

Branching Processes  
Dem260 Math Demog  
Spring 2020  
Lecture 8

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# Agenda for today

1. A viral example, simulating a branching process, our questions
2. Generating functions and extinction
3. Next time, finish up BP and get started on Fisher-Wright

# Stochastic modeling

- ▶ Until now, we've focused on the hidden structures of heterogeneity.
- ▶ Now, we're wwitching gears:
  - ▶ Stochastic not deterministic
  - ▶ In small populations, randomness matters. (Even when risks are homoteneous.)
  - ▶ Todayn: branching processes (“parents producing children”), next Fisher-Wright (“children choosing parents”), and then historical reconstruction from contemporary diversity (“coalescent”).

# Very brief history of Branching Processes

- ▶ Bienayme's lost notes
- ▶ Galton and Watson's (extinction of families)
- ▶ Genetics (survival of a mutant)
- ▶ The bomb (chain reactions)
- ▶ Anywhere “incipient dynamics” matter.

# Applicability to the Coronavirus? Yes and no.

- ▶ Perhaps the beginning, with first few cases.
- ▶ But once scale gets large, we'll see that deterministic dynamics take over.
- ▶ One lesson: beyond  $R_0$ .

## Our viral example

Here are the chances that the first carrier passes on the virus to  $k$  people?

$k$	$p_k$	digits
0	.3	0-2
1	.4	3-5
2	.3	6-9

- ▶ What is  $R_0$ , (aka  $m$ )? Calculate.
- ▶ Let's diagram one chance outcome, using my number "(xxx) xxx-9056"

9, 0, 5, 6

$k$	$p_k$	digits
0	.3	0-2
1	.4	3-5
2	.3	6-9

## Let's repeat

$k$	$p_k$	digits
0	.3	0-2
1	.4	3-5
2	.3	6-9

1. Let's try another diagram as a group. First name alphabetically (last cell digit)
2. Everyone do their last 4 digits



# What is a (Bienaymé-)Galton-Watson branching process?

$p_k$  Each individual in each generation reproduces independently, following same offspring distribution, with  $p_k$  as the probability of having  $k$  offspring.

$Z_n$  The size of the  $n$ 'th generation  $Z_n$ . ( $Z_1 \equiv 1$ )

$p_0 > 0$  Some non-zero probability of no children.

Variance None of the  $p_k$  are 1

## Some questions

- ▶ What is the chance  $d$  of eventual extinction (no “outbreak”)?
- ▶ Or, what is the distribution of surviving family sizes?
- ▶ What are the aggregate properties of many branching processes? (Mean growth, variance, time-paths, eventual size)?

# Galton's original question

**4001.** (Proposed by FRANCIS GALTON.)—A large nation, of whom we will only concern ourselves with the adult males,  $N$  in number, and who each bear separate surnames, colonise a district. Their law of population is such that, in each generation,  $a_0$  per cent. of the adult males have no male children who reach adult life;  $a_1$  have one such male child;  $a_2$  have

two; and so on, up to  $a_5$  who have five. Find (1) what proportion of the surnames will have become extinct after  $r$  generations; and (2) how many instances there will be of the same surname being held by  $m$  persons.

[The Proposer remarks that a general solution of this problem would be of much aid in certain rather important statistical enquiries, and that he finds it a laborious matter to work it out numerically, in even the simplest special cases, and to only a few generations. In reality, the generations would overlap and mix, but it is not necessary to suppose them otherwise than as occurring in successive steps.]

# A simulation

```
k = 0:2
p0 = .3; p1 = .3; p2 = .4;
p_k = c(p0, p1, p2)
Z1 = 1
set.seed(9)
(kids.of.Z1 = sample(x = k, size = Z1, replace = T, prob = p_k))
## [1] 2
(Z2 = sum(kids.of.Z1))
## [1] 2
(kids.of.Z2 = sample(x = k, size = Z2, replace = T, prob = p_k))
## [1] 2 2
(Z3 = sum(kids.of.Z2))
## [1] 4
(kids.of.Z3 = sample(x = k, size = Z3, replace = T, prob = p_k))
## [1] 2 1 2 2
(Z4 = sum(kids.of.Z3))
## [1] 7
```

Let's draw the tree.

