

model	data	initial values
{	{	{
}	}	}
	matrices vectors scalars	all parameters need to be initialized

we want to model $\left\{ \begin{array}{l} Y_i = \beta_0 + \beta_1 X_i + \epsilon_i, \quad i=1, \dots, N_{SUB} \\ \epsilon_i \sim N(0, \sigma_\epsilon^2) \end{array} \right.$

```

model
{ # Begin model
  for (i in 1:NSUB)
  { Y[i] ~ dnorm(m[i], tau_eps)
    m[i] ← beta0 + beta1 * X[i]
  }
  beta0 ~ dnorm(0, 1.0E-6)
  beta1 ~ dnorm(0, 1.0E-6)
  tau_eps ~ dgamma(.001, .001)
} # end model

```

Annotations:

- Arrow from τ_{ϵ} to $1/\sigma_\epsilon^2$ with label "precision = $1/\sigma_\epsilon^2$ "
- Arrow from $X[i]$ to a box labeled "data"
- Arrow from β_0 to "auxiliary variable"
- Arrows from β_1 to "parameters"

Y[]

X[]

2

.1

1

1

3

0

4

2

5

3

END

list(beta0 = 0, beta1 = 0, tau_eps = \$)

the sup priors
are compatible
w/ model

Warnings:

WINBUGS is
powerful!

So use it
with responsibility.

Will run even
if not identifiable!

WINBUGS Documentation:

③

- 1) The WINBUGS manual
- 2) Ciprian Crainiceanu's website
- 3) Bayesian Statistical Modeling, Peter Congdon
- 4) MCMC in practice Spiegelhalter et al.
- 5) JSS → Bayesian analysis for P-splines, Crainiceanu

Tricks & Tips (not exhaustive)

1. Nested indexing

Y_{ij} → Geomarker for subject i at visit j

X_{ij} → BMI ... " "

$$Y_{ij} = \beta_0 + b_i + \beta_1 X_{ij} + \epsilon_{ij}$$

$$b_i \sim N(0, \sigma_b^2)$$

$$\epsilon_{ij} \sim N(0, \sigma_\epsilon^2)$$

Need to rearrange the data ...

First step: restructure the data

Y	X	subject	"unbalanced design"
2	0	1	
3	2	1	
2	2	1	
1	3	2	
24	1	2	
3	2	3	
⋮	⋮		
END	EMP		

model

{ # begin model

for (i in 1:NBBS)

{ $Y[i] \sim \text{dnorm}(m[i], \text{tau-eps})$

$m[i] \leftarrow \text{beta0} + \text{beta1} * X[i]$

}

for (R in 1:NSUB)

{ $b[R] \sim \text{dnorm}(0, \text{tau-b})$ }

smoothing parameter

beta0 ~ same as before

beta1 ~ same as before

tau-eps ~ same as before

tau-b ~ same as tau-eps

} # end model

2nd trick

$$\begin{cases} \beta_0 + b_i \\ b_i \sim N(0, \sigma_b^2) \end{cases} \Leftrightarrow b_i \sim N(\beta_0, \sigma_b^2)$$

Hierarchical centering

mathematically equivalent
not WINBUGS technically equivalent

Hierarchical centering
take β_0 out of $w[i]$ statement
in previous model block, and
put β_0 in for 0 in $\text{dnorm}(\beta_0, \tau_0)$

1. convergence of chains
2. mixing
3. plot your priors and posteriors
→ you can see if parameters are learning from data

if there's no change between prior/posterior
1) could be parameter is weakly or non-identifiable
4. check whether parameters make any sense / Naive estimators
5. Do simulations
↳ the frequentist properties of your Bayesian Analysis

Simplify, Simplify, Simplify.

1. Go back to simple regression.

2. Fix parameters

$$\tau_{\text{As}} \leftarrow 1$$

instead of a distribution.

Arsenic - "As" is bad for you

		threshold
As(III)	Arsenite	.1
As(V)	Arsenate	.5

They'd like to measure $Y_{ij}^R \rightarrow$ true concentration of arsenic for subject i at visit j

R=1 As(III)

R=2 As(V)

C^R limit of detection

$W_{ij}^R \rightarrow$ observed concentration

60 subjects, 3 obs. / subject

Data seen:

$$w_{ij}^R = \begin{cases} y_{ij}^R, & y_{ij}^R > c^R \\ \text{NA}, & \text{otherwise} \end{cases}$$

censoring,
but not
independent

multilevel, lognormal, ~~probit~~ ... tobit model

$$y_{ij}^R = \mu_{ij}^k + b_i^k + \varepsilon_{ij}^k$$

$$\varepsilon_{ij}^k \sim N(0, \sigma_{\varepsilon, k}^2)$$

$$\mu_{ij}^k = x_{ij}^k \beta_R$$