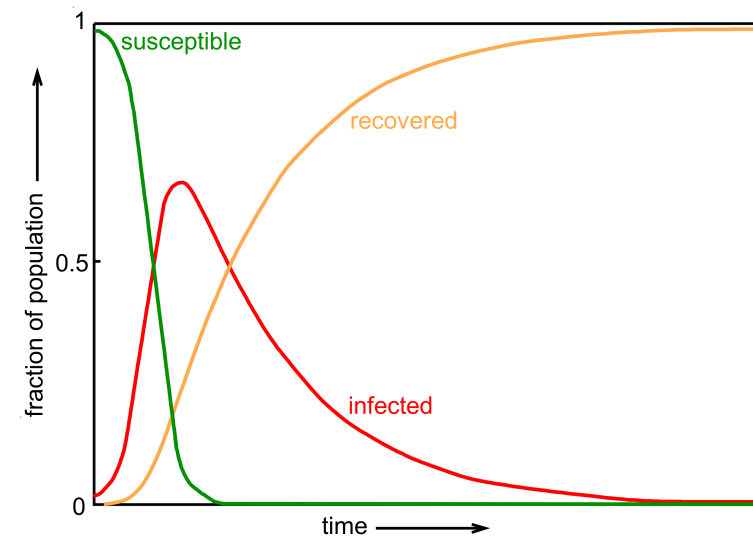
*Sensitivity*:  $1 - \delta$ , with  $\delta$  = probability of false negative

*Specificity*: assume no false positives

*Compliance*:  $q$  = probability of quarantine compliance

## Process evolution

- ▷ All random variables are independent
- ▷ This produces a large, continuous-time Markov process  $\{S_t, I_t, R_t\}$
- ▷ Typical SIR behaviour for large  $N$



# SIR epidemic + random testing: Process evolution

## Testing parameters

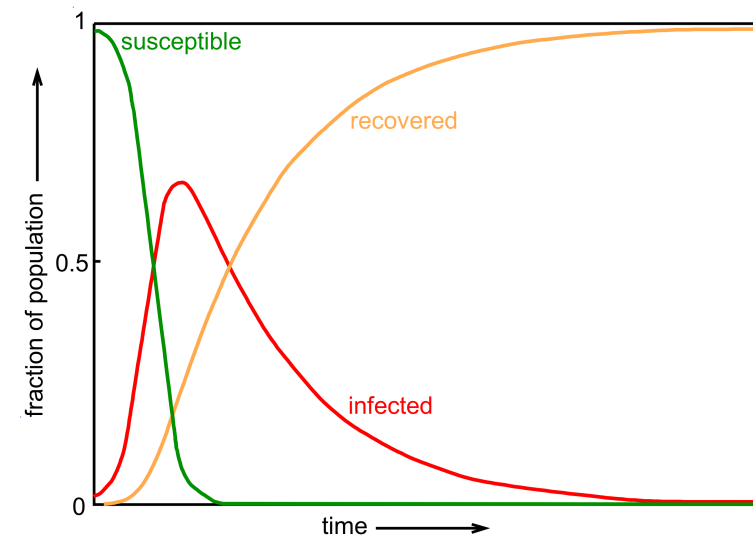
*Sensitivity*:  $1 - \delta$ , with  $\delta$  = probability of false negative

*Specificity*: assume no false positives

*Compliance*:  $q$  = probability of quarantine compliance

## Process evolution

- ▷ All random variables are independent
- ▷ This produces a large, continuous-time Markov process  $\{S_t, I_t, R_t\}$
- ▷ Typical SIR behaviour for large  $N$



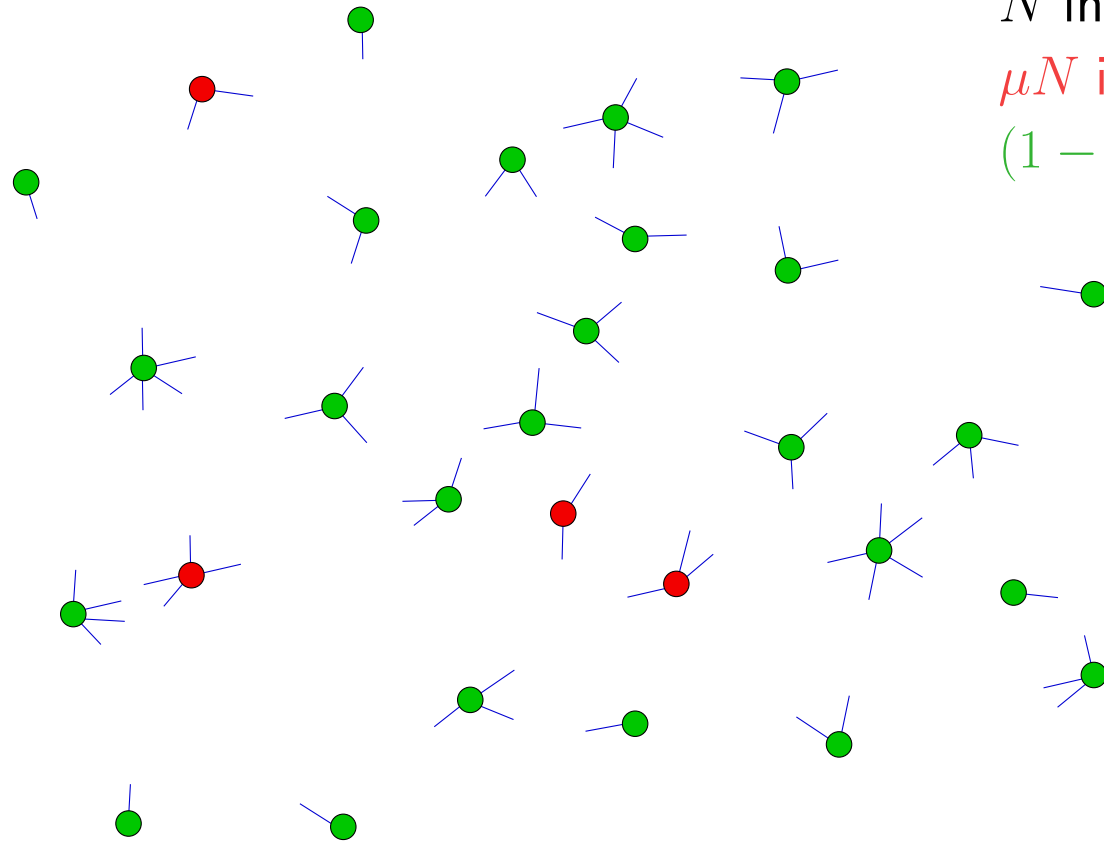
## Assumptions

↪ Model contains several **unrealistic assumptions**:

Poisson degree distr, no false positives, no geographical structure, no disease-specific characteristics, completely random testing

