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Title Calculates the Probability of Informed Trading (PIN)

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Description Estimates the probability of informed trading (PIN) initially introduced by Easley et. al. (1996) <doi:10.1111/j.1540-6261.1996.tb04074.x> . Contribution of the package is that it uses likelihood factorizations of Easley et. al. (2010) <doi:10.1017/S0022109010000074> (EHO factorization) and Lin and Ke (2011) <doi:10.1016/j.finmar.2011.03.001> (LK factorization). Moreover, the package uses different estimation algorithms. Specifically, the grid-search algorithm proposed by Yan and Zhang (2012) <doi:10.1016/j.jbankfin.2011.08.003> , hierarchical agglomerative clustering approach proposed by Gan et. al. (2015) <doi:10.1080/14697688.2015.1023336> and later extended by Ersan and Alici (2016) <doi:10.1016/j.intfin.2016.04.001> .

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Contents

`InfoTrad-package` *Calculates the Probability of Informed Trading (PIN)*

Description

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Author(s)

Duygu Celik and Murat Tinic

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References

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D. Easley, S. Hvidkjaer, and M. O'Hara. Factoring information into returns. Journal of Financial and Quantitative Analysis, 45(2): 293-309, 2010.

Ersan, O., and Alici, A. An unbiased computation methodology for estimating the probability of informed trading (PIN). Journal of International Financial Markets, Institutions and Money, 43(1):74-94, 2016.

Q. Gan, W.C. Wei, and D. Johnstone. A faster estimation method for the probability of informed trading using hierarhical agglomerative clustering. Quantitative Finance, 15(11):1805-1821, 2015

H.W. W. Lin and W. C. Ke. A computing bias in estimating the probability of informed trading. Journal of Financial Markets, 14(4):625-640, 2011.

D. Mullner. fastcluster: Fast hierarchical, agglomerative clustering routines for r and python. Journal of Statistical Software, 53(9):1-18, 2013.

Y. Yan and S. Zhang. An improved estimation method and empirical properties of the probability of informed trading. Journal of Banking & Finance, 36(2):454-467, 2012.

EA *Ersan and Alici (2016) Cluster analysis with the altered steps.*

Description

It estimates PIN using Ersan and Alici (2016) modified clustering algorithm.

Usage

```
EA(data, likelihood = c("LK", "EHO"))
## S3 method for class 'EA_class'
print(obj)
```

Arguments

data	Data frame with 2 variables
likelihood	Character strings for likelihood algorithm. Default is "LK".
obj	object variable

Details

Argument for data must be a data frame with 2 columns that only contain numbers. Not any other type. You do not have to give names to the columns. We will assign first one as "Buy" and second as "Sell", therefore you should put order numbers with respect to this order.

Value

Returns a list of parameter estimates (output)

alpha	A Number
delta	A Number
mu	A Number
eb	A Number
es	A Number
LikVal	A Number
PIN	A Number

Warning

This function does not handle NA values. Therefore the datasets should not contain any missing value. This function do not conduct the iterative estimation procedure proposed in the same paper.

Author(s)

Duygu Celik and Murat Tinic

References

Ersan, Oguz, and Asli Alici . "An unbiased computation methodology for estimating the probability of informed trading (PIN)." Journal of International Financial Markets, Institutions and Money 43 (2016): 74-94.

Examples

```
# Sample Data
#   Buy Sell
#1  350  382
#2  250  500
#3  500  463
#4  552  550
#5  163  200
#6  345  323
#7  847  456
#8  923  342
#9  123  578
#10 349  455

Buy=c(350,250,500,552,163,345,847,923,123,349)
Sell=c(382,500,463,550,200,323,456,342,578,455)
data=cbind(Buy,Sell)

# Parameter estimates using the LK factorization of Lin and Ke (2011)
# with the modified clustering algorithm of Ersan and Alici (2016).
# Default factorization is set to be "LK"

result=EA(data)
print(result)

# Alpha: 0.9511418
# Delta: 0.2694005
# Mu: 76.7224
# Epsilon_b: 493.7045
# Epsilon_s: 377.4877
# Likelihood Value: 43973.71
# PIN: 0.07728924


# Parameter estimates using the EHO factorization of Easley et. al. (2010)
# with the modified clustering algorithm of Ersan and Alici (2016).

result=EA(data,likelihood="EHO")
print(result)

# Alpha: 0.9511418
# Delta: 0.2694005
# Mu: 76.7224
# Epsilon_b: 493.7045
# Epsilon_s: 377.4877
# Likelihood Value: 43973.71
```

```
# PIN: 0.07728924
```

EHO *Likelihood factorization of Easley et. al. (2010) - EHO Factorization*

Description

The function calculates the likelihood factorization of Easley et. al. (2010) and computes paramaters for estimation of PIN value.

