

Mortality Heterogeneity: Multiplicative Fixed  
Frailty  
Dem260 Math Demog  
Spring 2020  
Lecture 2

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# Agenda for today

1. Conclusions from last time
2. Review of Mortality Mathematics
3. Multiplicative-fixed-frailty and alternatives to it.
4. Population Survival and Hazards under fixed frailty
  - \* Break
5. Gamma frailty

# Conclusions from last time

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## Conclusions from last time

- ▶ Heterogeneity as variation in risk (not just outcome)
- ▶ Constantly growing parts  $\neq$  constantly growth whole
- ▶ Keyfitz result: Change in growth rate = variance of growth rates
- ▶ Poisson growth gives us a closed-form solution.

# Review of mortality mathematics

$\ell(x)$  or  $S(x)$  probability of survival to age  $x$

$\mu(x)$  or  $h(x)$  hazard rate at age  $x$  (“minus the exponential rate of change in survival”)

Let's treat  $\mu$  as a definition.

$$\mu(x) \equiv -\frac{d}{dx} \log \ell(x)$$

Can anti-differentiate (integrate) to solve for survival:

$$\ell(x) = s(x) = e^{-\int_0^x \mu(a) da}$$

## Application: what is $\ell'(x)$ ?

- ▶ in words?
- ▶ taking derivative of  $\ell(x)$
- ▶ interpretation



## Two special cases

- ▶ Constant hazards  $\mu(x) = \mu$ . What's  $\ell(x)$ ?
- ▶ Gompertz hazards  $\mu(x) = ae^{bx}$ . What's  $\ell(x)$ ?

# Extending Keyfitz to mortality

$$\frac{d}{dx}\bar{\mu}(x) = \text{average rate of change} - \sigma_{\mu}^2$$

What is  $\bar{\mu}$ ? It's a weighted average:

$$\bar{\mu}(x) = \frac{\int \mu(x|z)\ell(x|z)p(z) dz}{\int \ell(x|z)p(z) dz}$$

To derive Keyfitz extension, differentiate with respect to age  $x$ .  
(See V&M eq (13)). A good exercise.

# Multiplicative Fixed frailty

For individual  $i$ ,

$$\mu_i(x) = z_i \mu_0(x).$$

$z_i$  “frailty” of the  $i$ th individual. (Usually thought of as a random variable with mean 1.)

$\mu_0(x)$  “Baseline hazard” schedule. (Also, the schedule of a person with  $z = 1$ ).

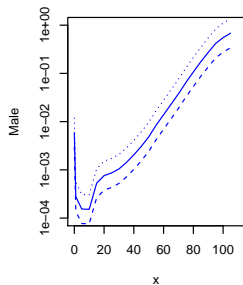
# What are some alternatives?

Let's think of at least three.

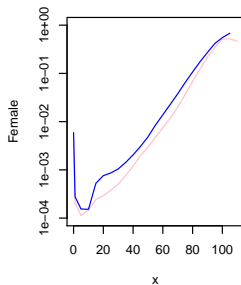
$(\beta, i, \Delta)$

Which look like multiplicative fixed frailty?

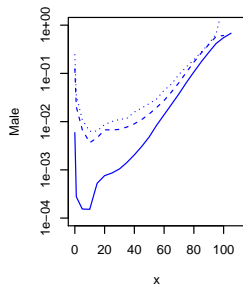
Male (5Mx) 1990, x 0.5 and x 1.5



By Sex (1990–1994)



Male 1790, 1890, 1990



## Part II. Results from Fixed Frailty

## 2. A simulation

### Our questions

- ▶ How do we do a micro-simulation, with individuals?
- ▶ How does fixed frailty fit in?
- ▶ How do we compute pop survival, hazards, etc.
- ▶ How does life table of heterogeneous pop differ from baseline?

# Homework

- ▶ Look at frailty, distribution and mean frailty of survivors.
- ▶ How does life expectancy  $e(x)$  differ in pop and baseline?



## Let's derive pop survival (Note: $\bar{s} = \bar{\ell}$ )

Pop survival will be a weighted average of group survival curves

$$\bar{s}(x) = \frac{p(z_1)s_1(x) + p(z_2)s_2(x) + \dots}{p(z_1) + p(z_2) + \dots}$$

With continuous  $z$  (what are limits of integration?)

$$\bar{s}(x) = \int s(x|z)p(z) dz$$

Under Multiplicative Fixed Frailty use

$$\mu(x|z) = z\mu_0(x)$$

to derive

$$\bar{s}(x) = \int s_0(x)^z p(z) dz.$$

## Now population hazards (stepping stones)

Definition of hazards:

$$\bar{\mu}(x) = -\frac{d}{dx} \log \bar{s}(x)$$

$$\bar{\mu}(x) = \mu_0(x) \frac{\int z s_0(x)^z p(z) dz}{\int s_0(x)^z p(z) dz}$$

$$\bar{\mu}(x) = \mu_0(x) \bar{z}(x)$$

Let's fill in steps.

