# Package: PDMIF (via r-universe)

September 12, 2024

```
Title Fits Heterogeneous Panel Data Models
```

Version 0.1.0

```
Description Fits heterogeneous panel data models with interactive
      effects for linear regression, logistic, count, probit,
      quantile, and clustering. Based on Ando, T. and Bai, J. (2015)
      ``A simple new test for slope homogeneity in panel data models
      with interactive effects" <doi:10.1016/j.econlet.2015.09.019>,
      Ando, T. and Bai, J. (2015) ``Asset Pricing with a General
      Multifactor Structure" <doi:10.1093/jjfinex/nbu026>, Ando, T.
      and Bai, J. (2016) ` Panel data models with grouped factor
      structure under unknown group member-
      ship" <doi:10.1002/jae.2467>, Ando, T. and Bai, J. (2017) ``Clustering huge
      number of financial time series: A panel data approach with
      high-dimensional predictors and factor struc-
      tures" <doi:10.1080/01621459.2016.1195743>, Ando, T. and Bai, J. (2020)
      ``Quantile co-movement in financial mar-
      kets" <doi:10.1080/01621459.2018.1543598>, Ando, T., Bai, J. and Li, K.
      (2021) `Bayesian and maximum likelihood analysis of large-scale
      panel choice models with unobserved heterogeneity" <doi:10.1016/j.jeconom.2020.11.013.>.
```

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**Encoding UTF-8** 

LazyData true

LazyDataCompression xz

**Roxygen** list(markdown = TRUE)

RoxygenNote 7.1.1

Imports diagonals, nevreg, quantreg

**Depends** R (>= 2.10)

Repository https://tomohiro-ando.r-universe.dev

RemoteUrl https://github.com/tomohiro-ando/pdmif

RemoteRef HEAD

**RemoteSha** 3a275ee7e25d319d2b744a6221b1ed079f174e76

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## Description

A synthesized input variable dataset to fit a linear model on a panel dataset.

# Usage

data1X

data1Y 3

### **Format**

A data frame with 5,000 rows and 2 columns:

**columns** the two independent variables

rows each 100 rows represent the timeseries of each of the 50 individuals ...

data1Y

A synthesized output variable dataset to fit a linear model on a panel dataset.

### **Description**

A synthesized output variable dataset to fit a linear model on a panel dataset.

#### Usage

data1Y

#### **Format**

A data frame with 100 rows and 50 columns:

columns the individuals

rows the time points in the timeseries of each individual ...

data2X

A synthesized input variable dataset to fit a binomial model on a panel dataset.

### Description

A synthesized input variable dataset to fit a binomial model on a panel dataset.

#### Usage

data2X

#### **Format**

A data frame with 5,000 rows and 2 columns:

columns the two independent variables

rows each 50 rows represent the timeseries of each of the 100 individuals ...

4 data 3X

data2Y	A synthesized output variable dataset to fit a binomial model on a panel dataset.

### Description

A synthesized output variable dataset to fit a binomial model on a panel dataset.

### Usage

data2Y

#### **Format**

A data frame with 50 rows and 100 columns:

columns the individuals

rows the time points in the timeseries of each individual ...

data3X	A synthesized input variable dataset to fit a poisson model on a panel
	dataset.

### Description

A synthesized input variable dataset to fit a poisson model on a panel dataset.

### Usage

data3X

#### **Format**

A data frame with 5,000 rows and 3 columns:

columns the three independent variables

rows each 50 rows represent the timeseries of each of the 100 individuals ...

data3Y 5

data3Y	A synthesized output variable dataset to fit a poisson model on a panel dataset.

### Description

A synthesized output variable dataset to fit a poisson model on a panel dataset.

### Usage

data3Y

### **Format**

A data frame with 50 rows and 100 columns:

columns the individuals

rows the time points in the timeseries of each individual ...

