

# Introduction to Demographic Heterogeneity

## Dem260 Math Demog

### Spring 2020

### Lecture 1

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# Welcome

- ▶ Quick overview of course
- ▶ Options for last part of semester
- ▶ Resources (lecture, notes, readings, videos)
- ▶ Practice (some standard math problems, some open ended questions, some simulations, and two applied investigations)

# Agenda for today

1. What demographic heterogeneity is (and isn't)
2. Dynamics of population growth with two sub-groups
  - \* Break
3. Keyfitz's result  $\bar{r}'(t) = \sigma_r^2(t)$ .
4. Ken's model of Poisson heterogeneity

# Part I. Conceptual Introduction

# 1. What is Demographic Heterogeneity?

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- ▶ If we see different outcomes (e.g., people dying at different ages), is this **Demographic Heterogeneity**? NO.
- ▶ Demographic heterogeneity = different rates for different folks.

*In a demographically heterogeneous population, people are of different types, with different type-specific rates.*

(These types can be discrete, with individuals being homogeneous within their type, or they can be continuous with possibly no individual having exactly the same risk as another.)

# 1. An example

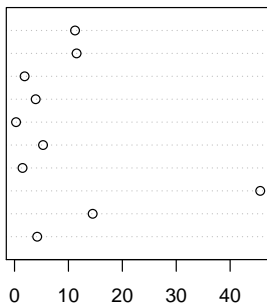
Let's draw 10 individuals from a homogeneous population and heterogeneous population.

```
## Homogeneous hazard of 1/10
set.seed(13)
x.homo <- rexp(10, rate = 1/10)
## Heterogeneous hazard (half 1/6 and half 1/13)
## Note: I didn't pick these particular numbers for any sp
x.hetero <- c(rexp(5, rate = 1/6),
              rexp(5, rate = 1/13))
par(mfrow = c(2,1))
dotchart(x.homo, main = "homogenous variation")
dotchart(x.hetero, main = "heterogeneous variation")
```

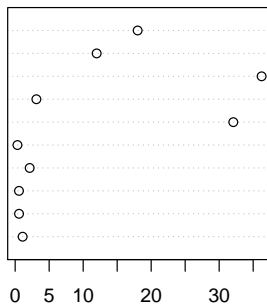


# 1. An example, continued

**homogenous variation**



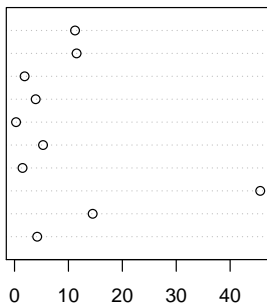
**heterogeneous variation**



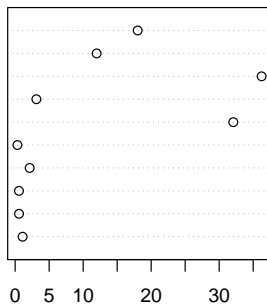
► Can you tell which is which?

# 1. An example, continued

**homogenous variation**



**heterogeneous variation**



- ▶ Can you tell which is which?
- ▶ Would we expect to see a difference if we increased sample?

# 1. Components of variation

- ▶ **Homogeneous:** Chance only
- ▶ **Heterogeneous:** Chance + group variation in risk

# 1. Analogies

Social inequality: equal opportunity vs. equal outcomes

Analysis of variance

total variance = within group + between group

Statistical models  $y = a + bx + \epsilon$

# What's new? Dynamics.

Heterogeneous populations evolve differently. Aggregates  $\neq$  Individuals

- ▶ Rates of growth (or decline)
- ▶ Changes over time or age or duration
- ▶ The trajectory of even the **average** individual differs from population average
- ▶ Relative positions, change of groups, may be misleading.

# 1. Terminology

- ▶ Heterogeneity
- ▶ Unobserved heterogeneity
- ▶ Selection
- ▶ Selective survival
- ▶ Other terms?

