Copy of another model component: "copy"

This describes the way to copy another model component with an optional scaling. This issue arise frequently, like with a model like

$$\eta_{ij} = u_i + u_j$$

where u is the **same model component**. This is not doable with normal use of 'formula', hence we introduce an identical (up to numerical error)

$$v = \mathsf{copy}(u)$$

so we can write

$$\eta = u + v$$

We can also enable an optional scaling, so that

$$v = \beta \times \text{copy}(u)$$

where β is estimated as well.

Note the following.

- One model component can be copied many times.
- v will inherit the main properties from u, such as replicate, nrep, group, ngroup and values. So it is natural to use same kind of indexing for v as u. For example

y
$$\sim$$
 f(i, replicate=r, group=g) + f(j, copy="i", replicate=rr, group=gg)

Here, the argument copy, say that this model component is a copy of "i".

• If the same indexing is **not** used, then the equivalent one-dimensional indexing is used as defined in inla.idx().

Hyperparameters

The optional hyperparameter is β which is **default fixed to** $\beta = 1$.

We can estimate β by setting fixed=FALSE, like

We can control β , in two ways.

- 1. We can fix β to be in an interval defined by argument range=c(r1,r2), where $r_1 \leq r_2$
 - If $r_1 = r_2$, then there is no restrictions on β .
 - If $-\infty < r_1 < r_2 < \infty$, then β is defined to be in the interval (r_1, r_2) , and the prior, initial values are defined on $\tilde{\beta}$, where

$$\beta = r_1 + (r_2 - r_1) \frac{1}{1 + \exp(-\tilde{\beta})}$$

• If $-\infty < r_1 < r_2 = \infty$, then β is defined to be in the interval (r_1, ∞) and the prior, initial values are defined on $\tilde{\beta}$, where

$$\beta = r_1 + \exp(\tilde{\beta})$$

2. We can make β the same as a β in another copy, with argument same.as, so that

```
\eta_{ijk} = u_i + \beta u_j + \beta u_k
```

with

```
y \sim f(i) + f(j, copy="i", hyper=list(beta=list(fixed=FALSE))) + f(k, copy="i", same.as="j")
```

3. How cloes a the copy is, is determined by the argument precision=..., and the default value is

```
f(j, copy="i", precision=exp(14))
```

meaning that $v = u + \epsilon$, where ϵ is iid zero mean Normal with precision exp(14).

Spesification

doc Create a copy of a model component

hyper

```
theta
    hyperid 36001
    name beta
    short.name b
    initial 0
    fixed TRUE
    prior normal
    param 1 10
    to.theta function(x, REPLACE.ME.low, REPLACE.ME.high) {
                                                                                          } e]
                                              return(x)
                                              stopifnot(low < high)</pre>
                                          } else if (is.finite(low) && is.infinite(high) &&
                                              return(log(x - low))
                                              stop("Condition not yet implemented")
    from.theta function(x, REPLACE.ME.low, REPLACE.ME.high) {
                                                                                          } e]
                                              return(x)
                                              stopifnot(low < high)</pre>
```

} else if (is.finite(low) && is.infinite(high) &&

}

stop("Condition not yet implemented")

return(low + exp(x))

constr FALSE

nrow.ncol FALSE

augmented FALSE

aug.factor 1

}

```
aug.constr
```

n.div.by

n.required FALSE

set.default.values FALSE

pdf copy

Example

Just simulate some data and estimate the parameters back.

```
## simple example to illustrate the use of 'copy'
set.seed(1234)
N <- 100
n <- 50
u <- scale(rnorm(n))
s <- 0.1
i <- sample(1:n, N, replace = TRUE)</pre>
j <- sample(1:n, N, replace = TRUE)</pre>
k <- sample(1:n, N, replace = TRUE)</pre>
y \leftarrow u[i] + u[j] + u[k] + rnorm(N, sd = s)
r \leftarrow inla(y \sim -1 +
               f(i, values = 1:n) +
               f(j, copy = "i") +
              f(k, copy = "i"),
          data = data.frame(y, i, j, k),
          control.family = list(hyper = list(
                                  prec = list(initial = log(1/s^2),
                                              fixed = TRUE))))
plot(u, r$summary.random$i$mean, pch = 19)
abline(a = 0, b = 1, lwd = 3, col = "blue")
## estimate scaling parameters, assuming
## y <- u[i] + beta.j * u[j] + beta.k * u[k] + rnorm(N, sd = s)
## where the true values are beta.j=1 and beta.k=1
rr \leftarrow inla(y \sim -1 +
               f(i, values = 1:n) +
               f(j, copy = "i", hyper = list(
                                    beta = list(fixed = FALSE))) +
              f(k, copy = "i", hyper = list(
                                     beta = list(fixed = FALSE))),
          data = data.frame(y, i, j, k),
          control.family = list(hyper = list(
                                  prec = list(initial = log(1/s^2),
                                              fixed = TRUE))))
rr$summary.hyperpar[,c("mean","sd")]
inla.dev.new()
plot(u, rr$summary.random$i$mean, pch = 19)
abline(a = 0, b = 1, lwd = 3, col = "blue")
## now we assume that we know that beta.k = beta.j, and we estimate
## just beta.j
```

Notes