	A synthesized vector of memberships needed to fit a linear model on a panel dataset under known group memberships.
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### Description

A synthesized vector of memberships needed to fit a linear model on a panel dataset under known group memberships.

### Usage

data4LAB

#### **Format**

A vector with 300 entries indicating the group membership of each individual.

6 data4Y

data4X	A synthesized input variable dataset to fit a linear model on a panel
	dataset under known group memberships.

### Description

A synthesized input variable dataset to fit a linear model on a panel dataset under known group memberships.

### Usage

data4X

#### **Format**

A data frame with 30,000 rows and 2 columns:

columns the two independent variables

rows each 100 rows represent the timeseries of each of the 300 individuals ...

data4Y	A synthesized output variable dataset to fit a linear model on a panel dataset under known group memberships.
	See of the

### Description

A synthesized output variable dataset to fit a linear model on a panel dataset under known group memberships.

### Usage

data4Y

#### **Format**

A data frame with 100 rows and 300 columns:

columns the individuals

data5X 7

data5X	A synthesized input variable dataset to cluster individuals by hetero-
	geneous panel data models with interactive effects.

### Description

A synthesized input variable dataset to cluster individuals by heterogeneous panel data models with interactive effects.

### Usage

data5X

#### **Format**

A data frame with 30,000 rows and 2 columns:

columns the two independent variables

rows each 100 rows represent the timeseries of each of the 300 individuals ...

data5Y	A synthesized output variable dataset to cluster individuals by hetero-
	geneous panel data models with interactive effects.

### Description

A synthesized output variable dataset to cluster individuals by heterogeneous panel data models with interactive effects.

### Usage

data5Y

#### **Format**

A data frame with 100 rows and 300 columns:

columns the individuals

8 data6Y

data6X	A synthesized input variable dataset to cluster individual units by non-
	linear heterogeneous panel data models with interactive effects when
	the group membership is unknown

### Description

A synthesized input variable dataset to cluster individual units by nonlinear heterogeneous panel data models with interactive effects when the group membership is unknown

### Usage

data6X

#### **Format**

A data frame with 4,500 rows and 2 columns:

columns the two independent variables

rows each 50 rows represent the timeseries of each of the 90 individuals ...

A synthesized output variable dataset to cluster individual units by
nonlinear heterogeneous panel data models with interactive effects
when the group membership is unknown.

### Description

A synthesized output variable dataset to cluster individual units by nonlinear heterogeneous panel data models with interactive effects when the group membership is unknown.

### Usage

data6Y

#### **Format**

A data frame with 50 rows and 90 columns:

columns the individuals

data7X

data7X	A synthesized input variable dataset to fit a quantile panel data model on a panel dataset.

### Description

A synthesized input variable dataset to fit a quantile panel data model on a panel dataset.

### Usage

data7X

### **Format**

A data frame with 20,000 rows and 3 columns:

columns the three independent variables

rows each 100 rows represent the timeseries of each of the 200 individuals ...

data7Y	A synthesized output variable dataset to fit a quantile panel data model
	on a panel dataset.

### Description

A synthesized output variable dataset to fit a quantile panel data model on a panel dataset.

### Usage

data7Y

### **Format**

A data frame with 100 rows and 200 columns:

columns the individuals

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data8Y	A synthesized output variable dataset to fit a quantile VAR model with
	interactive effects and lag=2.

### Description

A synthesized output variable dataset to fit a quantile VAR model with interactive effects and lag=2.

### Usage

data8Y

#### **Format**

A data frame with 102 rows and 15 columns:

columns the individuals

rows the time points in the timeseries of each individual ...

HOMTEST	

### Description

This function tests homogeneity of the regression coefficients in heterogeneous panel data models with interactive effects.

### Usage

```
HOMTEST(X, Y, Nfactors, Maxit = 100, tol = 0.001)
```

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

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### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated factor loadings for the common factors.
- pvalue: The p-value of the homogeneity test.

