

# Bayesian model averaging

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# model uncertainty

$kid\_score \sim hs + iq + work + age$

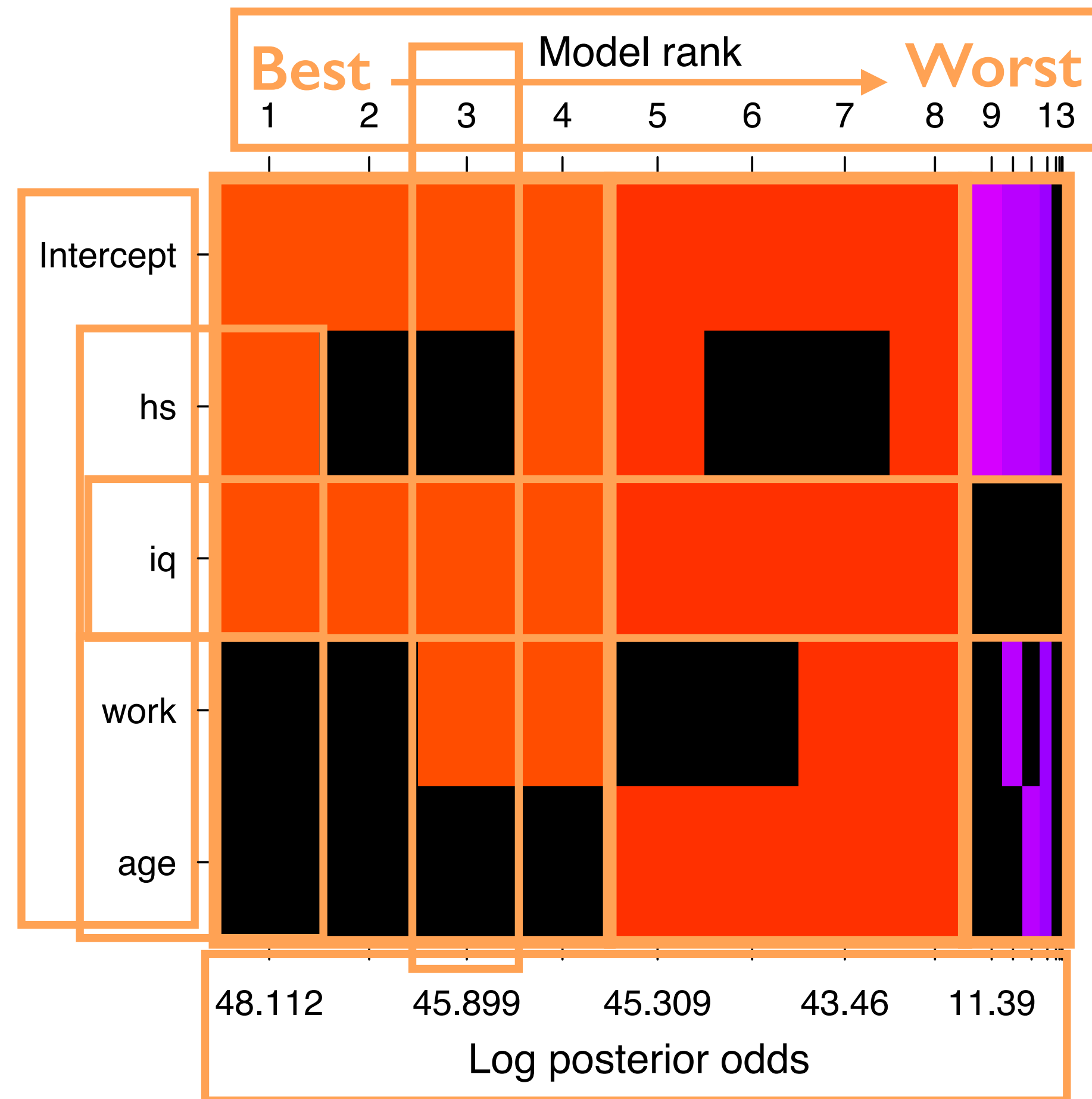
- ▶  $p$  predictors (4)
- ▶  $2^p$  possible models (16)



# visualizing model uncertainty

R

```
> image(cog_bas, rotate = F)
```



# Bayesian model averaging

- ▶ let  $\Delta$  be a quantity of interest

$Y^*$ ,  $\beta_j$ ,  $\gamma_j$  indicator variable  $j$  is included,  $p(\beta_j \mid \text{data})$

$$p(\Delta \mid \text{data}) = \sum_m^{2^p} p(\Delta \mid \mathcal{M}_m, \text{data}) p(\mathcal{M}_m \mid \text{data})$$

$$E[\Delta \mid \text{data}] = \sum_m^{2^p} E[\Delta \mid \mathcal{M}_m, \text{data}] p(\mathcal{M}_m \mid \text{data})$$

- ▶ weighted average of model specific quantities
- ▶ BMA predictions  $\hat{Y}^* = \sum \hat{Y}_m^* p(\mathcal{M}_m \mid \text{data})$