## A simulation (2)

### A function

```
branch <- function(n_max = 30, pk = c(p0, p1, p2), Z1 = 1)
{
  Z.vec <- rep(NA, n_max)
  Z.vec[1] <- Z1
  for (i in 1:(n_max-1))
  {
    Z.vec[i+1] <- sum(sample(x = k,
                             size = Z.vec[i],
                             replace = T,
                             prob = p_k))
  }
  return(Z.vec)
}
```

```
set.seed(19); branch()
```

```
## [1] 1 2 2 4 5 2 2 3 2 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
```

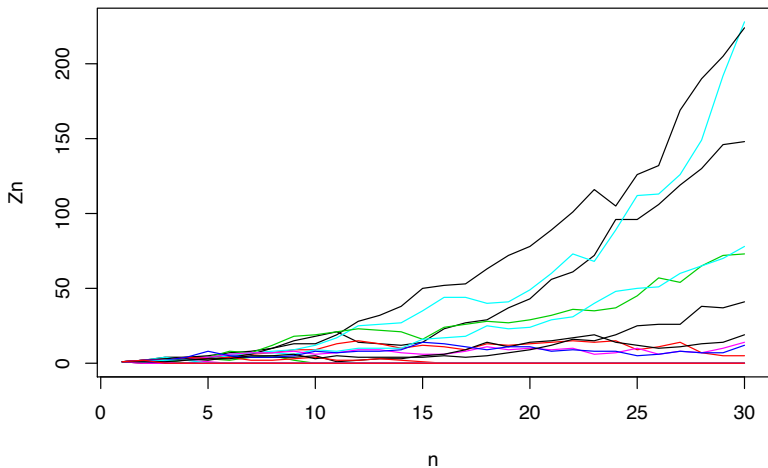
```
set.seed(99); branch()
```

```
## [1] 1 1 2 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
```

## A simulation (3)

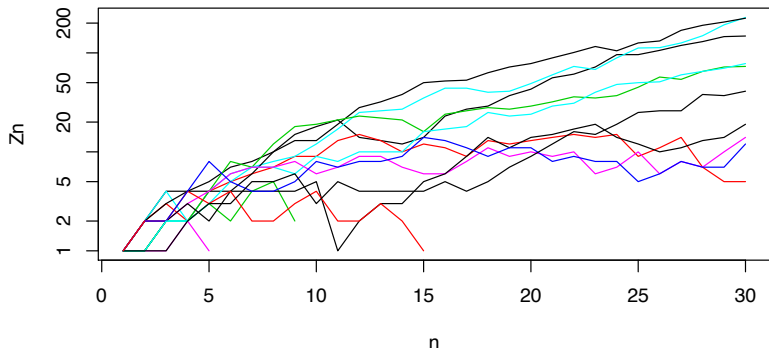
Let's see what happens with 20 trials (up to 30 generations)

```
n_trials = 20; n_gen = 30
Z.mat <- matrix(NA, n_trials, n_gen)
set.seed(131)
for (i in 1:n_trials)
  Z.mat[i,] <- branch(n_max = n_gen)
matplot(t(Z.mat),
         type = "l", lty = 1, ylab = "Zn", xlab = "n")
```



How many survive (out of 20)?

## log-scale

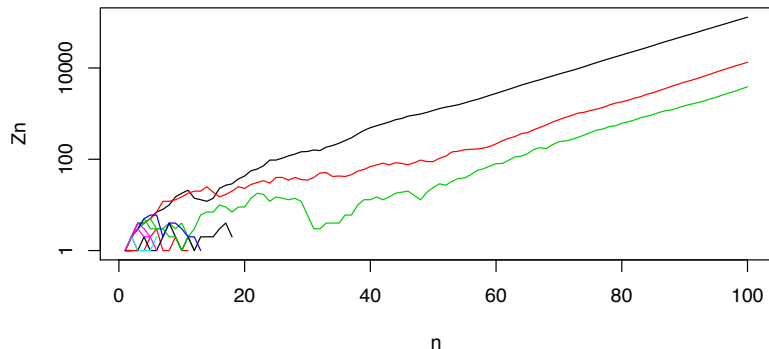


```
## surviving
## extinct survive
##      0.5      0.5
```

- How would you describe the time path of the surviving lines?



## Long term



What does this remind you of? (Hint: “Leslie”). (See Harris figure)

## “Extinction” vs “breakout”

- ▶ We see that in a super-critical ( $m > 1$ ) branching process, if a line can survive a few generations and reach a large enough size, it will grow exponentially.
- ▶ What happens if  $m < 1$ , if  $m = 1$ ? Discuss.

# Mathematical analysis

# The Probability Generating Function: Our mathematical tool

$$h(z) = p_0 + p_1z + p_2z^2 + \dots$$

The PGF “keeps book” on the probabilities. The chance of  $k$  is the coefficient on  $z^k$ .