# 1. Heterogeneity at work? Black-White Crossover

# 1. Heterogeneity at work? Mortality Plateaus



# 1. Big Caveat: Fundamental Unidentifiability

Same data of  $N$  observations

- ▶  $N$  draws from 1 distribution
- ▶ 1 draw from  $N$  distributions
- ▶ Something in-between

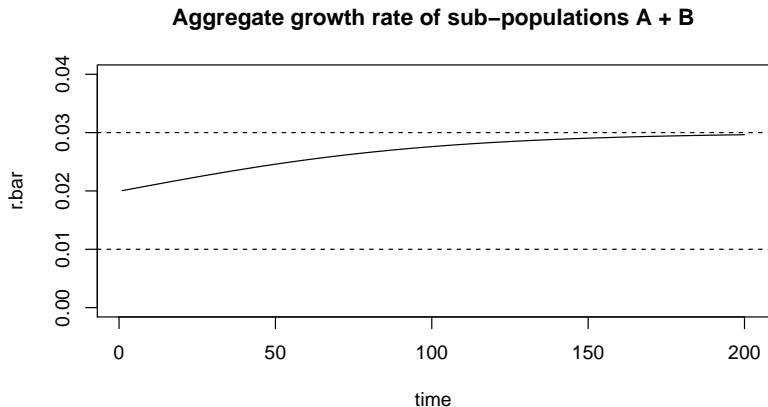
Abel (66) and Beth (76) example.

## 2. A 2nd example: Exponential growth, two countries

Two countries start equal size, but grow at different rates. What happens to aggregate growth rate?

```
rA = .03 ## growth rate of A
rB = .01 ## growth rate of A
KA = 100 ## starting pop size
KB = 100
t = 0:200
KA.t = KA*exp(rA*t) ## exp growth of A
KB.t = KB*exp(rB*t) ## exp growth of B
K <- KA.t + KB.t ## combined pop
r.bar = diff(log(K)) ## growth rate
plot(t[-1], r.bar, type = "l", ylim = c(0, 0.04),
      ylab = "r.bar", xlab = "time")
abline(h = c(rA, rB), lty = 2)
title("Aggregate growth rate of sub-populations A + B")
```

## 2. A 2nd example: Exponential growth, two countries



- ▶ What determines growth rate?
- ▶ How does it change over time?
- ▶ Does the process converge?

## 2. Some more examples

1. Differential, constant mortality ( $\mu_A = .03$ ;  $\mu_B = .01$ )
2. Differential, *time-varying* mortality or growth.
3. “Movers and Stayers” (Migration)
4. “Movers and Stayers” (Marriage)

## 2. Even more examples

1. Fecundity: aging or heterogeneity?
2. Divorce: duration or heterogeneity?
3. Duration of unemployment: duration or heterogeneity?
4. Recidivism by time out of prison
5. What else?

## 2. Fun with App

[https://shiny.demog.berkeley.edu/josh/het\\_ruse/](https://shiny.demog.berkeley.edu/josh/het_ruse/)

- ▶ Can you create a plateau?
- ▶ Can you create a crossover?
- ▶ Can you get aggregate rate to decline?
- ▶ Anything else?

Cookie break here?

## Part II. Formal Analysis

1. Keyfitz result
2. Ken's Poisson-Exponential Model
3. Keyfitz USA-Mexico example



### 3. Keyfitz result

$$\frac{d}{dt}\bar{r}(t) = \sigma_r^2(t)$$

When group-specific growth rates are constant, the rate of change of the aggregate growth rate equals the variance of the growth rates.

