

Overlapping Expert Information: Learning about Dependencies in Expert Judgment

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Extending Winkler's Multivariate Normal Aggregation

$P(\text{Event})$

$P(\text{Event} \mid X_1, \dots, X_n)$

$$P(\text{Event} \mid X_1, \dots, X_n)$$

Event = Collision between two vessels

X_1 = Type of other vessel

X_2 = Proximity of other vessel

X_3 = Wind speed

X_4 = Wind direction

X_5 = Current speed

X_6 = Current direction

X_7 = Visibility

$$P(\text{Event} \mid X_1, \dots, X_n)$$

Event = Incoming vessel contains RDD

X_1 = Last country docked

X_2 = 2nd to last country docked

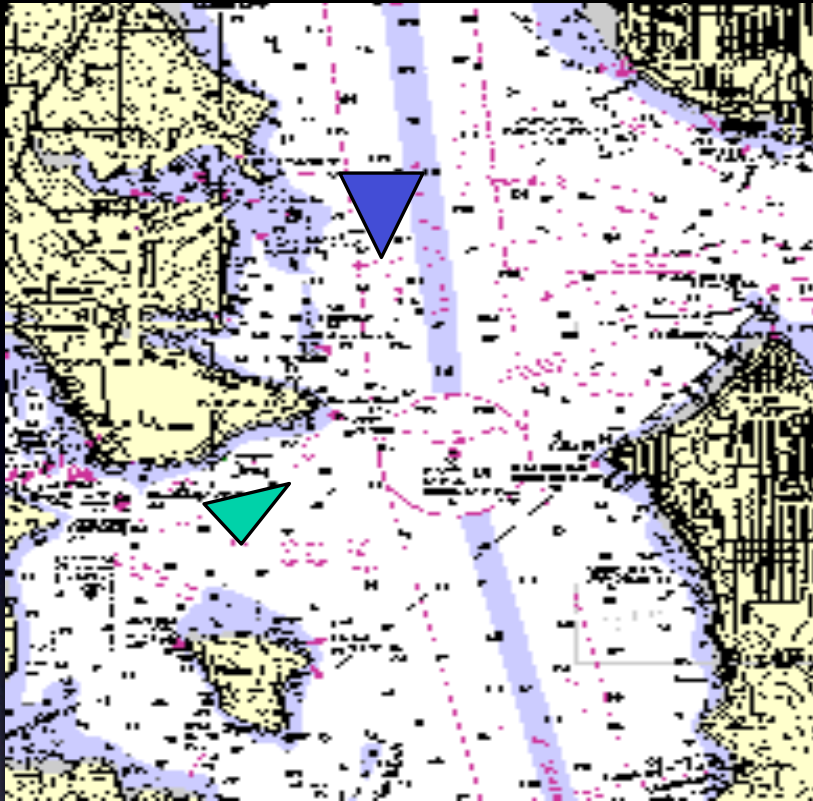
X_3 = 3rd to last country docked

X_4 = Frequency of US calls

X_5 = Vessel ownership

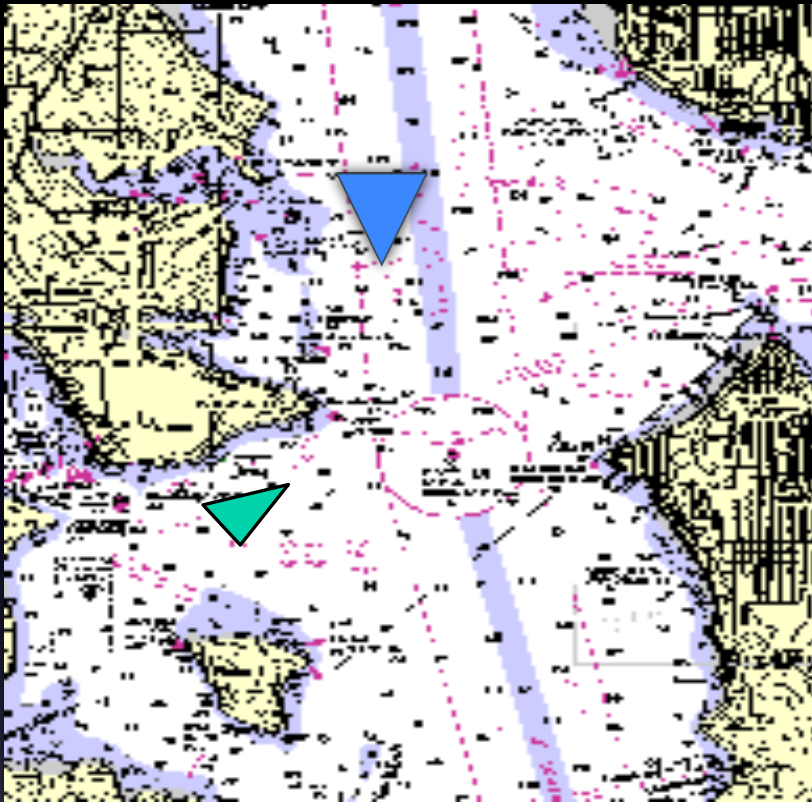
X_6 = Type of vessel

X_7 = Type of crew

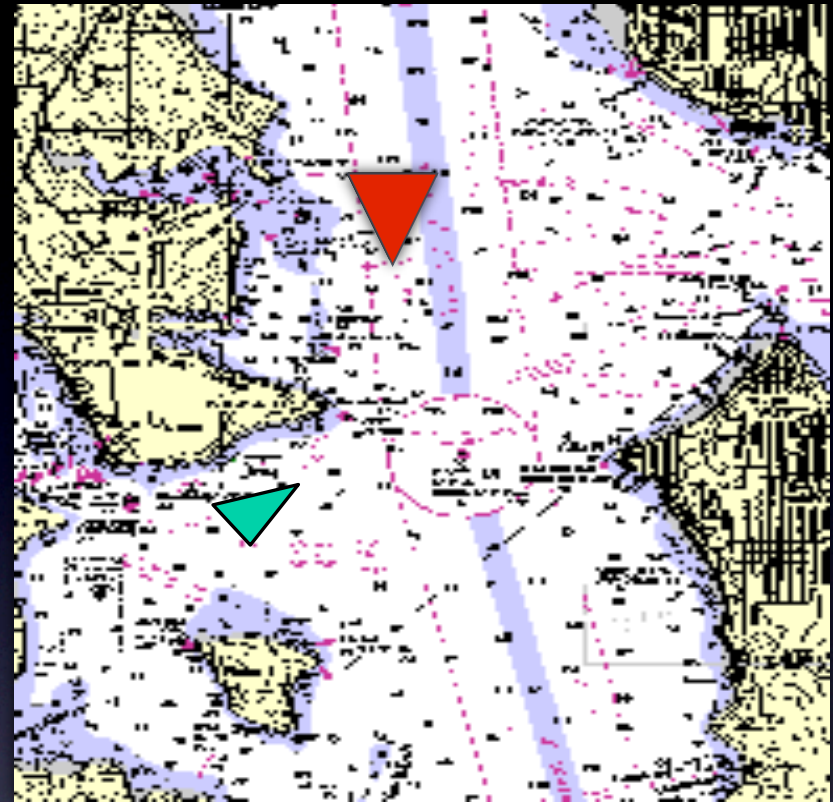


What is the probability
of a collision?

Issaquah class ferry
On the Bremerton to Seattle route
Crossing situation within 15 minutes
Other vessel is a navy vessel
No other vessels around
Good visibility
Negligible wind



Issaquah class ferry
On the Bremerton to Seattle route
Crossing situation within 15 minutes
Other vessel is a navy vessel
No other vessels around
Good visibility
Negligible wind



Issaquah class ferry
On the Bremerton to Seattle route
Crossing situation within 15 minutes
Other vessel is a product tanker
No other vessels around
Good visibility
Negligible wind

Issaquah	Ferry Class	-
SEA-BRE(A)	Ferry Route	-
Navy	1st Interacting Vessel	Product
Crossing	Traffic Scenario 1st Vessel	-
< 1 mile	Traffic Proximity 1st Vessel	-
No Vessel	2nd Interacting Vessel	-
No Vessel	Traffic Scenario 2nd Vessel	-
No Vessel	Traffic Proximity 2nd Vessel	-
