



**UNIVERSITÀ  
DEGLI STUDI  
DI TRIESTE**



Dipartimento di scienze economiche,  
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# Bayesian Statistics

## Multiple parameter models

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## Example: Kidney cancer death rates

For the 3130 counties of the US we know the death counts  $y_j$  in 1980-89 and populations  $n_j$ , death rates are computed as  $r_j = y_j/n_j$ .

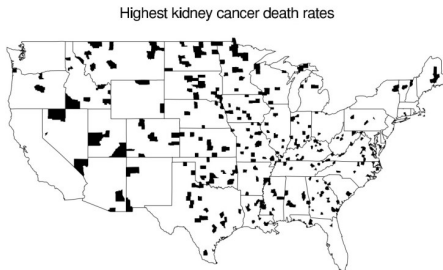


Figure 2.6 *The counties of the United States with the highest 10% age-standardized death rates for cancer of kidney/ureter for U.S. white males, 1980–1989. Why are most of the shaded counties in the middle of the country? See Section 2.7 for discussion.*

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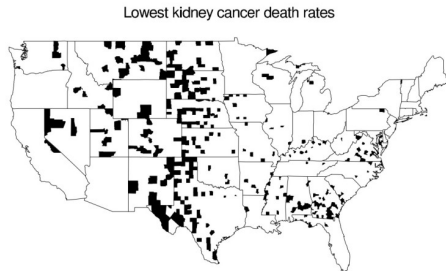


Figure 2.7 *The counties of the United States with the lowest 10% age-standardized death rates for cancer of kidney/ureter for U.S. white males, 1980–1989. Surprisingly, the pattern is somewhat similar to the map of the highest rates, shown in Figure 2.6.*

## Pattern on the map

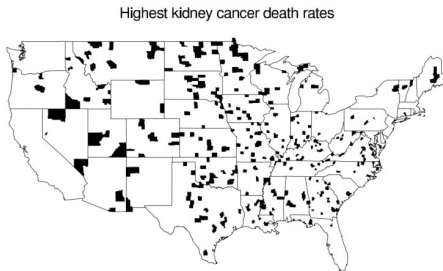


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Many of the counties in the Great Plains in the middle of the country, but relatively few counties near the coasts, are shaded.

## Pattern on the map

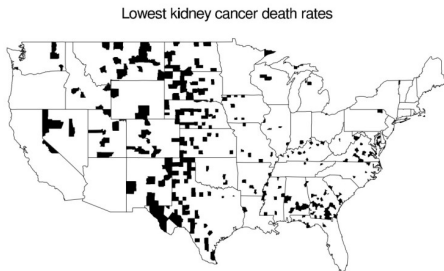


Figure 2.7 *The counties of the United States with the lowest 10% age-standardized death rates for cancer of kidney/ureter for U.S. white males, 1980–1989. Surprisingly, the pattern is somewhat similar to the map of the highest rates, shown in Figure 2.6.*

The pattern of the counties with the lowest rate is similar, the point is that this pattern is due to the size of the counties.

## Pattern on the map, explained

Kidney cancer is rare, the overall death rate is  $4.65 \times 10^{-5}$ .



Population of counties can be as low as 1000 up to 100s of thousand, in a scarcely populated county, let  $n_j = 1000$  one has, in 10 years,  $E(Y_j) = 0.4$  and if we assume a Poisson distribution for  $Y_j$

- $P(Y_j = 0) = 0.63$ 
  - these will have a rate of 0 and so will be among the lowest
- $P(Y_j = 1) = 0.29$ 
  - these will have a rate of 1/10000 and so will be among the highest



More populated countries are unlikely to have such extreme outcomes.