### References

Ando, T. and Bai, J. (2015) A simple new test for slope homogeneity in panel data models with interactive effects. Economics Letters, 136, 112-117.

### **Examples**

```
fit <- HOMTEST(data1X,data1Y,2,20,0.5)</pre>
```

HOMTESTGLM	HOMTESTGLM	
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### **Description**

This function tests homogeneity of the regression coefficients in heterogeneous generalized linear models with interactive effects.

### Usage

```
HOMTESTGLM(X, Y, FAMILY, Nfactors, Maxit = 100, tol = 0.001)
```

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
FAMILY	A description of the error distribution and link function to be used in the model just like in glm functions.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

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### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated factor loadings for the common factors.
- pvalue: The p-value of the homogeneity test.

#### References

Ando, T. and Bai, J. (2015) A simple new test for slope homogeneity in panel data models with interactive effects. Economics Letters, 136, 112-117.

### **Examples**

```
fit <- HOMTESTGLM(data2X,data2Y,binomial(link=logit),2,10,0.5)</pre>
```

**HYPTEST** 

**HYPTEST** 

### **Description**

This function undergoes hypothesis testing for regression coefficients obtained from the various functions in the package.

### Usage

```
HYPTEST(
   B,
   B0,
   Se,
   test = "two",
   variables = seq(1, nrow(B)),
   individuals = seq(1, ncol(B))
)
```

В	A dataframe of Coefficients as obtained in the output of any function in the package.
В0	A dataframe of hypothetical coefficients to be evaluated in the test. (nrows should match number of variables and ncols should match number of individuals)
Se	A dataframe of Standard Errors as obtained in the output of any function in the package.

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test	A string to determine what kind of test to run ("two" for two-tailed, "right" for right-tailed and "left for left-tailed).
variables	A list of variables whose coefficients are to be tested. Default is all variables in the B dataframe.
individuals	A list of individuals whose coefficients are to be tested. Default is all individuals in the B dataframe.

### Value

A dataframe of p-values resulting from each individual test.

### **Examples**

```
 fit <- PDMIFLOGIT(data2X, data2Y, 2, 20, 0.5) \\ HYPTEST(fit$Coefficients, data.frame(c(0,1), c(-1,2)), fit$Se, "two", c(1,3), c(1,2)) \\
```

### Description

Under a pre-specified number of groups and the number of common factors, this function implements clustering for N individuals in the panels. Each of individuals in the group are subject to the group-specific unobserved common factors.

### Usage

```
PDMIFCLUST(X, Y, NGfactors, NLfactors, Maxit = 100, tol = 0.001)
```

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
NGfactors	A pre-specified number of common factors across groups (see example).
NLfactors	A pre-specified number of factors in each groups (see example).
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

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#### Value

A list with the following components:

- Label: The estimated group membership for each of the individuals.
- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- GlobalFactors: The estimated common factors across groups.
- GlobalLoadings: The estimated factor loadings for the common factors.
- GroupFactors: The estimated group-specific factors.
- GroupLoadings: The estimated factor loadings for each group.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

#### References

Ando, T. and Bai, J. (2016) Panel data models with grouped factor structure under unknown group membership Journal of Applied Econometrics, 31, 163-191.

Ando, T. and Bai, J. (2017) Clustering huge number of financial time series: A panel data approach with high-dimensional predictors and factor structures. Journal of the American Statistical Association, 112, 1182-1198.

#### **Examples**

```
fit <- PDMIFCLUST(data5X,data5Y,2,c(2,2,2),20,0.5)</pre>
```

**PDMIFCLUSTGLM** 

**PDMIFCLUSTGLM** 

#### **Description**

Under a pre-specified number of groups and the number of common factors, this function implements clustering for N individual units by nonlinear heterogeneous panel data models with interactive effects. Exponential family of distributions are used Each of individuals in the group are subject to the group-specific unobserved common factors.