> 0.5 Miles	Visibility	-
Along Ferry	Wind Direction	-
0	Wind Speed	-
	Likelihood of Collision	-

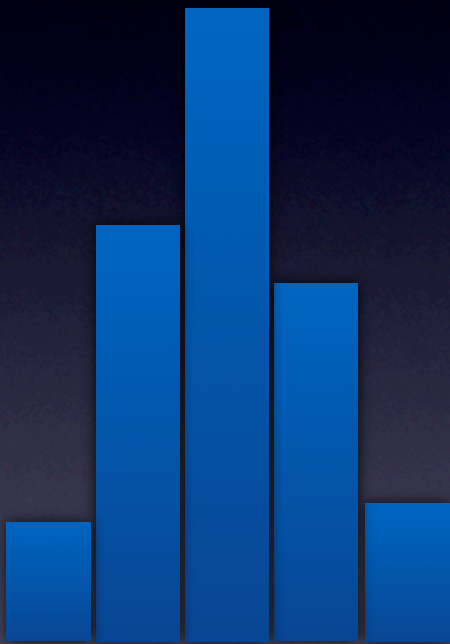
9 8 7 6 5 4 3 2 1 2 3 4 5 6 7 8 9

$$P(\text{Event} \mid X, p_0, \beta) = p_0 \exp(X^T \beta)$$

$$\frac{P(\text{Event} \mid R, \beta)}{P(\text{Event} \mid L, \beta)} = \frac{p_0 \exp(R^T \beta)}{p_0 \exp(L^T \beta)} = \exp((R - L)^T \beta)$$

$$y_{i,j} = \ln(z_{i,j}) = X_i^T \beta + u_{i,j}$$

$$y_{i,j} = \ln(z_{i,j}) = X_i^T \beta + u_{i,j}$$



9 8 7 6 5 4 3 2 1 2 3 4 5 6 7 8 9
Likelihood of Collision

$$\theta$$

$$u_i = \mu_i - \theta$$

$$\underline{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_p \end{pmatrix} \sim MVNormal(\underline{0}, \Sigma)$$

$$\pi(\theta; \mu, \Sigma) \propto \exp\left(-(\theta - \mu^*)^2 / 2\sigma^{*2}\right)$$

$$\mu^* = \frac{\underline{1}^T \Sigma^{-1} \mu}{\underline{1}^T \Sigma^{-1} \underline{1}}$$

$$\sigma^{*2} = \frac{1}{\underline{1}^T \Sigma^{-1} \underline{1}}$$

$$\theta$$

$$X_i^T \beta$$

$$u_i = \mu_i - \theta$$

$$u_{i,j} = y_{i,j} - X_i^T \beta$$

$$\underline{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_p \end{pmatrix} \sim MVNormal(\underline{0}, \Sigma)$$

$$\underline{u}_i = \begin{pmatrix} u_{i,1} \\ \vdots \\ u_{i,p} \end{pmatrix} \sim MVNormal(\underline{0}, \Sigma)$$

$$\begin{pmatrix} y_{1,1} & \cdots & y_{1,p} \\ \vdots & \ddots & \vdots \\ y_{N,1} & \cdots & y_{N,p} \end{pmatrix} = \begin{pmatrix} x_{1,1} & \cdots & x_{1,q} \\ \vdots & \ddots & \vdots \\ x_{N,1} & \cdots & x_{N,q} \end{pmatrix} \begin{pmatrix} \beta_1 & \cdots & \beta_1 \\ \vdots & \ddots & \vdots \\ \beta_q & \cdots & \beta_q \end{pmatrix} + \begin{pmatrix} u_{1,1} & \cdots & u_{1,p} \\ \vdots & \ddots & \vdots \\ u_{N,1} & \cdots & u_{N,p} \end{pmatrix}$$