## County Model

We assume

$$y_j \sim \text{Poisson}(10n_j\theta_j)$$

we also assume

$$\theta_j \sim \text{Gamma}(20, 430\,000)$$

(implying  $E(\theta_j) = 4.65 \times 10^{-5}$  and  $\sqrt{V(\theta_j)} = 1.04 \times 10^{-5}$  so the posterior distribution is

$$\theta_j|y_j \sim \text{Gamma}(\alpha = 20 + y_j, \beta = 430\,000 + 10n_j)$$

with mean and variance

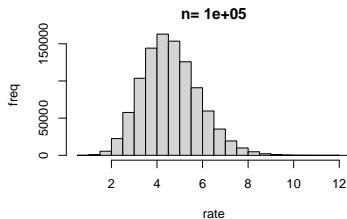
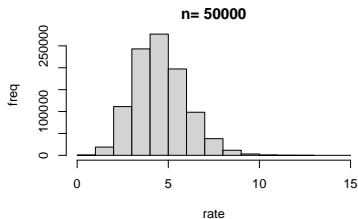
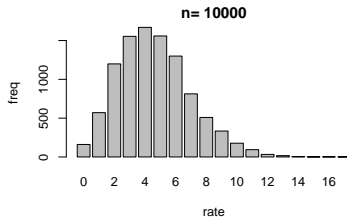
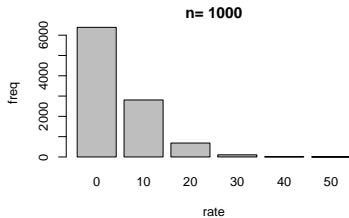
$$E(\theta_j|y_j) = \frac{20 + y_j}{430\,000 + 10n_j}; \quad \sqrt{V(\theta_j|y_j)} = \sqrt{\frac{20 + y_j}{(430\,000 + 10n_j)^2}}$$

## Results for some hypothetical countries

y	n	rate ( $\times 10^{-5}$ )	$\alpha$	$\beta$	mean ( $\times 10^{-5}$ )	sd ( $\times 10^{-5}$ )
0	1000	0.00	20	440000	4.55	1.02
1	1000	10.00	21	440000	4.77	1.04
2	1000	20.00	22	440000	5.00	1.07
0	50000	0.00	20	930000	2.15	0.48
1	50000	0.20	21	930000	2.26	0.49
2	50000	0.40	22	930000	2.37	0.50
50	50000	10.00	70	930000	7.53	0.90
100	50000	20.00	120	930000	12.90	1.18
0	1000000	0.00	20	10430000	0.19	0.04
1	1000000	0.01	21	10430000	0.20	0.04
2	1000000	0.02	22	10430000	0.21	0.04
1000	1000000	10.00	1020	10430000	9.78	0.31
2000	1000000	20.00	2020	10430000	19.37	0.43



# Predictive distribution of the observed rate



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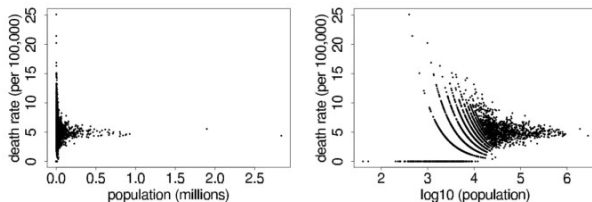


Figure 2.8 (a) Kidney cancer death rates  $y_j/(10n_j)$  vs. population size  $n_j$ . (b) Replotted on the scale of  $\log_{10}$  population to see the data more clearly. The patterns come from the discreteness of the data ( $n_j = 0, 1, 2, \dots$ ).

# Predictive distribution of the observed rate

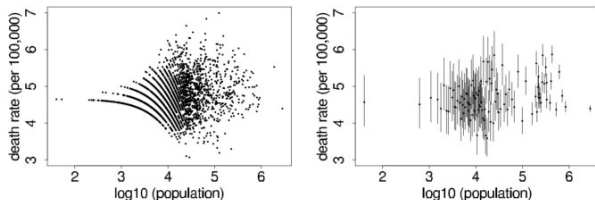


Figure 2.9 (a) Bayes-estimated posterior mean kidney cancer death rates,  $E(\theta_j|y_j) = \frac{20+y_j}{430000+10n_j}$  vs. logarithm of population size  $n_j$ , the 3071 counties in the U.S. (b) Posterior medians and 50% intervals for  $\theta_j$  for a sample of 100 counties  $j$ . The scales on the y-axes differ from the plots in Figure 2.8b.