↪ But these mostly only make our **results more conservative**

# SIR epidemic on the configuration model



$N$  individuals

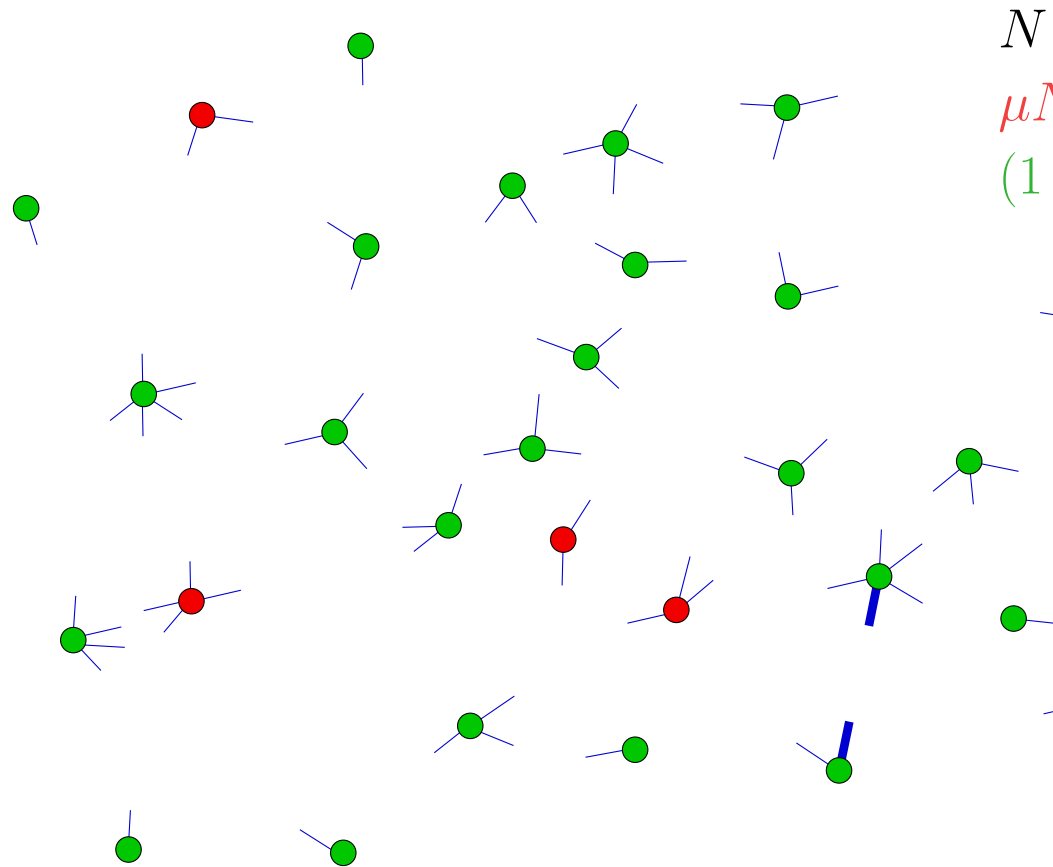
$\mu N$  infected

$(1 - \mu N)$  susceptible

● arbitrary degree sequence  $d_i$   
give each node  $i$  degree  $d_i$

draw  $d_i$  half-edges at each  $i$

# SIR epidemic on the configuration model



$N$  individuals

$\mu N$  infected

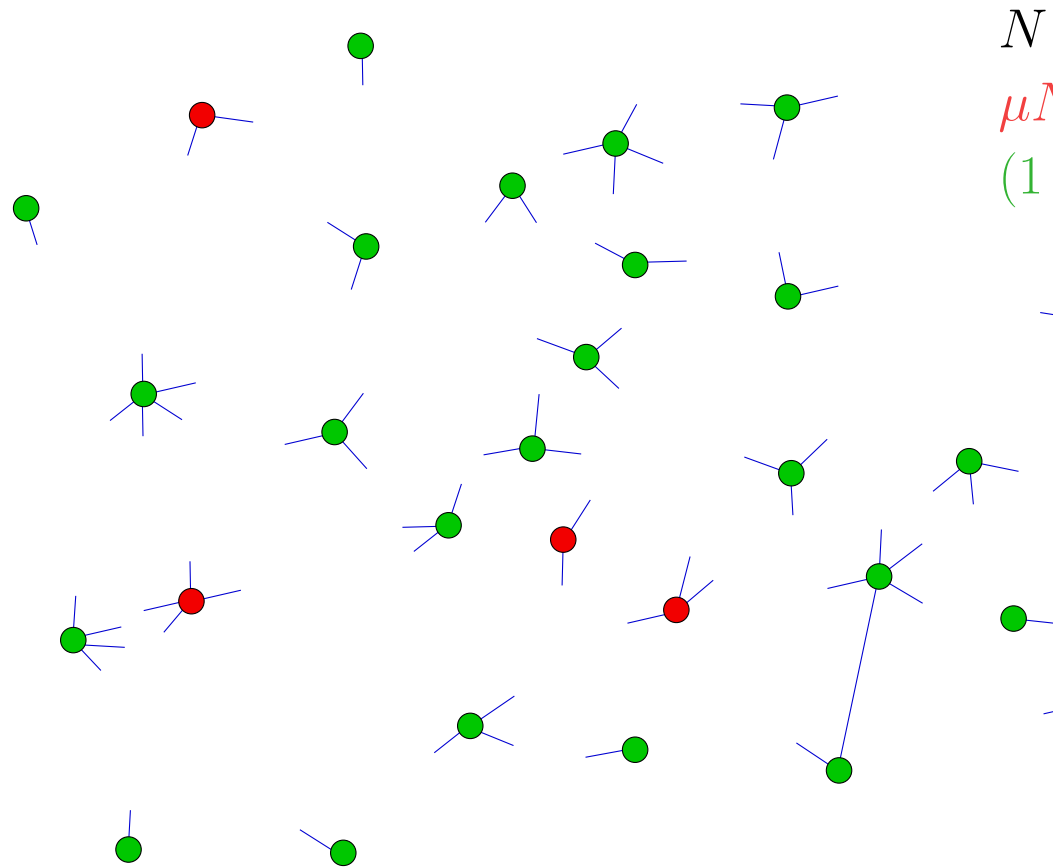
$(1 - \mu N)$  susceptible

● arbitrary degree sequence  $d_i$   
give each node  $i$  degree  $d_i$

draw  $d_i$  half-edges at each  $i$

● randomly select two

# SIR epidemic on the configuration model



$N$  individuals

$\mu N$  infected

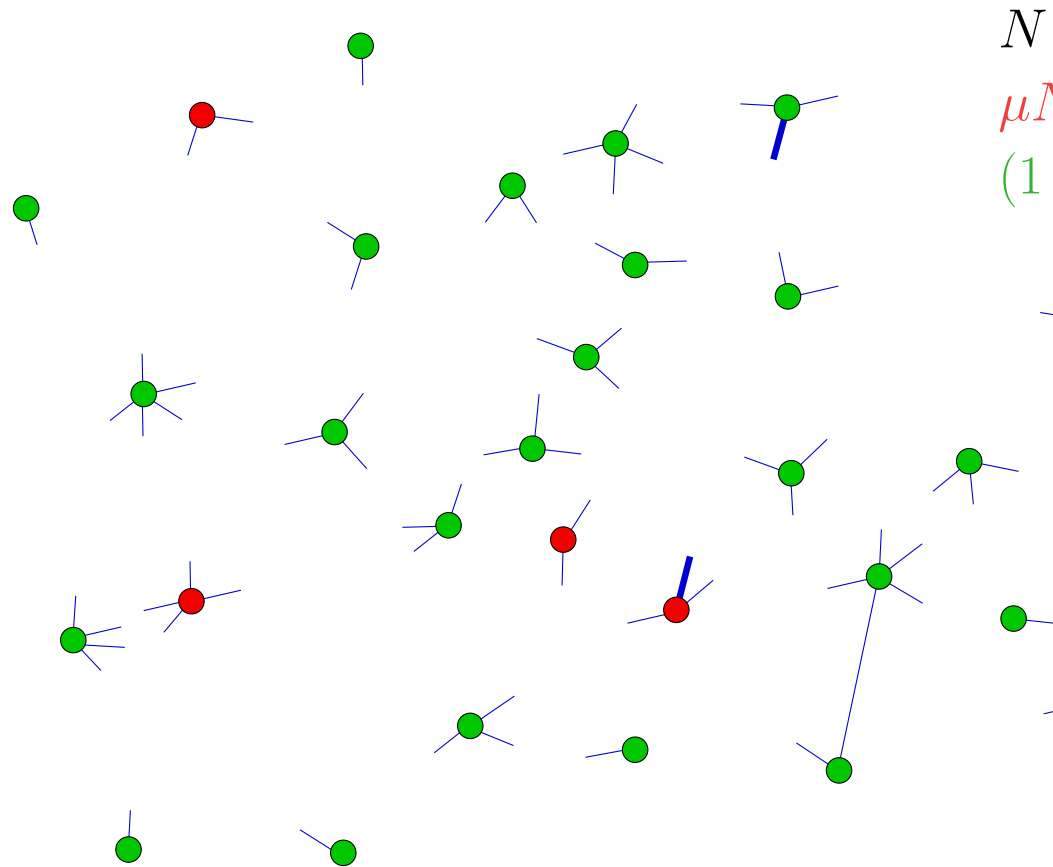
$(1 - \mu N)$  susceptible

● arbitrary degree sequence  $d_i$   
give each node  $i$  degree  $d_i$

draw  $d_i$  half-edges at each  $i$

● randomly select two  
join them

# SIR epidemic on the configuration model



$N$  individuals

$\mu N$  infected

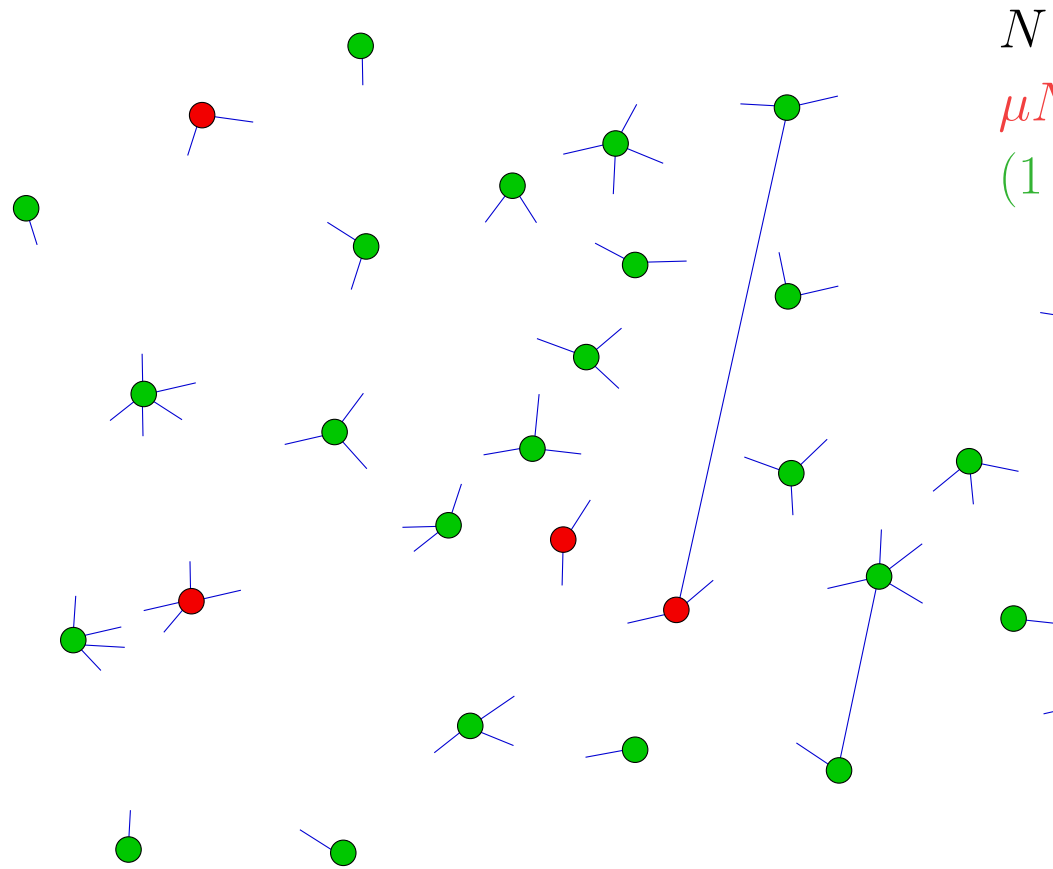
$(1 - \mu N)$  susceptible

● arbitrary degree sequence  $d_i$   