$$\mathbf{Y} = \mathbf{X}\beta_{\underline{1}}^T + \mathbf{U}$$

$$p(\mathbf{Y} | \mathbf{X}, \beta, \Sigma) \propto |\Sigma|^{-\frac{N}{2}} \exp\left\{-\frac{1}{2} \text{tr}(\mathbf{V}\Sigma^{-1})\right\} \exp\left\{-\frac{1}{2} \text{tr}\left(\left(\hat{\mathbf{B}} - \beta \underline{\mathbf{1}}^T\right)^T \mathbf{X}^T \mathbf{X} \left(\hat{\mathbf{B}} - \beta \underline{\mathbf{1}}^T\right) \Sigma^{-1}\right)\right\}$$

$$\mu_*^\beta = \frac{\hat{\mathbf{B}} \Sigma^{-1} \underline{\mathbf{1}}}{\underline{\mathbf{1}}^T \Sigma^{-1} \underline{\mathbf{1}}}$$

$$\Sigma_*^\beta = \frac{\left(X^T X\right)^{-1}}{\underline{\mathbf{1}}^T \Sigma^{-1} \underline{\mathbf{1}}}$$

$$(\Sigma) \sim \text{Inv-Wishart}(\mathbf{G}, m)$$

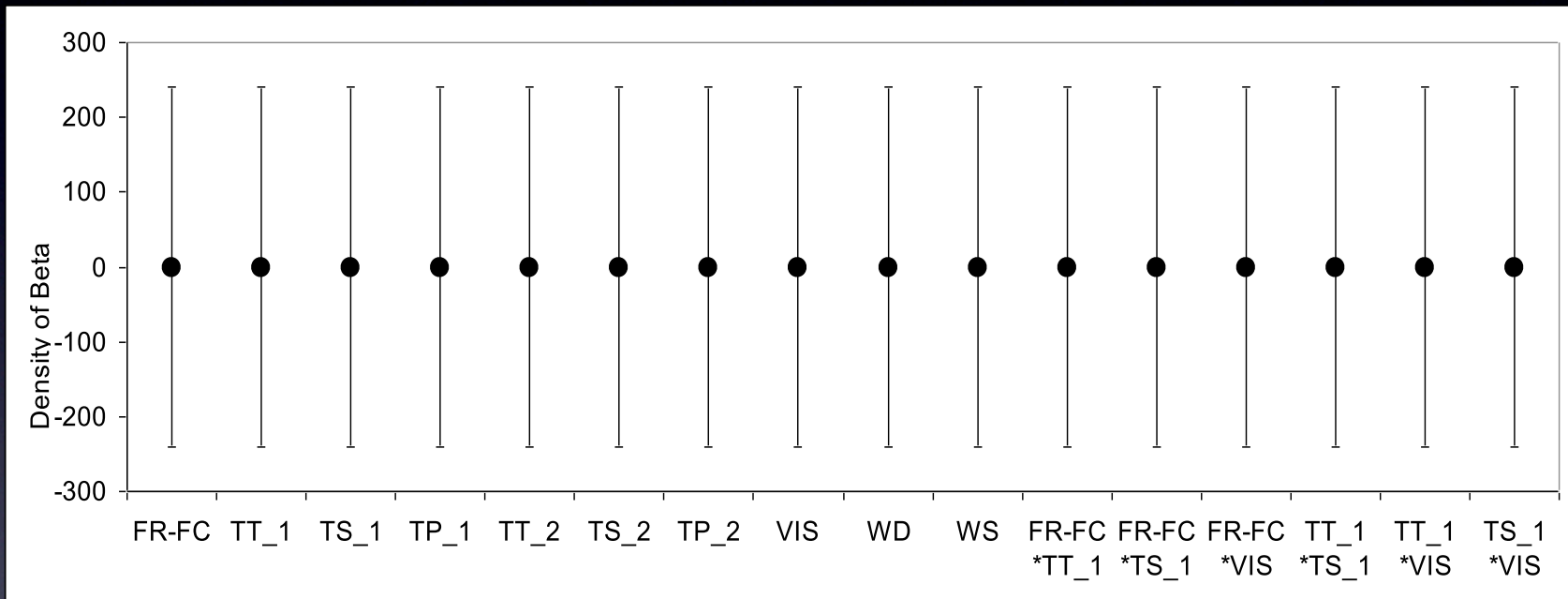
$$(\beta | \mathbf{Y}, \mathbf{X}, \Sigma) \sim \text{MVNormal}\left(\varphi, \frac{\mathbf{A}}{\underline{\mathbf{1}}^T \Sigma^{-1} \underline{\mathbf{1}}}\right)$$

$$\mathbf{V} = (\mathbf{Y} - \mathbf{X}\hat{\mathbf{B}})^T (\mathbf{Y} - \mathbf{X}\hat{\mathbf{B}})$$

$$(\Sigma | \mathbf{Y}, \mathbf{X}) \sim \text{Inv-Wishart}(\mathbf{G} + \mathbf{V}, m + N)$$