# coefficient summaries

R

```
> cog_coef
```

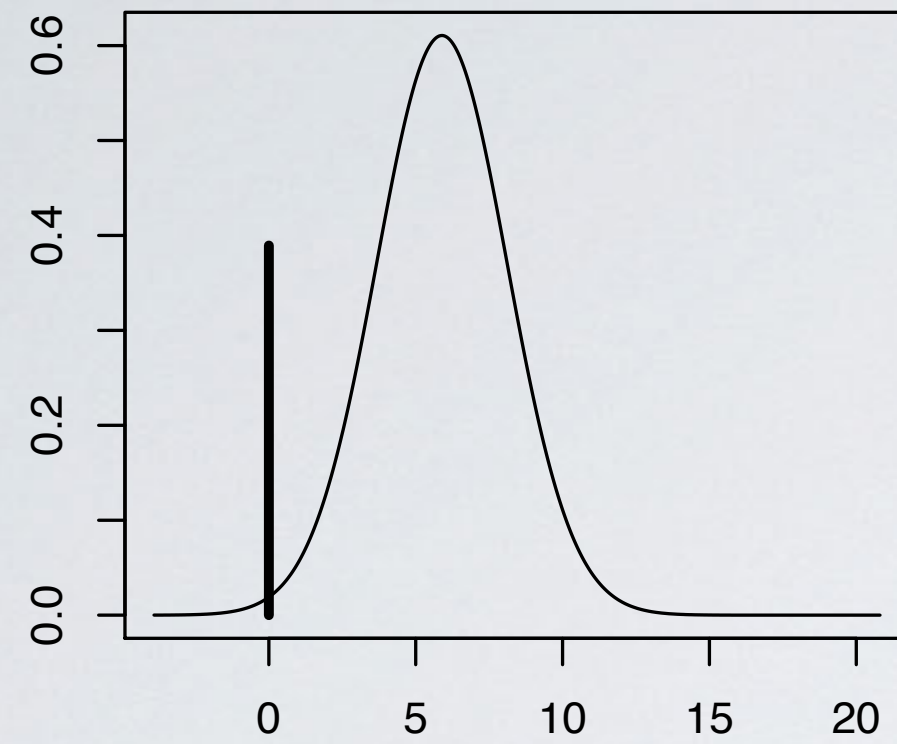
Marginal Posterior Summaries of Coefficients:

	post mean	post SD	post p(B != 0)
Intercept	86.79724	0.87287	1.00000
hs	3.59494	3.35643	0.61064
iq	0.58101	0.06363	1.00000
work	0.36696	1.30939	0.11210
age	0.02089	0.11738	0.06898

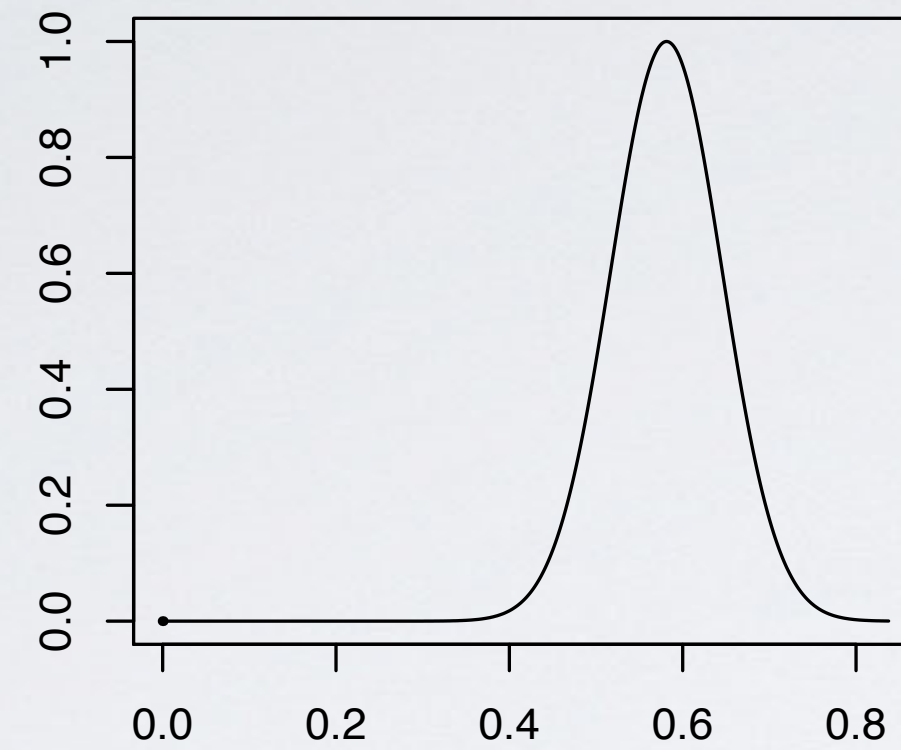
►  $P(\beta_{age} = 0 \mid \text{data}) = 0.931$

