# GaussianJW

### Parametrisation

The GaussianJW likelihood is a two-part likelihood for  $\{(y,v)_i\}$ . First a Gaussian observation y

$$y|\ldots \sim \mathcal{N}(p, V(p, n))$$

with mean (probability) p and a variance function

$$V(p, n) = \exp(\beta_1 + \beta_2 \log(p(1-p)) + \beta_3 \log(n))$$

and then an (conditional independent) observed variance v, where

$$\nu \frac{v}{V(p,n)} | \dots \sim \chi_{\nu}^2.$$

The case  $\beta_1 = 0, \beta_2 = 1, \beta_3 = -1$  resembles the case where a Binomial is approximated with a Normal.  $(n, \nu)$  is considered as fixed.

### **Link-function**

The probability p is linked to the linear predictor  $\eta$  with a (default) logit link

$$p = \frac{1}{1 + \exp(-\eta)}$$

but other links are also possible.

### Hyperparameters

The hyperparameters are

$$\theta_1 = \beta_1$$

$$\theta_2 = \beta_2$$

$$\theta_3 = \beta_3$$

and the prior is defined on  $(\theta_1, \theta_2, \theta_3)$ .

### **Specification**

- family="gaussianjw"
- Required arguments: y, n and  $\nu$  (all vectors of the same length) as an inla.mdata()-object with this spesific ordering, see the example.

### Hyperparameter spesification and default values

doc The GaussianJW likelihoood

hyper

## theta1

hyperid 65101 name beta1 short.name beta1 initial 0

```
fixed FALSE
         prior normal
         param 0 100
         to.theta function(x) x
         from.theta function(x) x
     theta2
         hyperid 65102
         name beta2
         short.name beta2
         initial 1
         fixed FALSE
         prior normal
         param 1 100
         to.theta function(x) x
         from.theta function(x) x
     theta3
         hyperid 65103
         name beta3
         short.name beta3
         initial -1
         fixed FALSE
         prior normal
         param -1 100
         to.theta function(x) x
         from.theta function(x) x
status experimental
survival FALSE
discrete FALSE
link default logit probit
pdf gaussianjw
Example
n <- 300
x \leftarrow rnorm(n, sd = 0.5)
eta <- 1 + x
p <- 1/(1 + exp(-eta))
df <- sample(10:100, n, replace = TRUE)</pre>
size <- df
va \leftarrow p * (1.0 - p) / size
v \leftarrow rchisq(n, df = df) * va / df
phat <- rnorm(n, mean = p, sd = sqrt(v))</pre>
Y <- inla.mdata(phat, v, size, df)
```