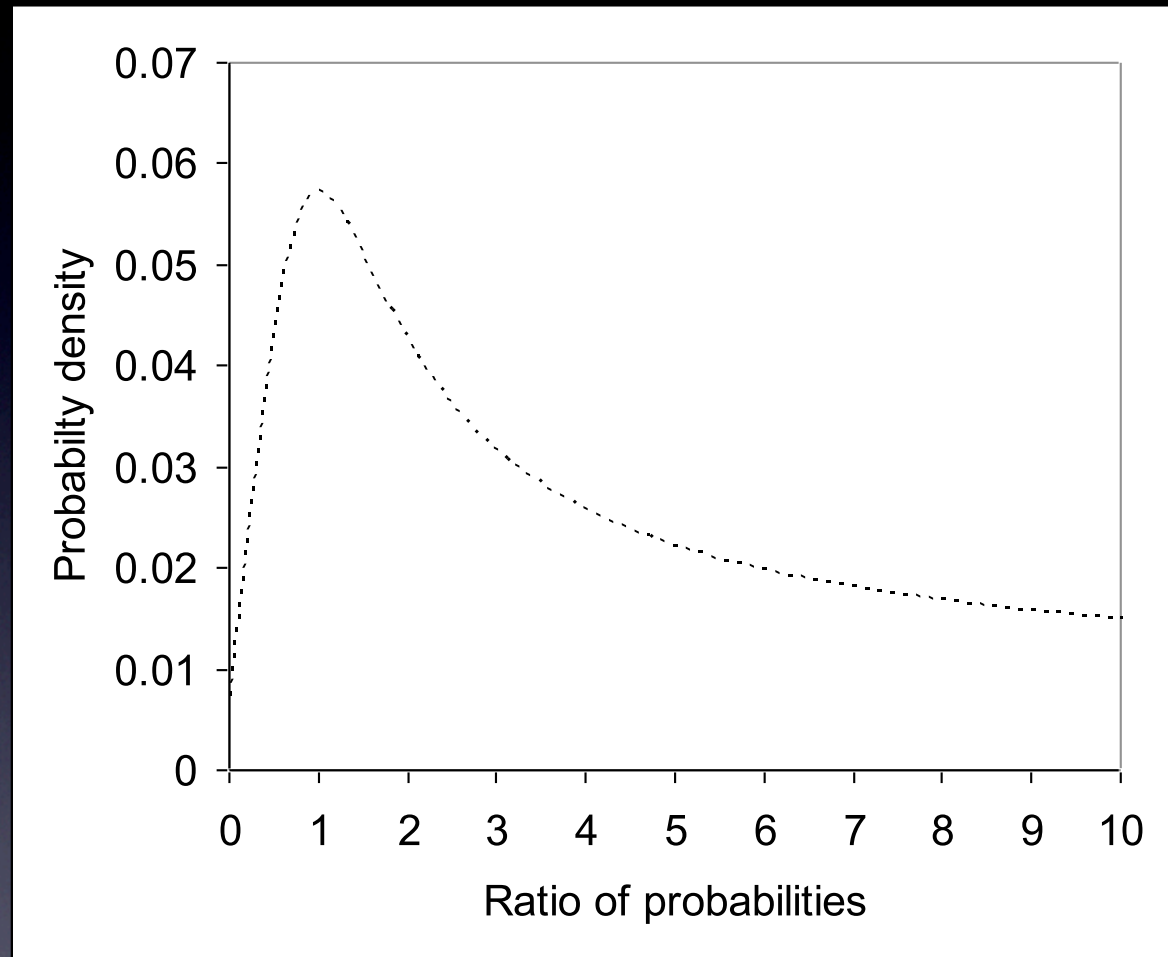
$$(\beta | \mathbf{Y}, \mathbf{X}, \Sigma) \sim \text{MVNormal}\left(\left(\mathbf{A}^{-1} + \mathbf{X}^T \mathbf{X}\right)^{-1} \left(\mathbf{X}^T \mathbf{X} \frac{\hat{\mathbf{B}} \Sigma^{-1} \underline{\mathbf{1}}}{\underline{\mathbf{1}}^T \Sigma^{-1} \underline{\mathbf{1}}} + \mathbf{A}^{-1} \varphi\right), \frac{\left(\mathbf{A}^{-1} + \mathbf{X}^T \mathbf{X}\right)^{-1}}{\underline{\mathbf{1}}^T \Sigma^{-1} \underline{\mathbf{1}}}\right)$$

Description	Notation	Values
Ferry route and class	FR_FC	26
Type of 1st interacting vessel	TT_1	13
Scenario of 1st interacting vessel	TS_1	4
Proximity of 1st interacting vessel	TP_1	Binary
Type of 2nd interacting vessel	TT_2	5
Scenario of 2nd interacting vessel	TS_2	4
Proximity of 2nd interacting vessel	TP_2	Binary
Visibility	VIS	Binary
Wind direction	WD	Binary
Wind speed	WS	Continuous