#### Usage

```
PDMIFCLUSTGLM(X, Y, FAMILY, NLfactors, Maxit = 100, tol = 0.001)
```

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### **Arguments**

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
FAMILY	A description of the error distribution and link function to be used in the model just like in glm functions.
NLfactors	A pre-specified number of factors in each groups (see example).
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

#### Value

A list with the following components:

- Label: The estimated group membership for each of the individuals.
- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- GroupFactors: The estimated group-specific factors.
- GroupLoadings: The estimated factor loadings for each group.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

#### References

Ando, T. and Bai, J. (2016) Panel data models with grouped factor structure under unknown group membership Journal of Applied Econometrics, 31, 163-191.

Ando, T. and Bai, J. (2017) Clustering huge number of financial time series: A panel data approach with high-dimensional predictors and factor structures. Journal of the American Statistical Association, 112, 1182-1198.

```
fit <- PDMIFCLUSTGLM(data6X,data6Y,binomial(link=logit),c(1,1,1),3,0.5)</pre>
```

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PDMIFCOUNT	PDMIFCOUNT		
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### **Description**

Under a known group membership, this function estimates heterogeneous poisson panel data models with interactive effects.

#### Usage

```
PDMIFCOUNT(X, Y, Nfactors, Maxit = 100, tol = 0.001)
```

### **Arguments**

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

#### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated factor loadings for the common factors.
- Predict: The conditional expectation of response variable.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

#### References

Ando, T., Bai, J. and Li, K. (2021) Bayesian and maximum likelihood analysis of large-scale panel choice models with unobserved heterogeneity, Journal of Econometrics.

```
fit <- PDMIFCOUNT(data3X,data3Y,3,30,0.5)</pre>
```

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PDMIFGLM PDMIFGLM
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### **Description**

This function estimates heterogeneous panel data models with interactive effects through generalised linear models.

#### Usage

```
PDMIFGLM(X, Y, FAMILY, Nfactors, Maxit = 100, tol = 0.001)
```

### **Arguments**

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
FAMILY	A description of the error distribution and link function to be used in the model just like in glm functions.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated factor loadings for the common factors.
- Predict: The conditional expectation of response variable.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

### References

Ando, T., Bai, J. and Li, K. (2021) Bayesian and maximum likelihood analysis of large-scale panel choice models with unobserved heterogeneity, Journal of Econometrics.

```
fit <- PDMIFGLM(data2X,data2Y,binomial(link=logit),2,20,0.5)</pre>
```

18 PDMIFLIN

PDMIFLIN	PDMIFLIN		
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### Description

This function estimates heterogeneous panel data models with interactive effects. This function is similar version of PDMIFLING which accommodates a group structure.

#### Usage

```
PDMIFLIN(X, Y, Nfactors, Maxit = 100, tol = 0.001)
```

#### **Arguments**

Χ	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

#### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated factor loadings for the common factors.
- Predict: The conditional expectation of response variable.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

#### References

Ando, T. and Bai, J. (2015) Asset Pricing with a General Multifactor Structure Journal of Financial Econometrics, 13, 556-604.

```
fit <- PDMIFLIN(data1X,data1Y,2)</pre>
```

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PDMIFLING	PDMIFLING	
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### **Description**

Under a known group membership, this function estimates heterogeneous panel data models with interactive effects. Together with the regression coefficients, this function estimates the unobserved common factor structures both for across/within groups.

#### Usage

```
PDMIFLING(X, Y, Membership, NGfactors, NLfactors, Maxit = 100, tol = 0.001)
```

### Arguments

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
Membership	A pre-specified group membership.
NGfactors	A pre-specified number of common factors across groups (see example).
NLfactors	A pre-specified number of factors in each groups (see example).
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

#### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- GlobalFactors: The estimated common factors across groups.
- GlobalLoadings: The estimated factor loadings for the common factors.
- GroupFactors: The estimated group-specific factors.
- GroupLoadings: The estimated factor loadings for each group.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

#### References

Ando, T. and Bai, J. (2015) Asset Pricing with a General Multifactor Structure Journal of Financial Econometrics, 13, 556-604.