give each node  $i$  degree  $d_i$

draw  $d_i$  half-edges at each  $i$

● randomly select two  
join them  
repeat

# SIR epidemic on the configuration model



$N$  individuals

$\mu N$  infected

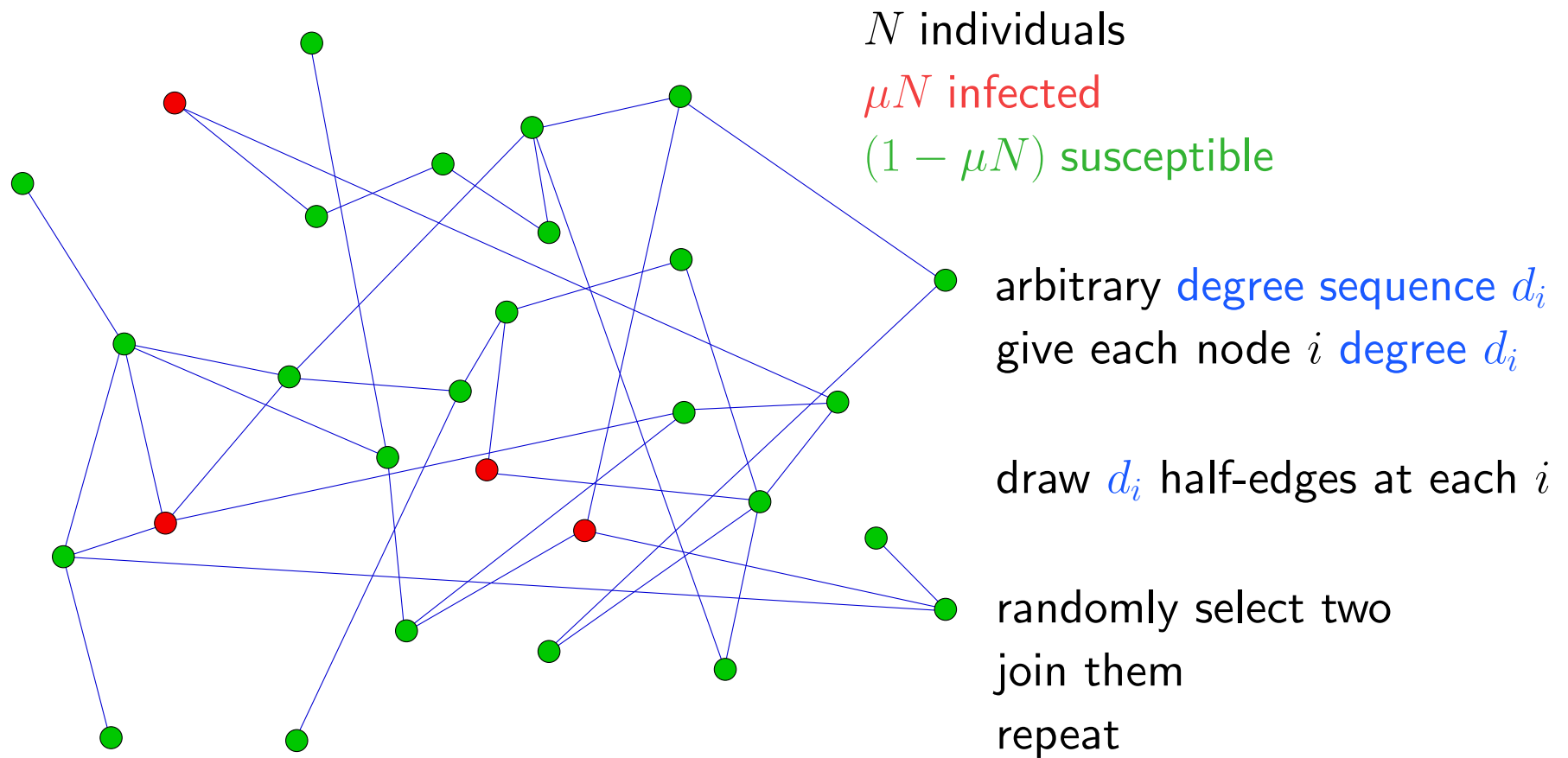
$(1 - \mu N)$  susceptible

● arbitrary degree sequence  $d_i$   
give each node  $i$  degree  $d_i$

draw  $d_i$  half-edges at each  $i$

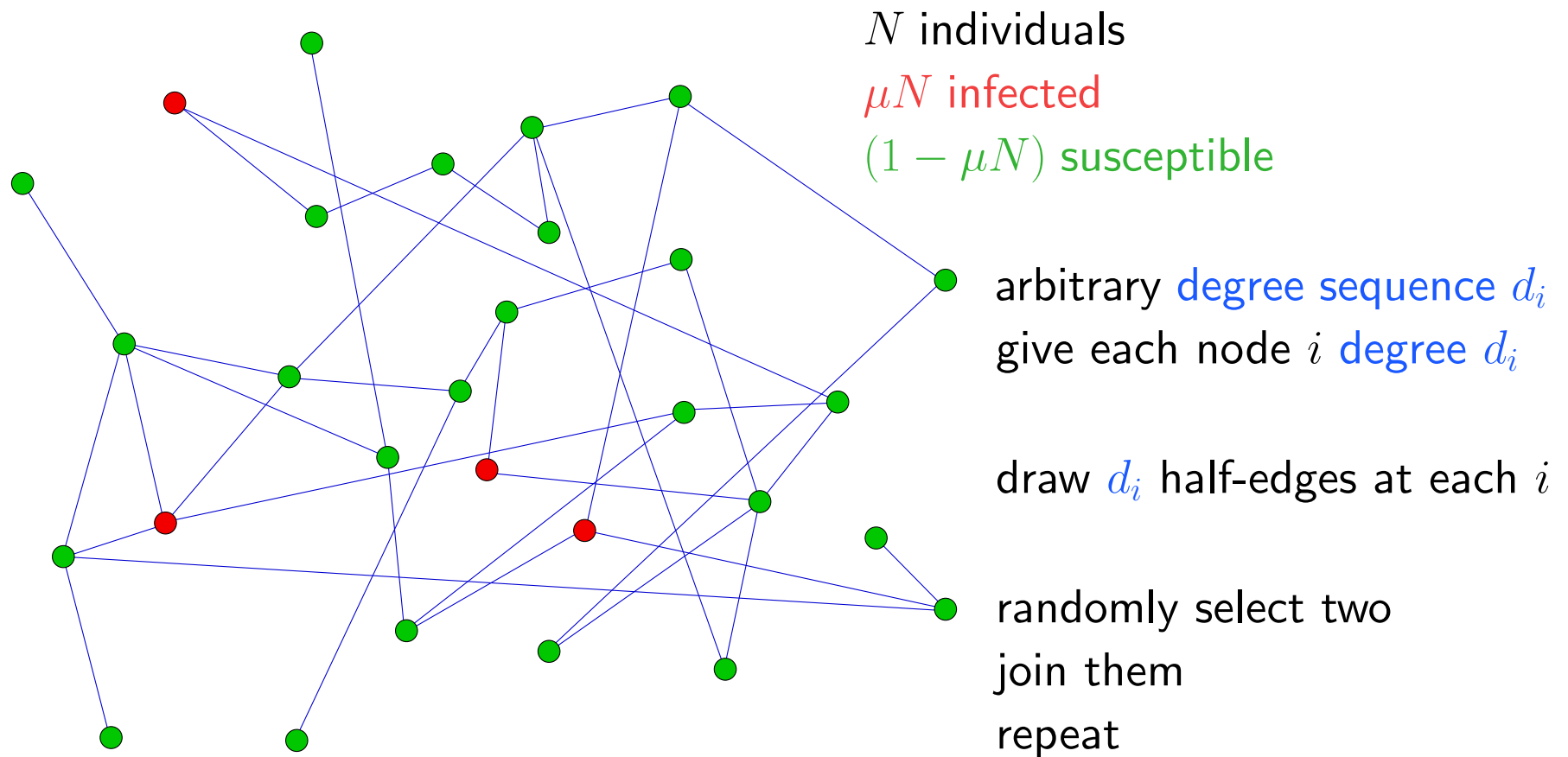
● randomly select two  
join them  
repeat

# SIR epidemic on the configuration model





# SIR epidemic on the configuration model



- ▷ **Possible problem:** There may be self-loops or multiple edges
- ▷ **Easy fix:** Re-do the pairing until there are no self-loops or multiple edges
- ▷ **Result:** Graph uniformly chosen among those with degree sequence  $\{d_i\}$

# SIR epidemic + testing on the configuration model

## Assumptions

- ▶ Degree sequence  $\{d_i\}$  can be chosen arbitrarily. But:
- ▶  $\{d_i\}$  usually chosen according to a given degree distr  $\{p_k\}$
- ▶ Proportion of individuals with degree  $k$  is  $\approx p_k$ 
  - $\leadsto$  This holds separately for both infected and susceptible nodes

# SIR epidemic + testing on the configuration model

## Assumptions

- ▶ Degree sequence  $\{d_i\}$  can be chosen arbitrarily. But:
- ▶  $\{d_i\}$  usually chosen according to a given degree distr  $\{p_k\}$
- ▶ Proportion of individuals with degree  $k$  is  $\approx p_k$ 
  - $\leadsto$  This holds separately for both infected and susceptible nodes

## Epidemic + testing process

- ▷ Exactly the same as before (once graph is fixed)
- ▷ Again, typical SIR behaviour for large  $N$

# SIR epidemic + testing on the configuration model

## Assumptions

- ▶ Degree sequence  $\{d_i\}$  can be chosen arbitrarily. But:
- ▶  $\{d_i\}$  usually chosen according to a given degree distr  $\{p_k\}$
- ▶ Proportion of individuals with degree  $k$  is  $\approx p_k$ 
  - $\leadsto$  This holds separately for both infected and susceptible nodes

## Epidemic + testing process

- ▷ Exactly the same as before (once graph is fixed)
  - ▷ Again, typical SIR behaviour for large  $N$
- $\leadsto$  How does the testing rate affect the evolution of the epidemic?