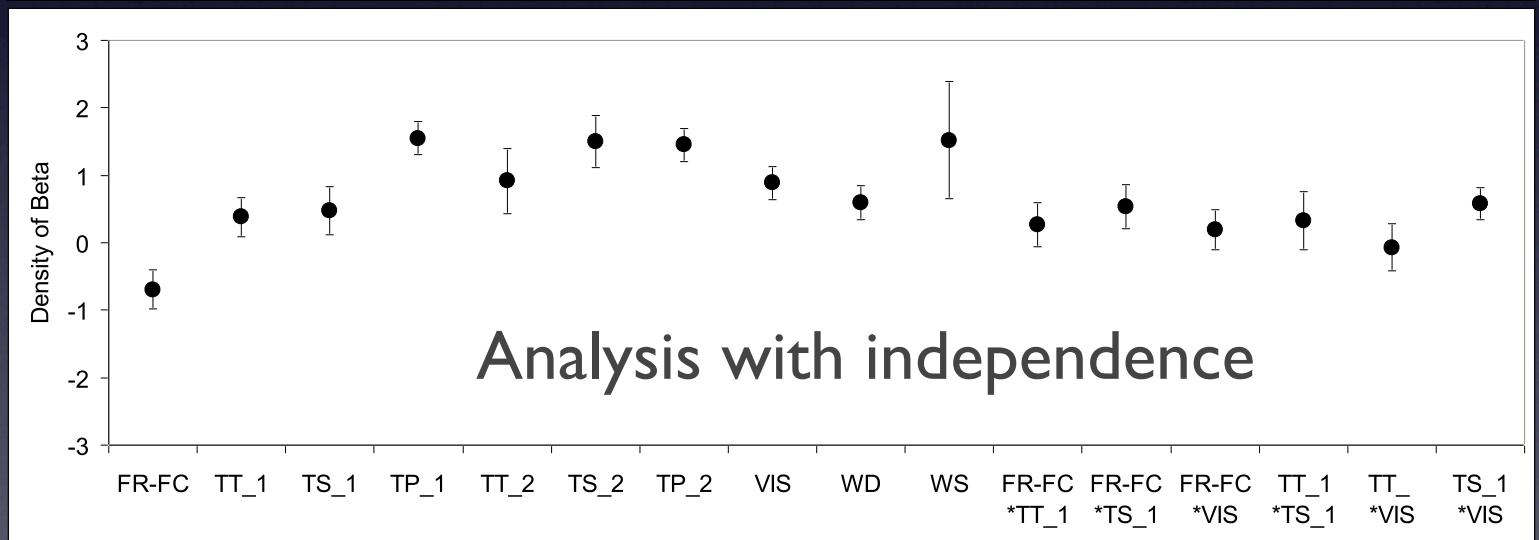
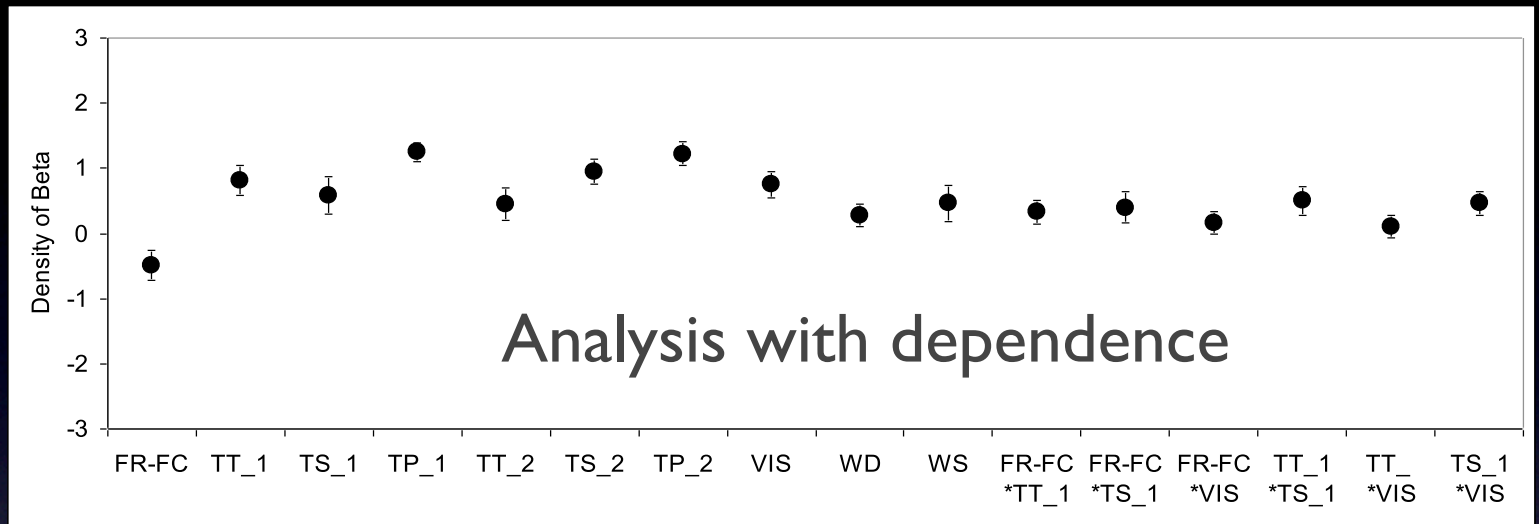


