linearFeedbackModel

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lfm	Estimate the linear feedback model in Blundell, Griffith and Windmeijer (2002)	

Description

Estimate the linear feedback model in Blundell, Griffith and Windmeijer "Individual effects and dynamics in count data models", Journal of Econometrics 108 (2002) 113-131

Usage

```
lfm(formula, data, effect = "individual", model = "onestep")
```

Arguments

formula	Similar to the pgmm() function in package plm. A symbolic description for the model to be estimated. Indicate a multi-part formula, the first two parts describing the covariates and the gmm instruments and, if any, the third part the 'normal' instruments. The first independent variable must be the lag of the dependent variable.
data	A pdata.frame.
effect	Either "individual" or "twoways". The former only includes individual fixed effects while the latter also includes time fixed effects.
effect	Either "onestep" or "twosteps". Whether to do one-step GMM or two-step GMM .

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Value

The matched call call coefficients The estimated coefficient The average of the Jacobian of the sample moment conditions over each indifitted.values data.frame of fitted values first The first stage estimates Estimates of the individual fixed effects fixed.effects mode1 The variables used for estimation for each individual residuals data.frame of residuals vcov The covariance matrix of the coefficients W1 The first-stage weight matrix used The second-stage (efficient) weight matrix used (only returned if model = "twosteps" W2 The instrument matrix for each individual Ζ

Examples

```
library(linearFeedbackModel)
library(plm)
set.seed(123)
# Create some data - follow example in Blundell, Griffith and Windmeijer (2002):
nΤ
           <- 8
                     # Number of time periods
           <- 500 # Number of individuals
           <- 0.5 # Coefficient on lagged dependent variable
gamma
beta
           <- 0.5
                    # Coefficient on x
rho
           <- 0.5
                    # For the starting values of x
           <- 0.1
                    # For effect of individual effects on x
tau
                    # Variance of distribution of individual fixed effects
           <- 0.5
var_epsilon <- 0.5 # Variance of distribution of x
# Parameters to be estimated:
theta <- c(gamma, beta)
# Draw individual effects:
eta <- rnorm(N, mean = 0, sd = sqrt(var_eta))
# Draws for starting values of x:
xi <- rnorm(N, mean = 0, sd = sqrt(var_epsilon / (1 - rho^2)))
# Draw initial conditions for x and y:
x0 <- (tau / (1 - rho)) * eta + xi
y0 <- sapply(1:N, function(i) rpois(1, lambda = exp(x0 * beta + eta[i])))
# Dataset to be created: List for each time period
```

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```
df <- vector("list", length = nT)</pre>
# Create first time period:
x \leftarrow rho * x0 + tau * eta + rnorm(N, mean = 0, sd = sqrt(var_epsilon))
y <- sapply(1:N, function(i)</pre>
            rpois(1, lambda = gamma * y0 + exp(beta * x0[i] + eta[i])))
df[[1]] \leftarrow data.frame(i = 1:N, t = 1, x = x, y = y)
# Create all following time periods:
for (t in 2:nT) {
  x <- rho * df[[t - 1]]$x + tau * eta +
       rnorm(N, mean = 0, sd = sqrt(var_epsilon))
  y \leftarrow sapply(1:N, function(i) rpois(1, lambda = gamma * df[[t - 1]]$y[i] +
                                       exp(beta * df[[t - 1]]x[i] + eta[i]))
 df[[t]] \leftarrow data.frame(i = 1:N, t = t, x = x, y = y)
}
# Combine into one dataset:
df <- do.call(rbind, df)</pre>
# Convert to pdata.frame:
data <- pdata.frame(df, index = c("i", "t"))</pre>
test <- lfm(y \sim lag(y, k = 1) + x \mid lag(y, k = 2:4) + lag(x, k = 1:4),
             data = data, effect = "individual", model = "onestep")
print(summary(test))
```

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