Some interesting properties

$$h(0) =$$

$$h(1) =$$

$$h'(1) =$$

But the magic is next.

# The story of two brothers

A father has two sons. The probability generating function of their children combined is:

$$[h(z)]^2 = (p_0 + p_1z + p_2z^2) \times (p_0 + p_1z + p_2z^2)$$

Multiply out, and tell me the coefficients on  $z^0, z^1, \dots$

## Even more amazing

What is the probability generating function for the distribution of grandsons?

- ▶ A man has two sons, with probability  $p_2$ , so PGF in that case is  $p_2[h(z)]^2$ .
- ▶ But let's sum over all possible numbers of sons.

$$p_0 + p_1 h(z) + p_2 [h(z)]^2 + p_3 [h(z)]^3 + \dots$$

- ▶ Which is? (Hint: write a new argument for PGF)

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- ▶ Can show PGF for the  $n$ 'th generation is

$$h(h(h \dots n \text{ times } h(z))) = h_n(z)$$



In-class exercise: write out  $h_2(z) = h(h(z))$  for

$$h(z) = p_0 + p_1z + p_2z^2.$$

# Extinction

# Extinction: some generalities

“Extinction is forever.” So, the probability  $d_n$  of extinction *by* generation  $n$  can never decline over time. (Must it always rise?)

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Is non-extinction “forever”? If  $\lim_{n \rightarrow \infty} d_n = d(\infty) < 1$ , then this says there's a chance  $1 - d(\infty)$  of eternal persistence. We'll try to figure out more about what this means.

## Extinction: a recursive trick

If the probability of a female line going extinct in  $n$  generations is  $d_n$ , then this is equivalent to her daughter(s) line(s) going extinct in  $n - 1$  generations. With  $p_k$  chance of having  $k$  daughters, we have

$$d_n = p_0 + p_1 d_{n-1} + \text{What is next term in series?}$$

## Recursive extinction, continued

What can we do with

$$d_n = h(d_{n-1})?$$

Well, remember that  $d_n$  is non-decreasing, and that it's maximum can be no greater than 1.0. When  $d_n$  reaches it's limit, say  $d$ , we won't need generational subscripts,  $d$  will be constant, and will obey

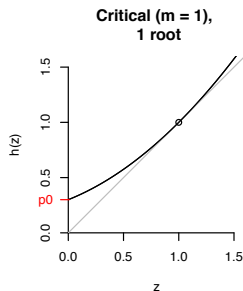
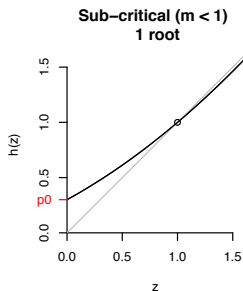
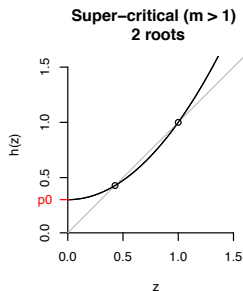
$$d = h(d)$$

Thus, an amazing result: the probability of ultimate extinction is when the argument equals the PGF of the argument.

Can  $d = 1$ , can  $d < 1$

1. Try  $d = 1$ . What happens?
2. If we were to find a solution less than 1.0, how would we interpret that?

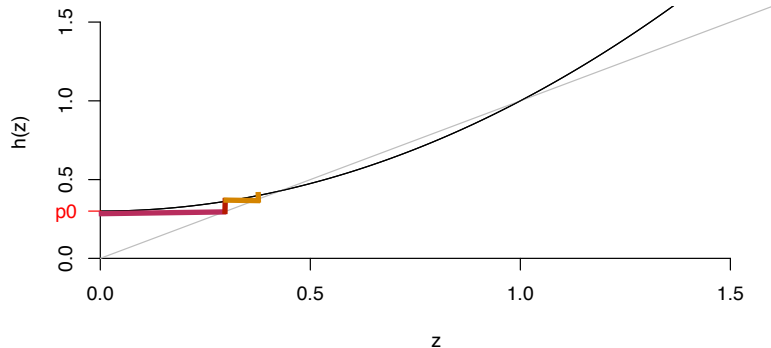
# Three cases



We can prove by answering: What is  $h'(1)$ ? What is  $h(0)$ ? Is  $h''(z) > 0$ ?



## A cobweb diagram (here a “staircase”)



Where is  $h(p_0), h(h(p_0)), h(h(h(p_0))), \dots$ ?