Assume independence between the
experts a priori

Comparing
the two
scenarios we
pictured
earlier a
priori

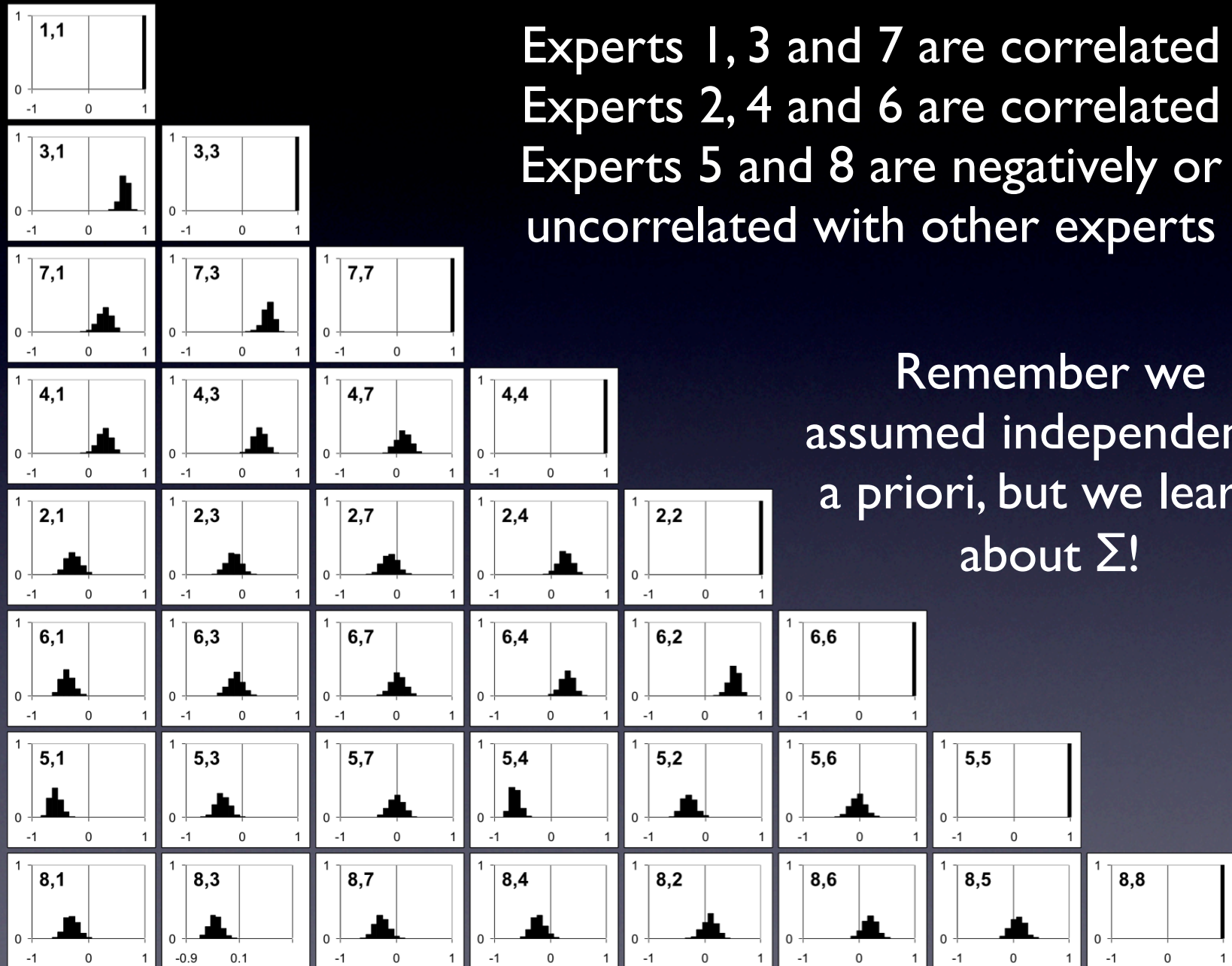


Doesn't
dependence
between
experts
increase
posterior
variance?

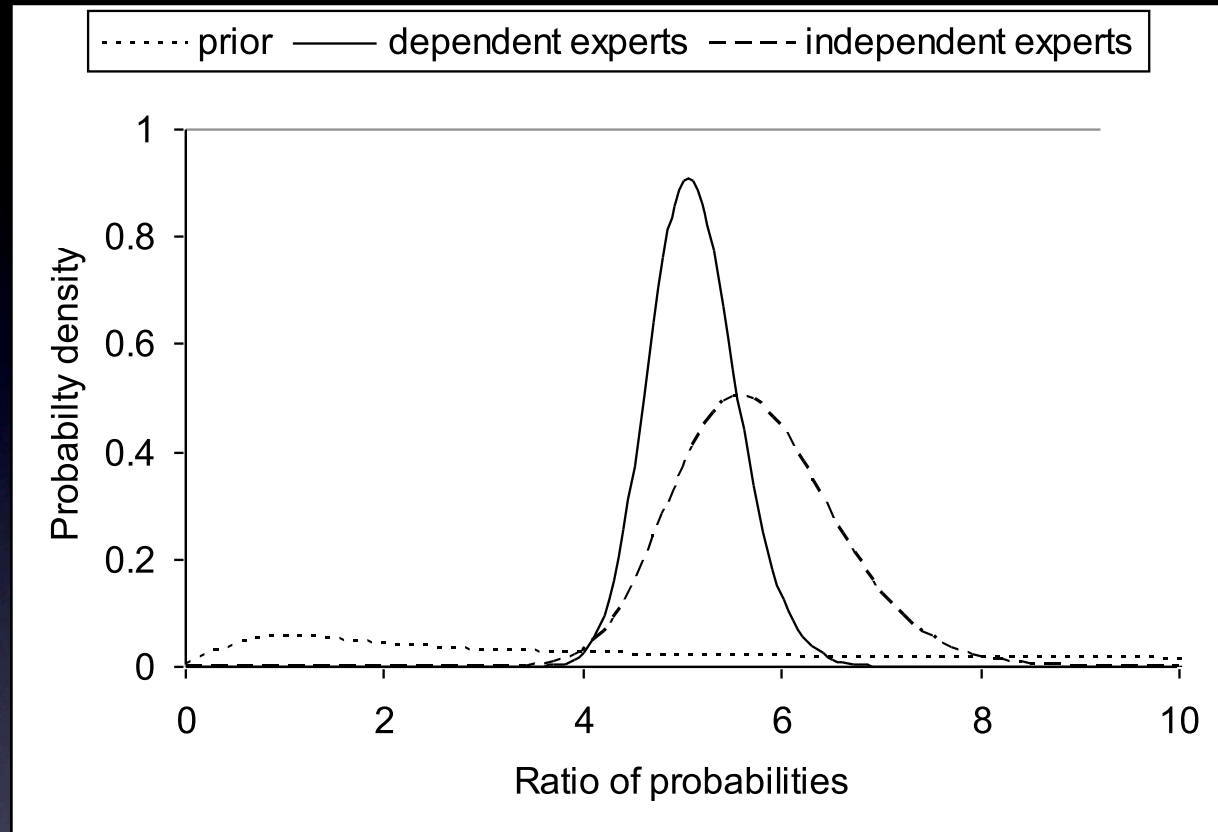


Experts 1, 3 and 7 are correlated
Experts 2, 4 and 6 are correlated
Experts 5 and 8 are negatively or
uncorrelated with other experts

Remember we
assumed independence
a priori, but we learnt
about Σ !



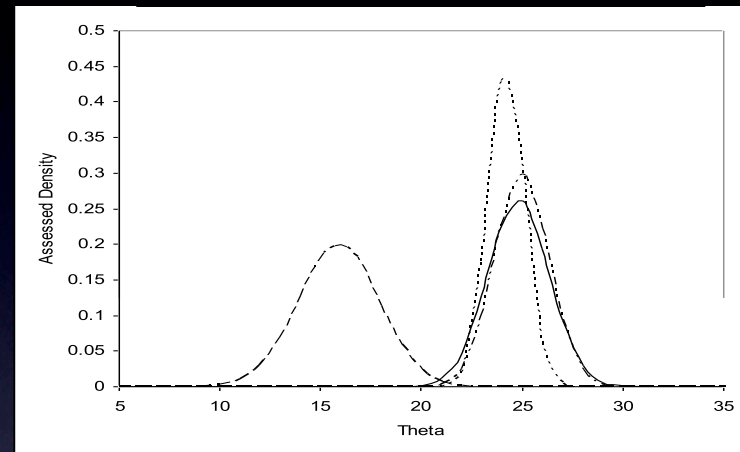
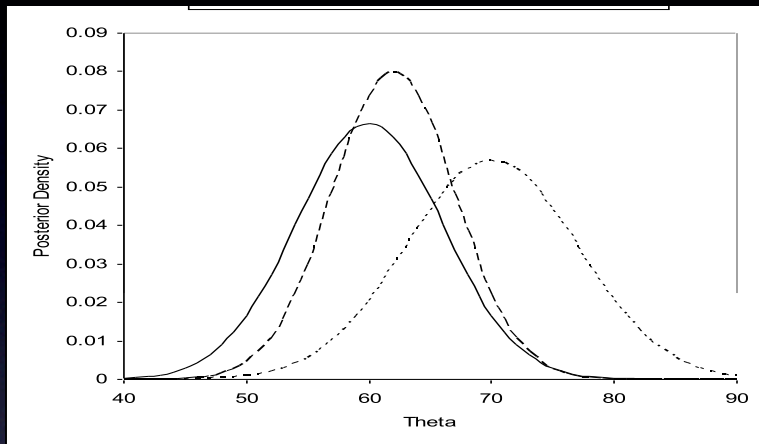
Comparing
the two
scenarios we
pictured
earlier