## Rodriguez question

Why isn't population hazard a (simple) average of individual hazards?

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Why isn't population hazard a (simple) average of individual hazards?

Answer: selected survival means that the distribution of frailty at age  $x$  differs from the starting frailty distribution at age 0.

# The rate of increase in hazards

(AKA “LAR: Lifetable Aging Rate”)

$$\beta(x) = \frac{d}{dx} \log \mu(x)$$

Example:

What is  $\beta(x)$  for Gompertz:  $\mu(x) = ae^{bx}$ ?

# Vaupel's result

$$\bar{\beta}(x) = \beta_0(x) - \bar{\mu}(x)CV_z^2(x)$$

The same kind of tedious derivation ... but conclusion is more interesting.

- ▶ Hazards rise less slowly in pop than in baseline
- ▶ If pop hazards plateau, then  $\bar{\beta}(x) = 0$
- ▶ Two special cases
  - ▶ Homogeneous pop and plateau in baseline
  - ▶ Gompertz and constant  $CV_z$  (e.g., from Gamma)

## Part II. Introduction to Gamma Frailty

# Gamma agenda (today)

- ▶ What do we want in a frailty distribution?
- ▶ What's the Gamma?
- ▶ Last math: closed form pop survival



# What do we want in a frailty distribution?

# What do we want in a frailty distribution?

- ▶ positive?
- ▶ a single dimension summarizing multiple factors? (Normal?)
- ▶ flexible?
- ▶ tractable?

# The Gamma distribution

$$p(z|k, \lambda) = \frac{\lambda^k}{\Gamma(k)} z^{k-1} e^{-\lambda z}$$

- $z$  the random variable representing frailty
- $k, \lambda$  parameters
- $\Gamma(k)$  A normalizing constant.

# Gamma in R

Our parameterization:

Mean:  $k/\lambda$

Variance:  $k/\lambda^2$

```
## with k and lambda
k = 3; lambda = 6
x <- rgamma(10000, shape = k, rate= lambda)
mean(x)

## [1] 0.4985961

sd(x)

## [1] 0.2892092
```

## Alternate parameterization

```
## with mean 1, sigma.sq
sigma.sq <- .25
z <- rgamma(10000, shape = 1/sigma.sq, rate = 1/sigma.sq)
mean(z)

## [1] 1.007105

var(z)

## [1] 0.2508519
```

# Population Survival of Gamma Frailty

Big picture

$$\bar{s}(x) = \int s_0(x)^z p(z) dz$$

Or, using our definition of survival,

$$\bar{s}(x) = \int e^{-zH_0(x)} p(z) dz$$

## “completing the gamma”

$$\bar{s}(x) = \int e^{-zH_0(x)} \frac{\lambda^k}{\Gamma(k)} z^{k-1} e^{-\lambda z} dz$$

Rearranging,

$$\bar{s}(x) = \lambda^k \int \frac{1}{\Gamma(k)} z^{k-1} e^{-z(H_0(x)+\lambda)} dz$$

Integral is like a  $\text{Gamma}(z|k, H_0(x) + \lambda)$ , but missing something.  
What?

## Our Result

$$\bar{S}(x) = \frac{\lambda^k}{[H_0(x) + \lambda]^k}$$

If mean = 1.0, then we can let  $\lambda = k = 1/\sigma^2$ ,

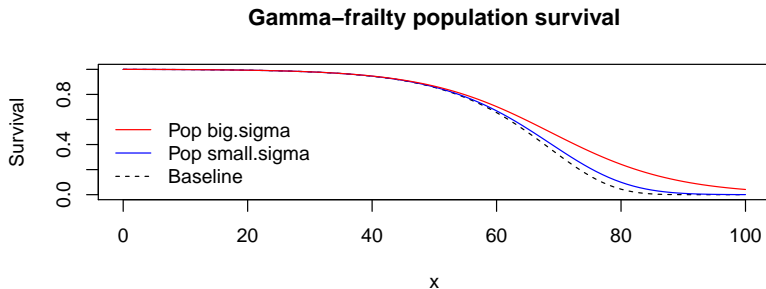
$$\bar{S}(x) = \frac{1/\sigma^2}{(H_0(x) + 1/\sigma^2)^{1/\sigma^2}} = \frac{1}{(1 + \sigma^2 H_0(x))^{1/\sigma^2}}$$



# Interpreting Gamma-frailty survival

$$\bar{S}(x) = \frac{1}{(1 + \sigma^2 H_0(x))^{1/\sigma^2}}$$

- ▶ Older ages, smaller survival.
- ▶ Variance not so clear



# Conclusions

- ▶ What is multiplicative fixed frailty?
- ▶ Analytical expressions for aggregate/pop/unconditional survival and hazards
- ▶ Extension of Keyfitz result
- ▶ Gamma intro

# Next time

We'll see the power of the Gamma for

- ▶ Explicit expression for hazards
- ▶ Inversion formula (baseline from aggregate)
- ▶ Plateaus
- ▶ Cross-overs
- ▶ Inequality and change over time