# Outline

## △ Covid-19 tests

- ▷ Antibody tests
- ▷ PCR tests
- ▷ Antigen tests
  - ~> Rapid, cheap, at-home tests

## △ Epidemiological models

- ▷ SIR models on random population networks
  - ▶ Erdős-Rényi graphs
  - ▶ Random graphs with given degree distribution
- ▷ SIR epidemics with mass testing

## △ Necessary testing rates for suppression

- ▷ Rigorous results for a broad class of models
- ▷ Explicit numerical examples

# Basic reproduction number $R_0$ for the E-R model

For the SIR epidemic on the Erdős-Rényi graph

## Theorem

As  $N \rightarrow \infty$ :

**i. no testing** [e.g. Andersson (1999), Neal (2003)]

$$R_0 = \frac{\alpha\beta}{\beta + \gamma}$$

# Basic reproduction number $R_0$ for the E-R model

For the SIR epidemic on the Erdős-Rényi graph

## Theorem

As  $N \rightarrow \infty$ :

**i. no testing** [e.g. Andersson (1999), Neal (2003)]

$$R_0 = \frac{\alpha\beta}{\beta + \gamma}$$

**ii. random testing**

$$R_0(\theta) = \frac{\alpha\beta}{\beta + \gamma + \theta(1 - \delta)q}$$

# Basic reproduction number $R_0$ for the E-R model

For the SIR epidemic on the Erdős-Rényi graph

## Theorem

As  $N \rightarrow \infty$ :

i. **no testing** [e.g. Andersson (1999), Neal (2003)]

$$R_0 = \frac{\alpha\beta}{\beta + \gamma}$$

ii. **random testing**

$$R_0(\theta) = \frac{\alpha\beta}{\beta + \gamma + \theta(1 - \delta)q}$$

## Proof.

Adding testing to the model is exactly equivalent to shortening the mean infection duration:

$$\gamma \mapsto \gamma + \theta(1 - \delta)q$$

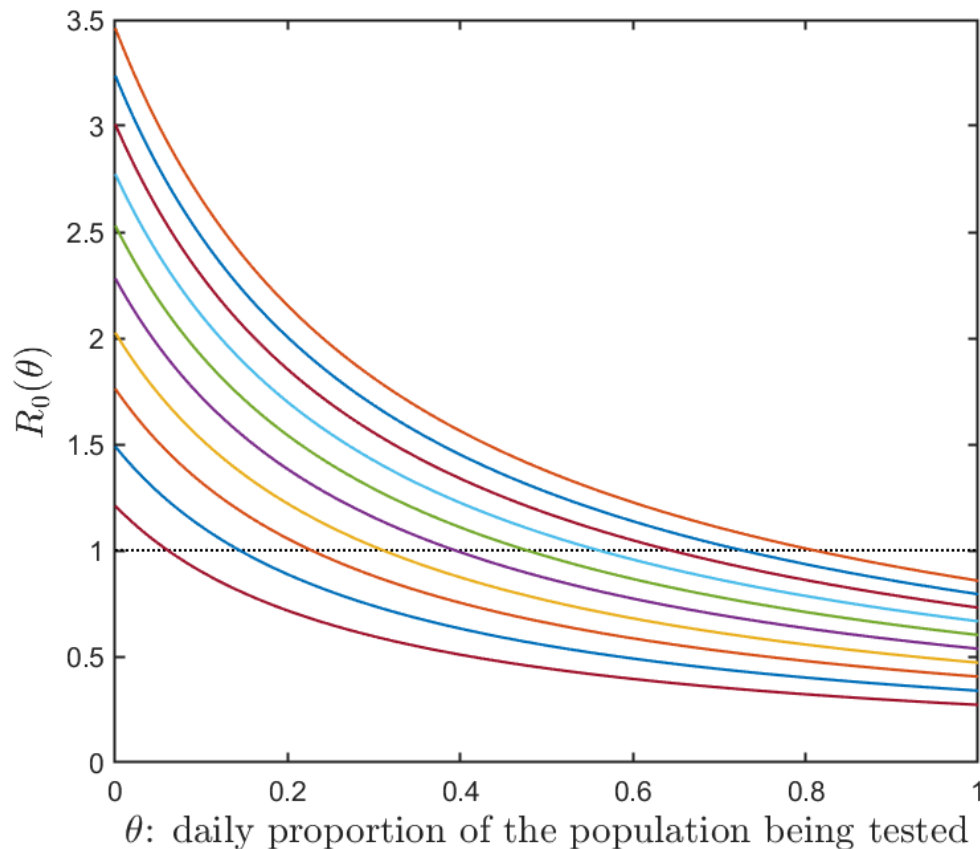
□



## $R_0$ for the E-R model: Examples

1-in-10000 initially infected  
20 acquaintances/individual on average  
average infectious period 7 days  
contact rate  $\beta$  varies  
 $\Rightarrow 1.2 \leq R_0 \leq 3.5$

quarantine compliance 75%  
**test sensitivity 70%**  
testing rate  $\theta$



## $R_0$ for the E-R model: Examples

1-in-10000 initially infected  
20 acquaintances/individual on average  
average infectious period 7 days  
contact rate  $\beta$  varies

$$\Rightarrow 1.2 \leq R_0 \leq 3.5$$

quarantine compliance 75%

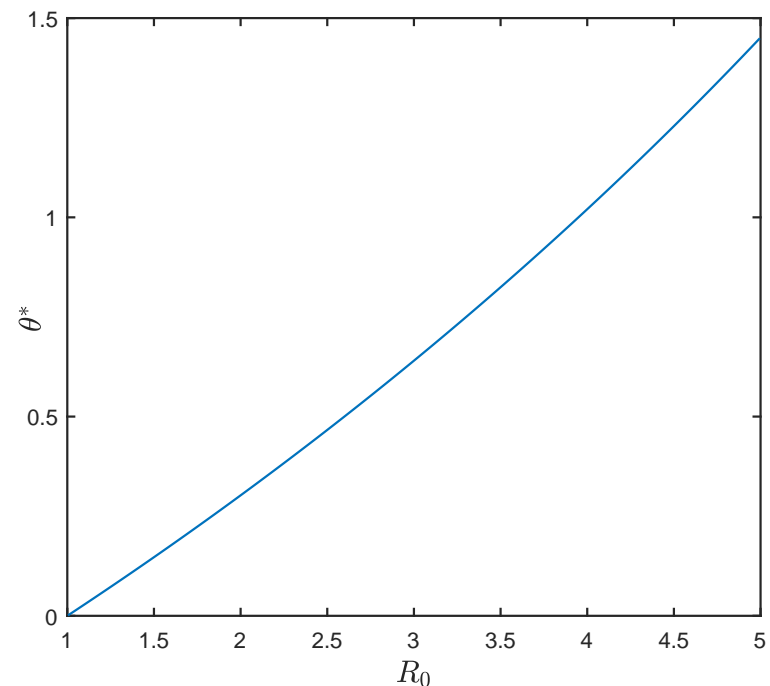
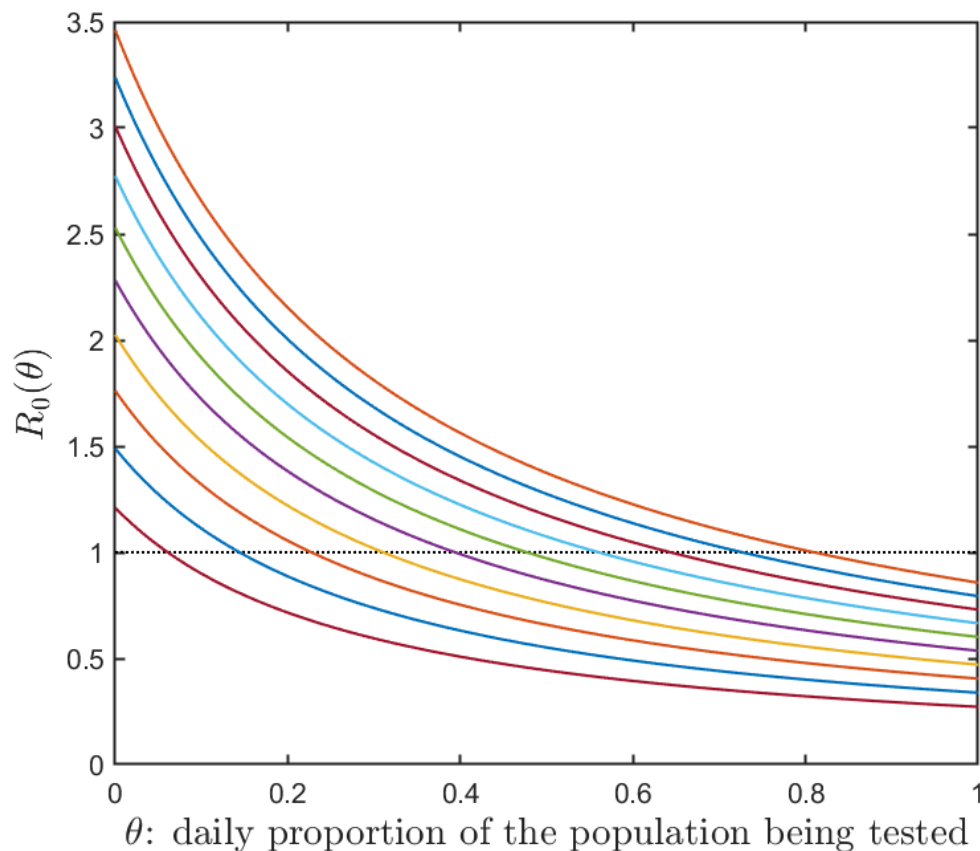
**test sensitivity 70%**