# plausible values of coefficients

**mom\_hsyas**

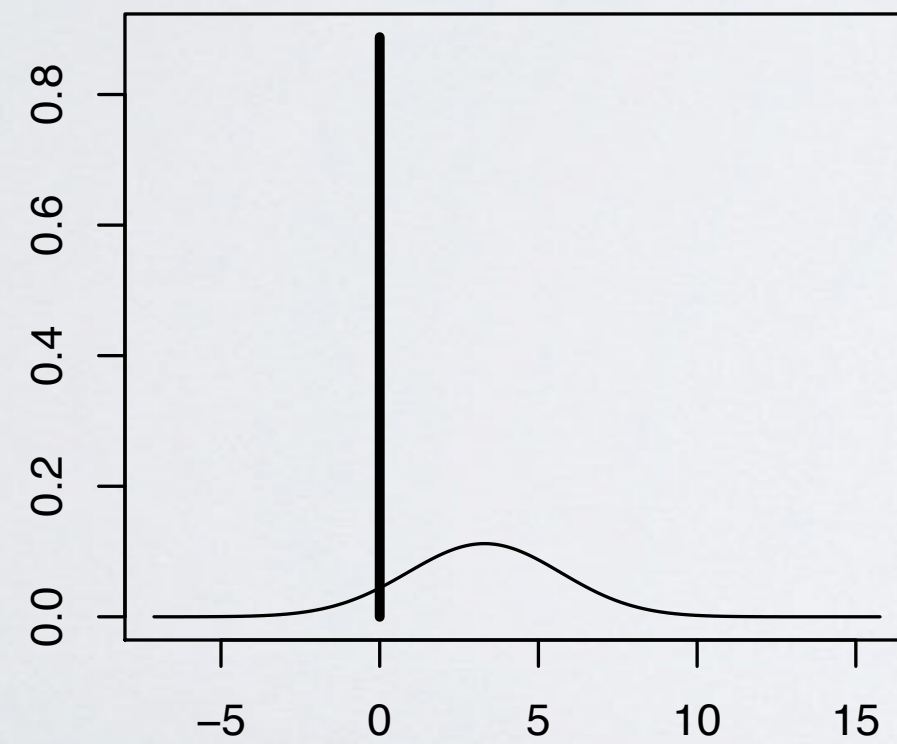


**mom\_iq**

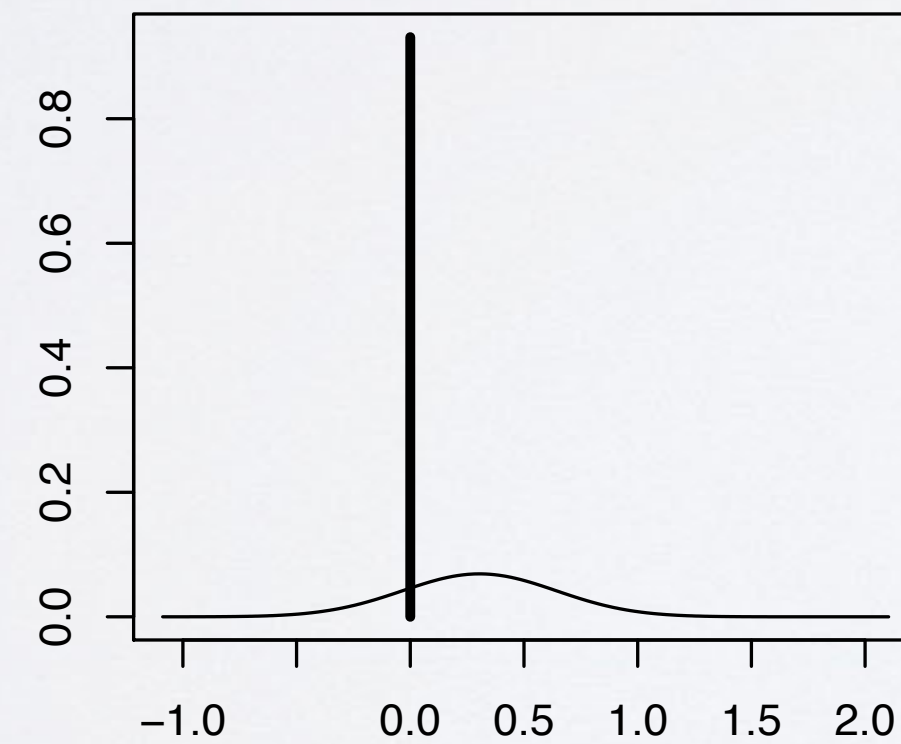


- ▶ spike represents probability coefficient is 0

**mom\_workyes**



**mom\_age**



- ▶ distribution of coefficient if it is not 0



# summary

- ▶ BMA accounts for model uncertainty
- ▶ software
- ▶ use BIC and reference prior

next:

- ▶ prior sensitivity