### 3. Derivation

By definition,

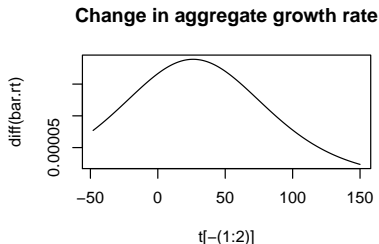
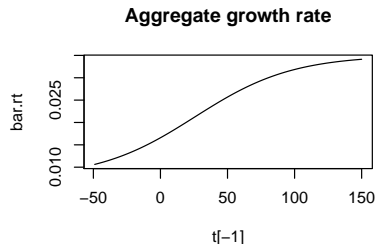
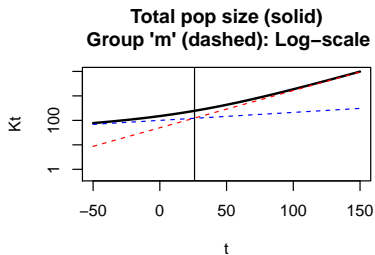
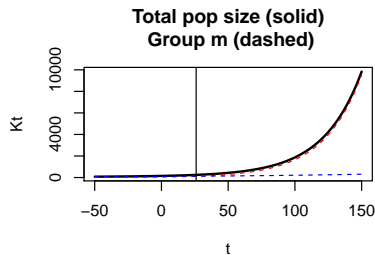
$$\bar{K}(t) = \sum_i K_i(t) = \sum_i K_i e^{r_i t}$$

and

$$\bar{r}(t) = \frac{\frac{d}{dt} \bar{K}(t)}{\bar{K}(t)}.$$

Let's take derivatives and simplify, recalling definition of variance.

### 3. US-Mexico Example



Appears that pop sizes maximizes variance. (Check at home?)

# US-Mexico Example

```
rm = 3.5/100
ru = .75/100
Km = 50
Ku = 100
t <- -50:150 ## go back in time to see rise and fall of variance
Kt = Km * exp(t*rm) + Ku * exp(t*ru)
bar.rt <- diff(log(Kt))
par(mfrow = c(2,2))
plot(t, Kt, lwd = 2, type = 'l')
title('Total pop size (solid)\n Group m (dashed)')
lines(t, Km * exp(t*rm), lty = 2, col = "red")
lines(t, Ku * exp(t*ru), lty = 2, col = "blue")
my.v = 26
abline(v = my.v)
plot(t, Kt, lwd = 2, type = 'l', log = 'y', ylim = c(.5, max(Kt)))
lines(t, Km * exp(t*rm), lty = 2, col = "red")
lines(t, Ku * exp(t*ru), lty = 2, col = "blue")
abline(v = my.v)
title("Total pop size (solid)\n Group 'm' (dashed): Log-scale")
plot(t[-1], bar.rt, type = 'l', main = 'Aggregate growth rate')
plot(t[-(1:2)], diff(bar.rt), type = 'l',
main = 'Change in aggregate growth rate')
```

Appears that pop sizes maximizes variance. (Check at home?)

### 3. Commentary on Keyfitz result

- ▶ Growth rates in heterogeneous populations start at pop average and then increase.
- ▶ Heterogeneity pop growth
- ▶ We will extend to cover non-constant growth

But

- ▶ Doesn't tell us how much bigger  $\bar{K}(t)$  is projection using constant aggregate rate  $\bar{r}(0)$ .
- ▶ Doesn't give us a formula for time path of aggregate  $\bar{K}(t)$  or  $\bar{r}(t)$

Note: our homework will try to address some of this using Taylor approximation.

## 4. The Origin of Ken's Poisson-Exponential Model

Given a world with many sub-populations, each growing exponentially at their own rate, what can we say about the time-path of world population growth?

