ADVANCED BAYESIAN MODELING

- Grouping and Regression Coefficients

Observations are divided into J groups, with scalar responses

$$y_{ij}$$
 = response from obs. i in group j $i = 1, \ldots, n_j$ $j = 1, \ldots, J$

The regression has K explanatory variables (perhaps including an intercept). For observation i in group j,

$$X_{ij} = (x_{ij,1}, \dots, x_{ij,K})$$

Note: The explanatory variables must *not* include indicators for the J groups!

Ordinarily, linear regression would be

$$E(y_{ij} \mid \theta, X_{ij}) = X_{ij}\beta = \beta_1 x_{ij,1} + \dots + \beta_K x_{ij,K}$$

Now instead allow coefficients to vary by group:

$$E(y_{ij} | \theta, X_{ij}) = X_{ij}\beta^{(j)} = \beta_1^{(j)}x_{ij,1} + \dots + \beta_K^{(j)}x_{ij,K}$$

where
$$eta^{(j)} = egin{pmatrix} eta_1^{(j)} \ dots \ eta_K^{(j)} \end{pmatrix}$$

.

Example: Rat Growth Data

$$y_{ij}$$
 = weight of rat j at age x_i $i = 1, \ldots, 5$ $j = 1, \ldots, 30$

Ages happened to be same for all rats:

$$x_1=8$$
 days $\qquad x_2=15$ days $\qquad x_3=22$ days $\qquad x_4=29$ days $\qquad x_5=36$ days

(but this is not necessary for the methods to work)

Data Source: Gelfand, A.E., Hills, S.E., Racine-Poon, A., and Smith, A.F.M. (1990). Illustration of Bayesian inference in normal data models using Gibbs sampling. *Journal of the American Statistical Association*, **85**, 972–985.

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At early ages, growth of a rat is well modeled by a straight line:

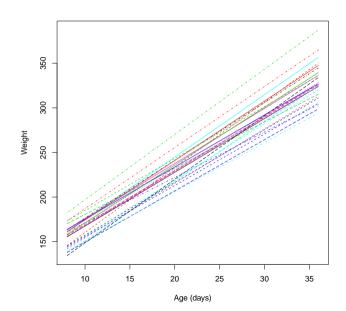
weight
$$\approx \beta_1 + \beta_2 \times age$$

But different rats could have

- ▶ Different intercepts β_1 : Some weigh more than others at birth.
- ▶ Different slopes β_2 : Some grow faster than others.

Individual ordinary least squares fitted simple linear regression lines (weight vs. age) for the 30 rats

Note: Some rats consistently weigh more, and some grow faster.



Thus, consider regression with coefficients allowed to vary by rat (group):

$$E(y_{ij} | \theta, X_{ij}) = \beta_1^{(j)} + \beta_2^{(j)} x_i$$

where

$$y_{ij}$$
 = weight of rat j at age x_i

Note: In earlier notation,

$$x_{ij,1} = 1$$
 $x_{ij,2} = x_i$