testing rate  $\theta$

### Corollary

Testing rate required for  $R_0(\theta) < 1$  :

$$\theta > \theta^* = \frac{\alpha\beta - \beta - \gamma}{q(1 - \delta)}$$



## Epidemic size for the E-R model

For the SIR epidemic on the Erdős-Rényi graph, write

$T_N =$  total size of the epidemic

and let

$$\begin{aligned}\tau(r, \mu) &= \min\{t > 0 : e^{-rt} = 1 + \mu - t\} \\ s(r, \mu) &= 1 - e^{-r\tau(r, \mu)}\end{aligned}$$

### Theorem

As  $N \rightarrow \infty$ :

- i. no testing [e.g. Neal (2003)]  $\frac{T_N}{N} \rightarrow s = s(R_0, \mu)$
- ii. random testing

$$\frac{T_N}{N} \rightarrow s(\theta) = s(R_0(\theta), \mu)$$

**Proof.**

Same as before

□

# Epidemic size for the E-R model: Examples

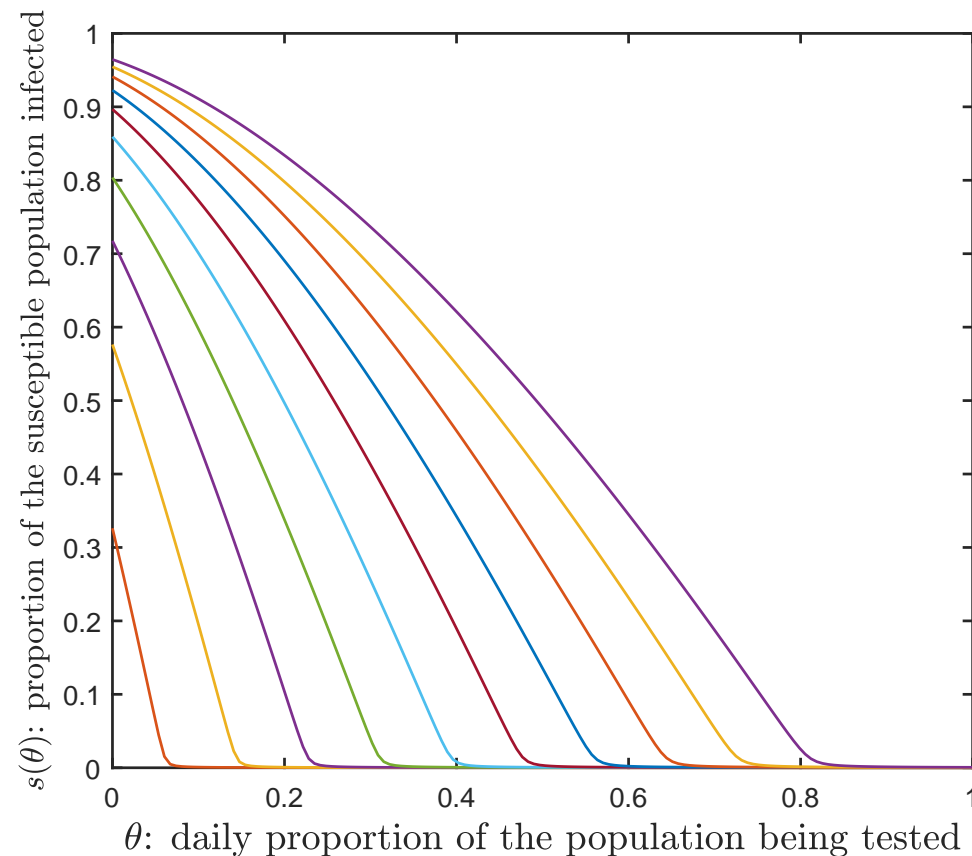
1-in-10000 initially infected  
20 acquaintances/individual on average  
average infectious period 7 days  
contact rate  $\beta$  varies

quarantine compliance 75%

**test sensitivity 70%**

testing rate  $\theta$

$$\Rightarrow 1.2 \leq R_0 \leq 3.5$$



## Small epidemics with the E-R model

For the SIR epidemic on the Erdős-Rényi graph suppose that instead of  $\mu N$  initially infected individuals we only have  $m$  of them

Let 
$$f(p, \gamma) = \gamma \int_0^\infty \exp \left\{ -\gamma z - \alpha(1-p)(1 - e^{-\beta z}) \right\} dz$$

We say there is a **small epidemic** if  $T_N = O(1)$  as  $N \rightarrow \infty$

## Small epidemics with the E-R model

For the SIR epidemic on the Erdős-Rényi graph suppose that instead of  $\mu N$  initially infected individuals we only have  $m$  of them

Let 
$$f(p, \gamma) = \gamma \int_0^\infty \exp \left\{ -\gamma z - \alpha(1-p)(1-e^{-\beta z}) \right\} dz$$

We say there is a **small epidemic** if  $T_N = O(1)$  as  $N \rightarrow \infty$

### Theorem

i. **no testing** [Martin-Löf (1986)]

Suppose  $R_0 > 1$

Let  $p$  be the smallest root of  $f(p, \gamma) = p$  in  $[0, 1]$

$\Rightarrow$  With prob  $p^m$  there is only a **small epidemic**

## Small epidemics with the E-R model

For the SIR epidemic on the Erdős-Rényi graph suppose that instead of  $\mu N$  initially infected individuals we only have  $m$  of them

Let 
$$f(p, \gamma) = \gamma \int_0^\infty \exp \left\{ -\gamma z - \alpha(1-p)(1-e^{-\beta z}) \right\} dz$$

We say there is a **small epidemic** if  $T_N = O(1)$  as  $N \rightarrow \infty$

### Theorem

#### i. no testing [Martin-Löf (1986)]

Suppose  $R_0 > 1$

Let  $p$  be the smallest root of  $f(p, \gamma) = p$  in  $[0, 1]$

$\Rightarrow$  With prob  $p^m$  there is only a **small epidemic**

#### ii. random testing

Suppose  $R_0(\theta) > 1$

Let  $p(\theta)$  be the smallest root of  $f(p, \gamma + \theta(1-\delta)q) = p$  in  $[0, 1]$