## 4. Ken's Poisson-Exponential Model

From an email:

*Josh asks: Suppose we have a discrete mix of subpopulations growing at different intrinsic rates  $r$  whose maximum is  $r_0$ . Is there a handy approximation for the growth path of the aggregate populations?*

*The assumption of a discrete mix is essential here. Otherwise Tauberian theorems apply and, with a vanishingly small portion of the population close to the maximum growth rate, we do not obtain long-run exponential growth.*

*I recommend modeling the discrete distribution of growth rates as a mixture of Poisson distributions.*

## 4. Poisson-based model

We are considering

$$\bar{K}(t) = \sum_i e^{r_i t} K_i(0). \quad (1)$$

Ken suggests

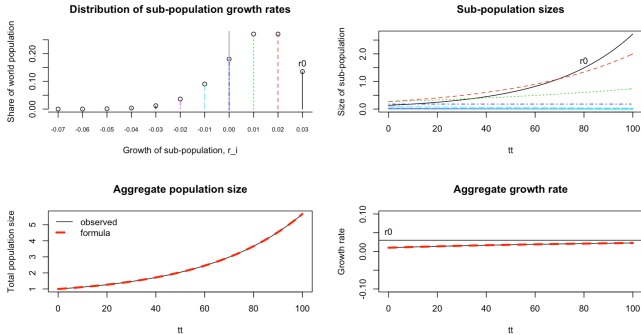
$$r_i = r_0 - s(\lambda) \cdot a, \quad (2)$$

- $r_0$  growth rate of the fastest growing sub-population
- $s$  a non-negative Poisson distributed integer
- $\lambda$  the parameter of the Poisson distribution (also it's mean and variance)
- $a$  gap between adjacent growth rates.

Example: sub-populations have growth rates 3, 2, 1, 0, -1, ... percent, then  $r_0 = 0.03$  and  $a = 0.01$ . Sizes of sub-pops determined by Poisson dis'n



Figure: A simulation of heterogeneous growth



## 4. Closed-form result

$$K(t) = K(0)e^{r_0 t} e^{-\lambda(1-e^{-at})}.$$

To derive:

- ▶ Write out mixture to get

$$K(t) = K(0)e^{r_0 t} \sum_i e^{-sat} f(s)$$

- ▶ Substitute for  $f(s)$ : *Pois*  $\frac{\lambda^s e^{-\lambda}}{s!}$
- ▶ Recognize that our mixture contains the series representation of  $e^{-at}$
- ▶ Done

## 4. Interpretation

$$K(t) = K(0)e^{r_0 t}e^{-\lambda(1-e^{-at})}.$$

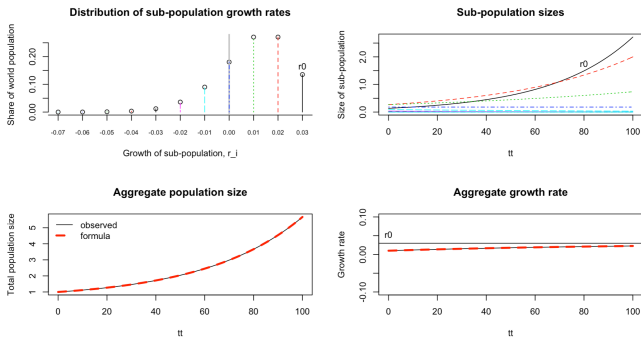
- ▶ Dominant term contains the maximum population growth rate  $r_0$ ,
- ▶ Second term gives the diminishing effect of the sub-populations with smaller population growth rates over time.

## 4. Some further analysis

What is the closed-form expression for  $\bar{r}(t)$ ?

## 4. And the formula works!

Figure: A simulation of heterogeneous growth



## 4. Some commentary

- ▶ Poisson and Exponential “fit”
- ▶ We'll this complementarity again (e.g., with Gamma)
- ▶ Tractable models are super powerful for enhancing our understanding.
- ▶ But be careful. Avoid extremes: the model is right/wrong.

# A BIG caveat

Are disaggregated models necessarily better?

# A BIG caveat

Are disaggregated models necessarily better? Some potential problems:

- ▶ Aggregate constraints?
- ▶ Interacting sub-populations?
- ▶ Illusion of precision?



## Next time ...

- ▶ Mortality
- ▶ Frailty (Multiplicative and fixed)
- ▶ Gamma