## So how do we actually get $d$ ?

Take the case where  $p_0 = .3$ ,  $p_1 = 0$ , and  $p_3 = .7$  (the one I just plotted).

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Take the case where  $p_0 = .3$ ,  $p_1 = 0$ , and  $p_3 = .7$  (the one I just plotted).

- ▶ Can do some algebra
- ▶ Or we can recursively iterate on the computer.

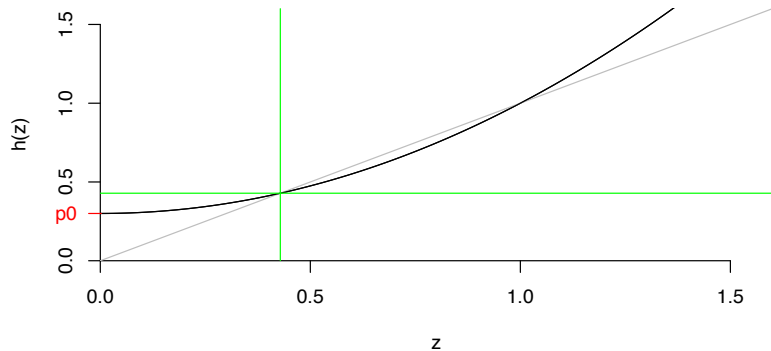
# Numerical recursion

```
pk = c(.3, .0, .7); names(pk) <- 0:2 ## our example
d <- pk["0"] # initial value
for (i in 1:20)
{
  d <- pk["0"] + pk["1"]*d + pk["2"]*d^2
  if (i %in% c(1,2,19,20))
    print(paste(i, d))
}

## [1] "1 0.363"
## [1] "2 0.3922383"
## [1] "19 0.428565882081349"
## [1] "20 0.428568100698915"
```

Did we get the right value?

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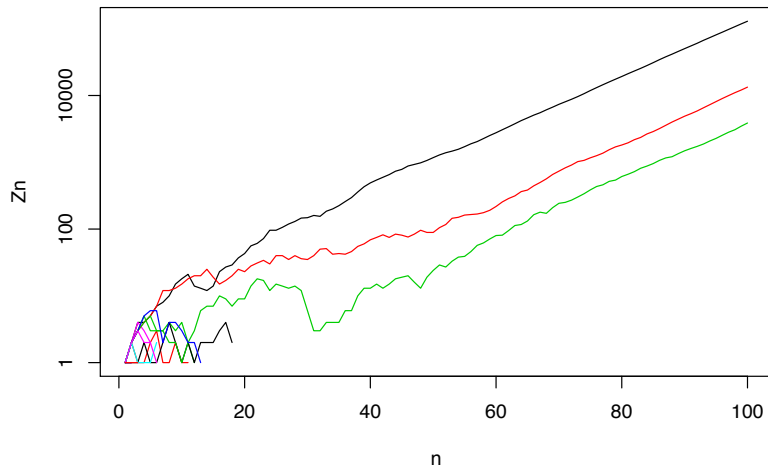


# Extinction and non-extinction revisited

- ▶ If  $m > 1$ , there exists  $d$  bigger than 0 and less than unity.
- ▶ This means there's some positive chance of extinction.
- ▶ But also some chance of never-extinction. (What form does never-extinction take?)



## Return to our simulation



Relevance to Corona virus?

# Good and bad set-ups for branching process

Good

Bad

- ▶ When offspring of 1 depends on offspring of other (e.g., brothers inheriting a farm)
- ▶ When resource constraints slow growth rates (e.g., Malthus: fertility of next gen depends on fertility of last; SIR model in disease spread)
- ▶ Analysis. PGF is powerful but still we often have to deal with listing all possibilities.
- ▶ Big populations – law of large numbers means randomness doesn't matter.

- ▶ Unrestricted growth (frontier, new disease, start of a reaction)
- ▶ A “null” model for understanding how apparent structure is just random outcomes. Families that die out didn't have to have low *NRR*. Just because most new viruses don't break out, doesn't mean they aren't potentially dangerous ( $R_0 \gg 1.0$ ).
- ▶ A model that corresponds our mental model of running a generative process forward. (cf. Fisher-Wright)

## Next time

- ▶ Do the PS exercises.
- ▶ We'll have BP part 2, next week. (means and variances of  $Z_n$ , the one tractable offspring distribution, time to extinction, etc.)
- ▶ I hope to have a mini-project with Mississippi last names.
- ▶ We may start Fisher-Wright. (Kennedy and I recommend Gillespie.)

# Questions?