$\Rightarrow$  With prob  $p(\theta)^m$  there is only a **small epidemic**

### Proof.

Same idea as before

□

# Probability of small epidemic: Examples

20 acquaintances/individual on average  
average infectious period 7 days

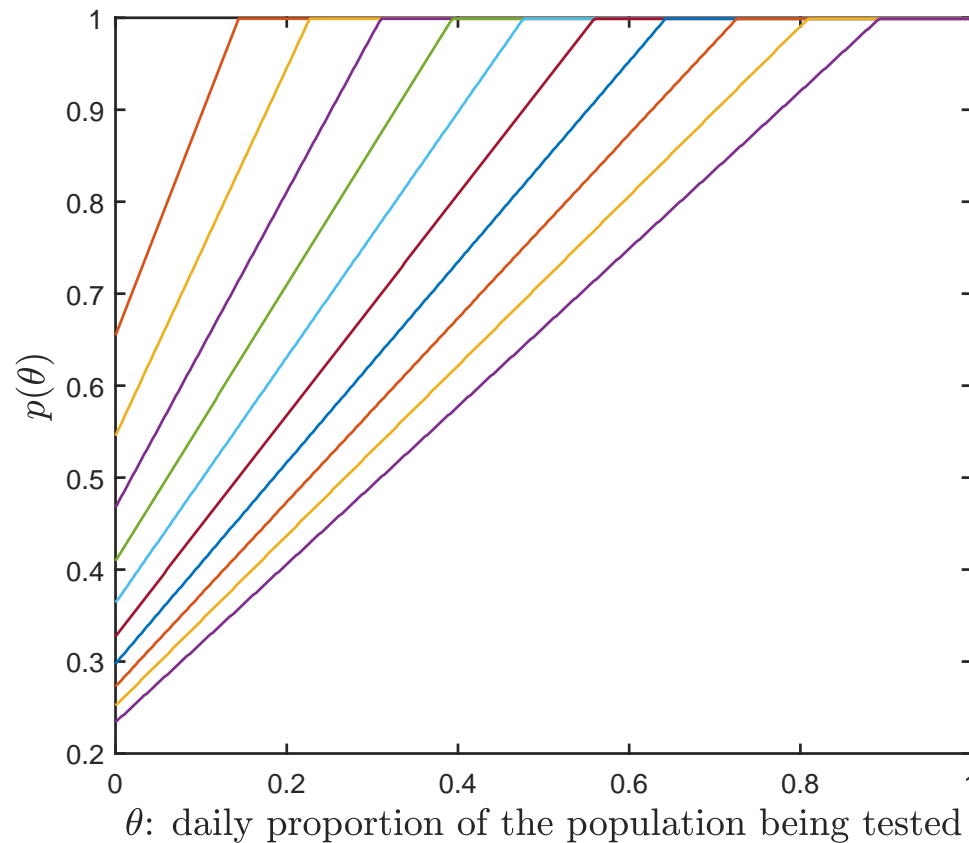
contact rate  $\beta$  varies

$$\Rightarrow 1.2 \leq R_0 \leq 3.5$$

quarantine compliance 75%

**test sensitivity 70%**

testing rate  $\theta$





# Basic reproduction number $R_0$ for the configuration model

For the SIR epidemic on the configuration model with degree distr  $\{p_k\}$  let

$$\lambda = \sum_{k=0}^{\infty} k p_k \qquad v^2 = \sum_{k=0}^{\infty} k(k-1)p_k$$

## Theorem

As  $N \rightarrow \infty$  (under mild conditions):

**i. no testing** [e.g. Janson-Luczak-Windridge (2014)]

$$R_0 = \left( \frac{\beta}{\beta + \gamma} \right) \left( \frac{(1 - \mu)v^2}{\lambda} \right)$$

## Basic reproduction number $R_0$ for the configuration model

For the SIR epidemic on the configuration model with degree distr  $\{p_k\}$  let

$$\lambda = \sum_{k=0}^{\infty} k p_k \qquad v^2 = \sum_{k=0}^{\infty} k(k-1)p_k$$

### Theorem

As  $N \rightarrow \infty$  (under mild conditions):

**i. no testing** [e.g. Janson-Luczak-Windridge (2014)]

$$R_0 = \left( \frac{\beta}{\beta + \gamma} \right) \left( \frac{(1 - \mu)v^2}{\lambda} \right)$$

**ii. random testing**

$$R_0(\theta) = \left( \frac{\beta}{\beta + \gamma + \theta(1 - \delta)q} \right) \left( \frac{(1 - \mu)v^2}{\lambda} \right)$$