**90%
Credibility
Interval**

Prior	$[1.88 \cdot 10^{-35}, 5.32 \cdot 10^{34}]$
Dependent	$[4.38, 5.84]$ $\frac{1}{2}$ width = 0.73
Independent	$[4.43, 7.04]$ $\frac{1}{2}$ width = 1.3

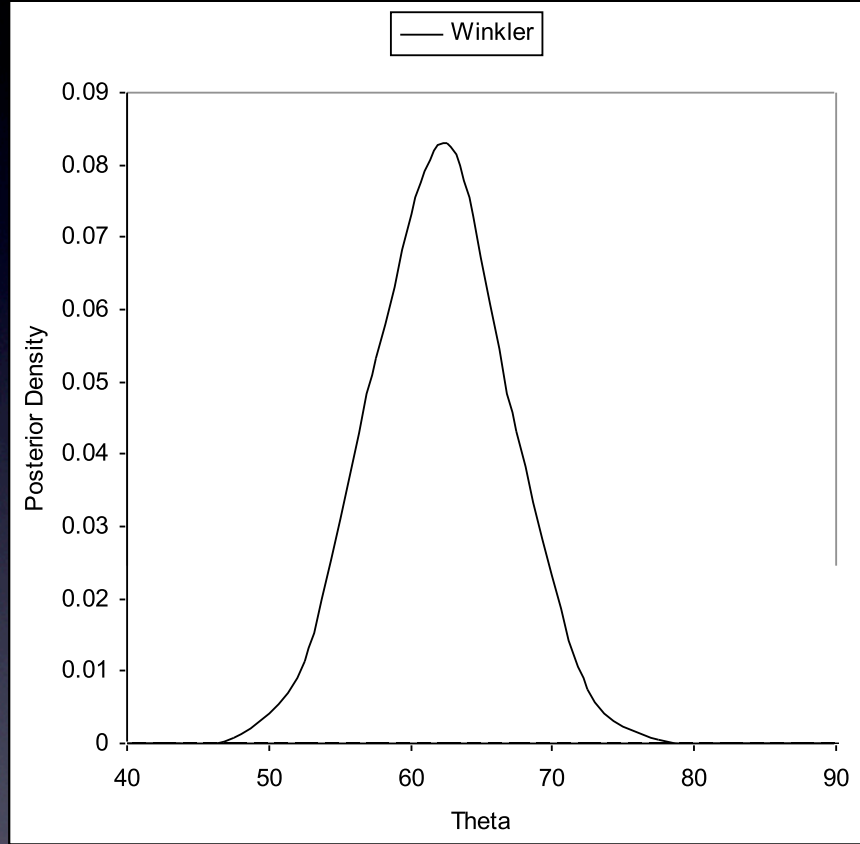
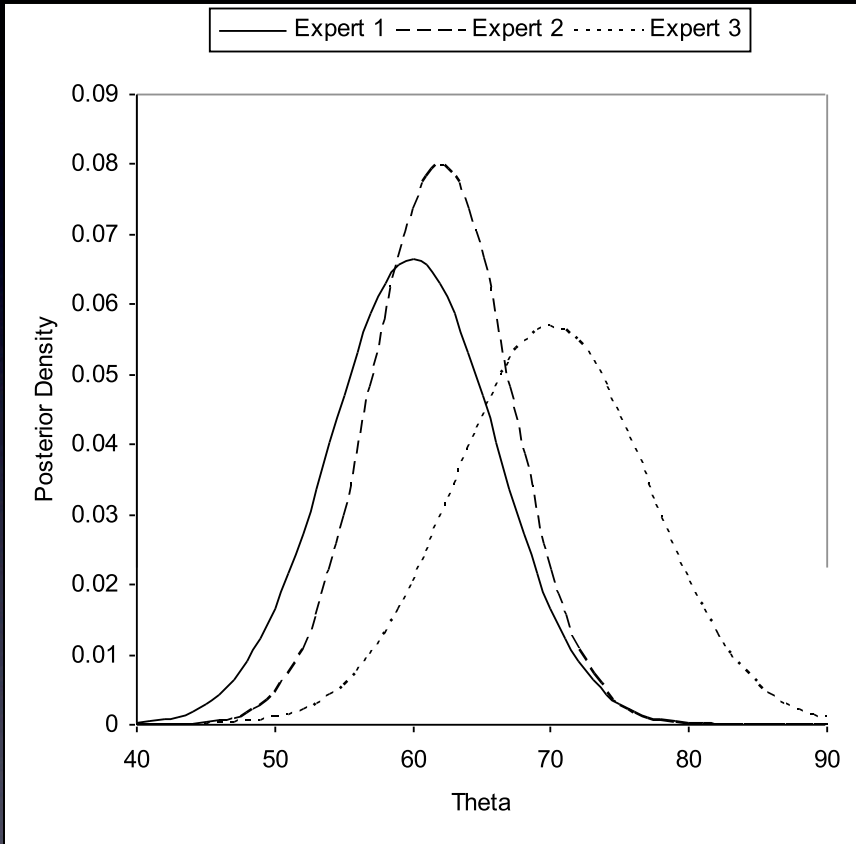
Getting the Right Mix of Experts