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### **Examples**

```
fit <- PDMIFLING(data4X,data4Y,data4LAB,2,c(2,2,2),30,0.1)</pre>
```

**PDMIFLOGIT** 

**PDMIFLOGIT** 

### **Description**

This function estimates heterogeneous logistic panel data models with interactive effects.

#### Usage

```
PDMIFLOGIT(X, Y, Nfactors, Maxit = 100, tol = 0.001)
```

#### **Arguments**

Χ	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

#### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated factor loadings for the common factors.
- Predict: The conditional expectation of response variable.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

### References

Ando, T., Bai, J. and Li, K. (2021) Bayesian and maximum likelihood analysis of large-scale panel choice models with unobserved heterogeneity, Journal of Econometrics.

```
fit <- PDMIFLOGIT(data2X,data2Y,2,20,0.5)</pre>
```

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PDMIFPROBIT	PDMIFPROBIT	

### **Description**

This function estimates heterogeneous probit panel data models with interactive effects.

### Usage

```
PDMIFPROBIT(X, Y, Nfactors, Maxit = 100, tol = 0.001)
```

### **Arguments**

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

#### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated factor loadings for the common factors.
- Predict: The conditional expectation of response variable.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

#### References

Ando, T., Bai, J. and Li, K. (2021) Bayesian and maximum likelihood analysis of large-scale panel choice models with unobserved heterogeneity, Journal of Econometrics.

```
fit <- PDMIFPROBIT(data2X,data2Y,2,20,0.5)</pre>
```

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PDMIFQUANTILE PDMIFQUANTILE	
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### **Description**

This function estimates heterogeneous quantile panel data models with interactive effects.

#### Usage

```
PDMIFQUANTILE(X, Y, TAU, Nfactors, Maxit = 100, tol = 0.001)
```

### Arguments

X	The (NT) times p design matrix, without an intercept where N=number of individuals, T=length of time series, p=number of explanatory variables.
Υ	The T times N panel of response where N=number of individuals, T=length of time series.
TAU	A pre-specified quantile point.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

#### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated quantile point under a given tau.
- Predict: The conditional expectation of response variable.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

#### References

Ando, T. and Bai, J. (2020) Quantile co-movement in financial markets Journal of the American Statistical Association.

```
fit <- PDMIFQUANTILE(data7X,data7Y,0.95,2,10,0.8)</pre>
```

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PDMIFQVAR PDMIFQVAR
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### **Description**

This function estimates heterogeneous quantile panel data VAR models with interactive effects.

### Usage

```
PDMIFQVAR(Y, LAG, TAU, Nfactors, Maxit = 100, tol = 0.001)
```

### Arguments

Υ	The T times N panel of response where N=number of individuals, T=length of time series.
LAG	The number of lags from y_t-1 to y_t-LAG used in the VAR.
TAU	A pre-specified quantile point.
Nfactors	A pre-specified number of common factors.
Maxit	A maximum number of iterations in optimization. Default is 100.
tol	Tolerance level of convergence. Default is 0.001.

### Value

A list with the following components:

- Coefficients: The estimated heterogeneous coefficients.
- Lower05: Lower end (5%) of the 90% confidence interval of the regression coefficients.
- Upper95: Upper end (95%) of the 90% confidence interval of the regression coefficients.
- Factors: The estimated common factors across groups.
- Loadings: The estimated quantile point under a given tau.
- Predict: The conditional expectation of response variable.
- pval: p-value for testing hypothesis on heterogeneous coefficients.
- Se: Standard error of the estimated regression coefficients.

#### References

Ando, T. and Bai, J. (2020) Quantile co-movement in financial markets Journal of the American Statistical Association.

```
fit <- PDMIFQVAR(data8Y,2,0.1,2,5,0.8)
```

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