**Proof.**

Same idea as before

□

## $R_0$ for the configuration model: Examples

1-in-10000 initially infected

degree distr  $p_k \propto k^{-1.75}e^{-0.02k}$

$\lambda \approx 3.5$  acquaintances/individual on average

average infectious period 7 days

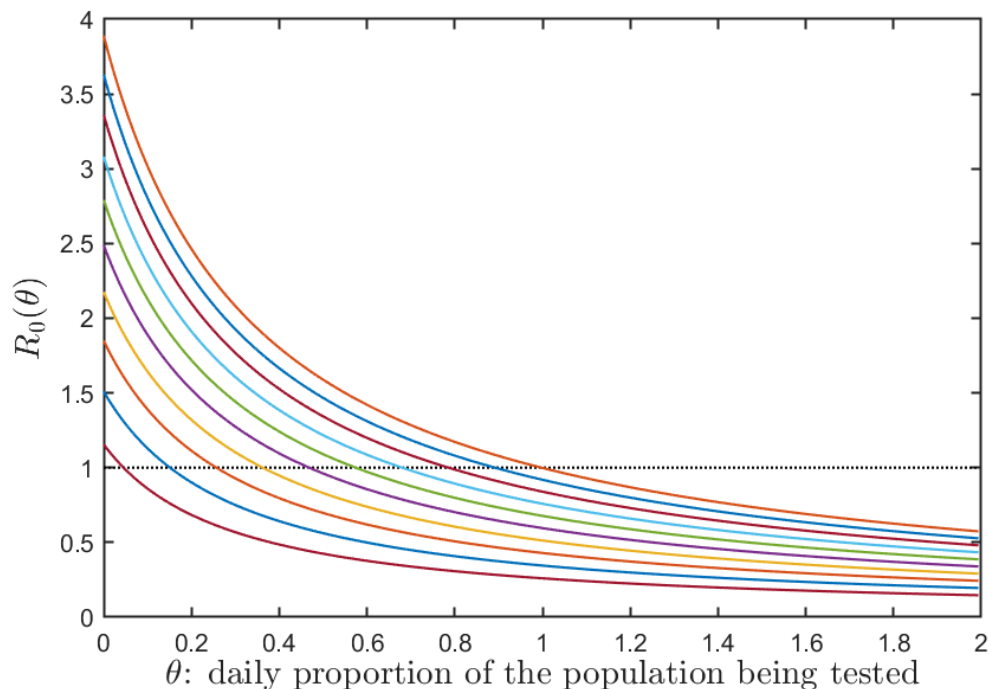
contact rate  $\beta$  varies

$$\Rightarrow 1.15 \leq R_0 \leq 3.9$$

quarantine compliance 75%

**test sensitivity 70%**

testing rate  $\theta$



## $R_0$ for the configuration model: Examples

1-in-10000 initially infected

degree distr  $p_k \propto k^{-1.75}e^{-0.02k}$

$\lambda \approx 3.5$  acquaintances/individual on average

average infectious period 7 days

contact rate  $\beta$  varies

$$\Rightarrow 1.15 \leq R_0 \leq 3.9$$

quarantine compliance 75%

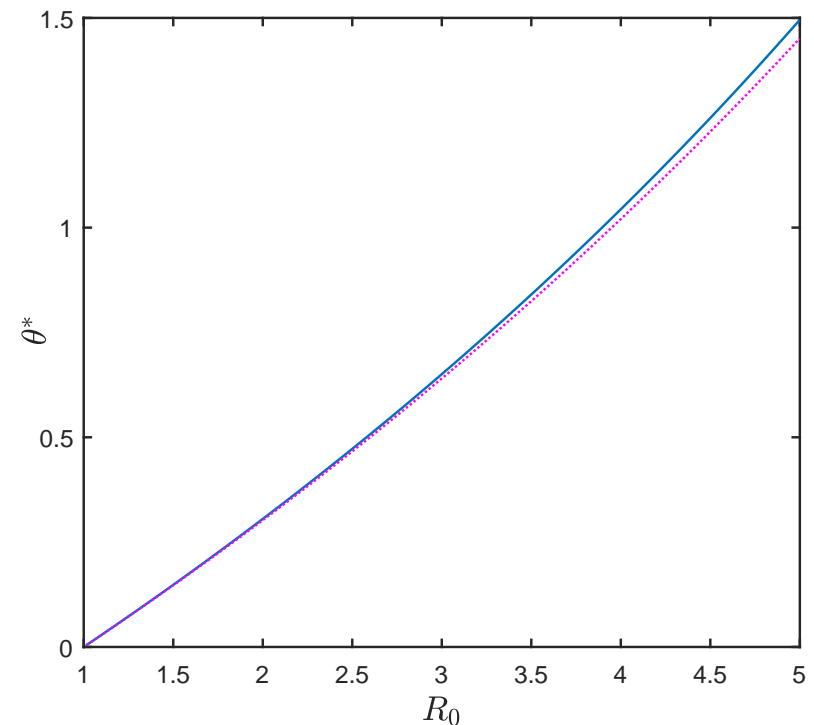
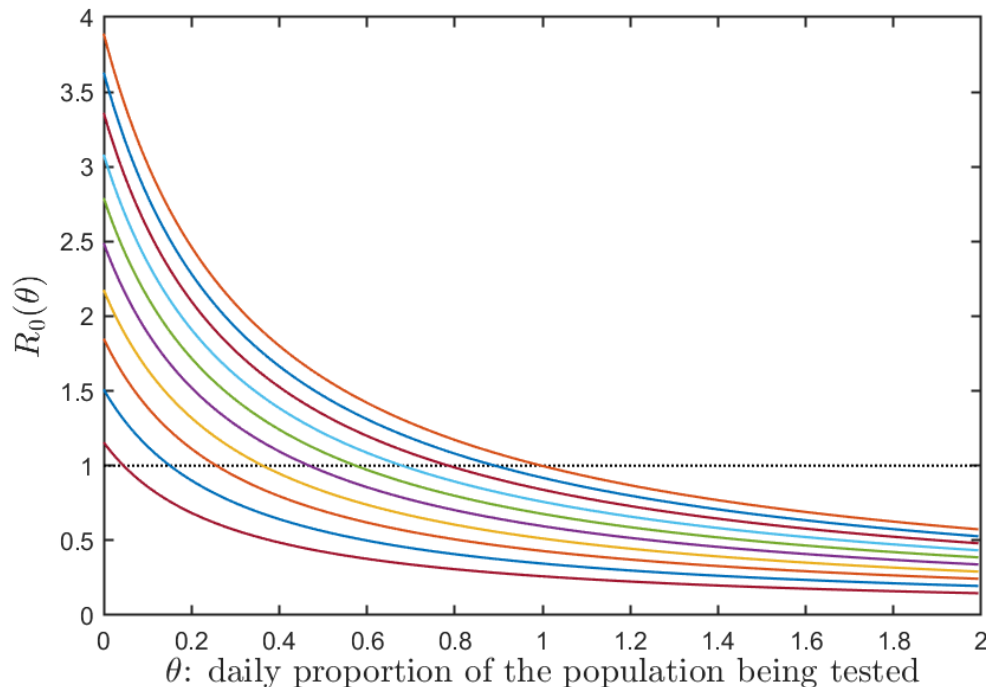
**test sensitivity 70%**

testing rate  $\theta$

### Corollary

Testing rate required for  $R_0(\theta) < 1$ :

$$\theta > \theta^* = \frac{1}{q(1-\delta)} \left[ \beta \left( \frac{(1-\mu)v^2}{\lambda} - 1 \right) - \gamma \right]$$



## Concluding remarks

### ~> A Covid-19 public policy proposal

- ▷ Rapid, cheap, at-home, saliva-based, paper tests
- ▷ Daily population-scale testing

### ~> Epidemiological models suggest

- ▷ Random testing is effective
  - ▶ Reduces  $R_0$
  - ▶ Decreases epidemic size
  - ▶ Its benefits are additive to other measures

## Concluding remarks

### ~> A Covid-19 public policy proposal

- ▷ Rapid, cheap, at-home, saliva-based, paper tests
- ▷ Daily population-scale testing

### ~> Epidemiological models suggest

- ▷ Random testing is effective
  - ▶ Reduces  $R_0$
  - ▶ Decreases epidemic size
  - ▶ Its benefits are additive to other measures
- ▷ Approximately daily testing *may* in fact be sufficient to suppress the pandemic

### ~> Precise mathematical results offer

- ▷ Strong, quantitative evidence of effectiveness
- ▷ Useful, conservative rules of thumb

## In recent news

### △ The rapid testing proposal is gaining traction

**Aug 27:** First FDA-approved rapid test: Abbott's \$5, 15-minute test  
White House announced \$750 million deal with Abbott

**Sept 1:** New rapid test by Roche-SD Biosensor partnership  
will be made available in Europe and the UK

**Sept 9:** UK PM announced “Operation Moonshot”, likely cost £100bn  
aiming for 10 million daily tests by spring

### △ Around the world

**Italy.** Approved 3-minute saliva test the “Daily Tampon”

**France.** New “antigénique rapide” test used by authorities

**Senegal.** UK-Senegal partnership developed  $\approx$  \$1 home antigen test

**India.** Authorities switching over to a rapid antigen test

*References*    [https://www.dpmms.cam.ac.uk/~ik355/PAPERS/Covid\\_talk\\_bib.pdf](https://www.dpmms.cam.ac.uk/~ik355/PAPERS/Covid_talk_bib.pdf)

*Slides*        [https://www.dpmms.cam.ac.uk/~ik355/PAPERS/Covid\\_talk\\_slides.pdf](https://www.dpmms.cam.ac.uk/~ik355/PAPERS/Covid_talk_slides.pdf)

*Our paper*    <https://www.dpmms.cam.ac.uk/~ik355/pubs.html>

## Technical assumptions for the configuration model

Assume initially  $N_I$  infected individuals s.t.  $N_{I,k}$  have degree  $k$  and  $N_S = N - N_I$  susceptible individuals s.t.  $N_{S,k}$  have degree  $k$

Assume  $N_I/N \rightarrow \mu$  and  $N_S/N \rightarrow (1 - \mu)$  for  $\mu \in (0, 1)$ , that  $\lambda \in (0, \infty)$  and that for all  $k$

$$\begin{aligned} \frac{N_{S,k}}{N_S} &\rightarrow p_k, & \frac{N_{I,k}}{N_I} &\rightarrow p_k \\ \sum_{k=0}^{\infty} k \frac{N_{S,k}}{N_S} &\rightarrow \sum_{k=0}^{\infty} k p_k = \lambda, & \sum_{k=0}^{\infty} k \frac{N_{I,k}}{N_I} &\rightarrow \sum_{k=0}^{\infty} k p_k = \lambda. \end{aligned}$$

Two last technical assumptions are required

For  $N_k =$  total no of individuals with degree  $k$ :

$$\max\{k ; N_{I,k} > 0\} = o(N) \quad \text{and} \quad \sum_{k=0}^{\infty} k^2 N_k = O(N)$$