$$(z_1, \dots, z_p)$$

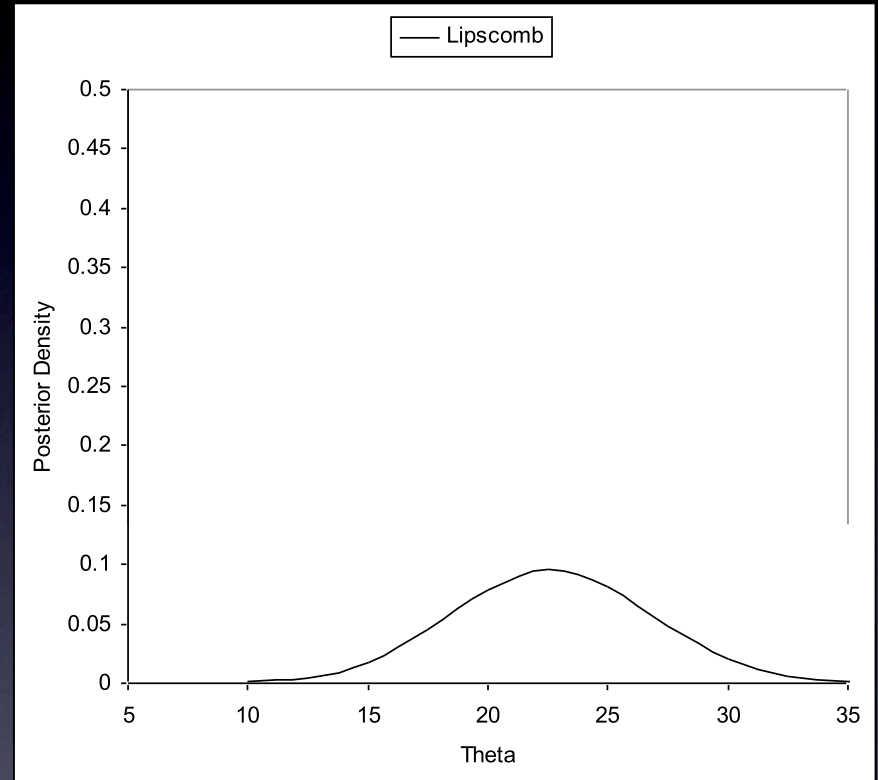
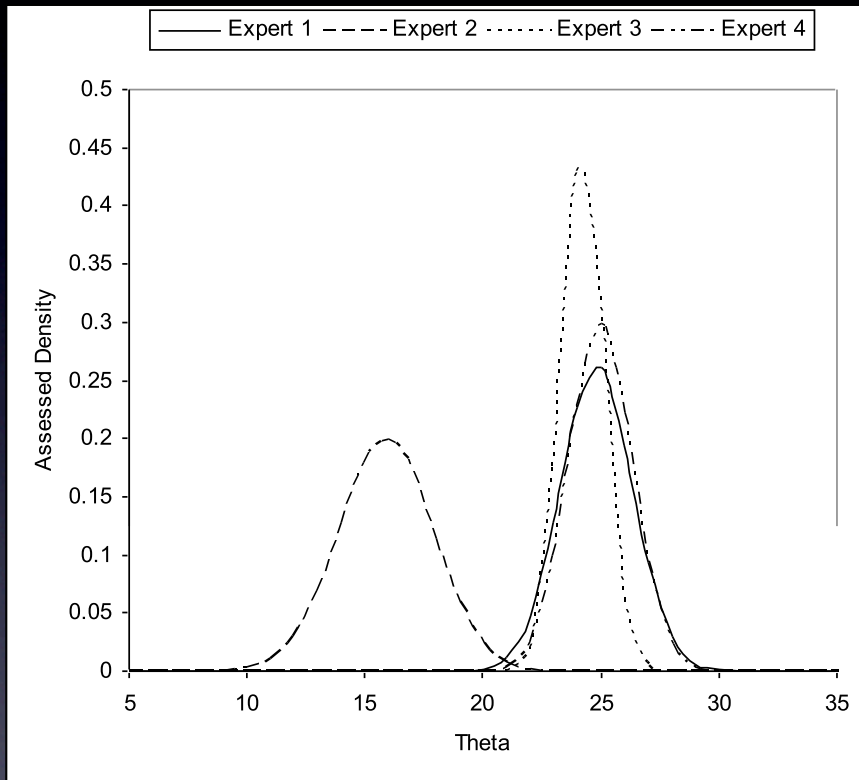
$$(r_1, \dots, r_p)$$

$$(z_1, \dots, z_n \mid \theta, \Sigma) \sim \text{MVNormal}(\underline{\theta}, \Sigma)$$



$$(z_i | \mu_i, \alpha_i, \gamma_i) \sim N(\mu_i, r_i)$$

$$(\mu_i | \theta, \lambda) \sim N(\theta, \lambda)$$



$$(z_i | \theta, r_i, \alpha_i, \gamma_i) \sim N(\theta + \alpha_i, \gamma_i r_i)$$

$$(r_i | \theta, z_i, a_\gamma, b_\gamma) \sim Ga(a_\gamma, b_\gamma)$$

$$\alpha_1, \dots, \alpha_p \sim N(0, \lambda)$$

$$\gamma_1, \dots, \gamma_p \sim Gamma(a, b)$$

$$(z_i | \theta, r_i, \alpha_i, \gamma_i) \sim N(\theta + \alpha_i, \gamma_i r_i)$$

$$(r_i | \theta, z_i, a_\gamma, b_\gamma) \sim Ga(a_\gamma, b_\gamma)$$

$$(\alpha_1, \gamma_1), \dots, (\alpha_p, \gamma_p) \sim G$$

$$G \sim DP(G_0, M)$$

$$G_0 = \text{gamma}(a, b)$$