Usage

```
EHO(data, fixed = c(FALSE, FALSE, FALSE, FALSE, FALSE))
```

Arguments

data	Data frame with 2 variables
fixed	Initial values for parameters in the following order: alpha, delta, mu, epsilon_b, epsilon_s

Details

In order to use EHO's return in optimization functions, please **omit** second argument. With this way, EHO will return a function instead of a value. Moreover, argument for data must be a data frame with 2 columns that contain numbers. Not any other type.

Value

LK_out	Returns an optim() object including parameter estimates for the likelihood factorization of Easley et. al. (2010)

Warning

This function does not handle NA values. Therefore the datasets should not contain any missing values.

Author(s)

Duygu Celik and Murat Tinic

References

Easley, D., Hvidkjaer, S., & O'Hara, M. Factoring information into returns. Journal of Financial and Quantitative Analysis, 45(2):293-309,2010.

Examples

```
# Sample Data
#   Buy Sell
#1  350  382
#2  250  500
#3  500  463
#4  552  550
#5  163  200
#6  345  323
#7  847  456
#8  923  342
#9  123  578
#10 349  455

Buy<-c(350,250,500,552,163,345,847,923,123,349)
Sell<-c(382,500,463,550,200,323,456,342,578,455)
data=cbind(Buy,Sell)

# Initial parameter values
# par0 = (alpha, delta, mu, epsilon_b, epsilon_s)
par0 = c(0.5,0.5,300,400,500)

# Call EHO function
EHO_out = EHO(data)
model = optim(par0, EHO_out, gr = NULL, method = c("Nelder-Mead"), hessian = FALSE)

## Parameter Estimates
model$par[1] # Estimate for alpha
# [1] 0.9111102
model$par[2] # Estimate for delta
#[1] 0.0001231429
model$par[3] # Estimate for mu
# [1] 417.1497
model$par[4] # Estimate for eb
# [1] 336.075
model$par[5] # Estimate for es
# [1] 466.2539

## Estimate for PIN
(model$par[1]*model$par[3])/((model$par[1]*model$par[3])+model$par[4]+model$par[5])
# [1] 0.3214394
####
```

GAN *GAN et al.(2015) Clustering based PIN Estimates*

Description

It estimates PIN using hierarchical agglomertaive clustering.

Usage

```
GAN(data, likelihood = c("LK", "EHO"))
## S3 method for class 'GAN_class'
print(obj)
```

Arguments

data	Data frame with 2 variables
likelihood	Character strings for likelihood algorithm. Default is "LK".
obj	object variable

Details

Argument for data must be a data frame with 2 columns that only contain numbers. Not any other type. You do not have to give names to the columns. We will assign first one as "Buy" and second as "Sell", therefore you should put order numbers with respect to this order. This package uses the hclust() function of Mullner (2013) to cluster the data at default settings.

Value

Returns a list of parameter estimates (output)

alpha	A Number
delta	A Number
mu	A Number
eb	A Number
es	A Number
LikVal	A Number
PIN	A Number

Warning

This function does not handle NA values. Therefore, the dataset should not contain any missing values.

Author(s)

Duygu Celik and Murat Tinic

References

D. Mullner. fastcluster: Fast hierarchical, agglomerative clustering routines for r and python. Journal of Statistical Software, 53(9):1-18, 2013.

Gan, Q., Wei, W. C., & Johnstone, D. A faster estimation method for the probability of informed trading using hierarchical agglomerative clustering. Quantitative Finance, 15(11), 1805-1821, 2015.

Examples

```
# Sample Data
#   Buy Sell
#1  350  382
#2  250  500
#3  500  463
#4  552  550
#5  163  200
#6  345  323
#7  847  456
#8  923  342
#9  123  578
#10 349  455

Buy<-c(350,250,500,552,163,345,847,923,123,349)
Sell<-c(382,500,463,550,200,323,456,342,578,455)
data<-cbind(Buy,Sell)

# Parameter estimates using the LK factorization of Lin and Ke (2011)
# with the algorithm of Gan et. al. (2015).
# Default factorization is set to be "LK"

result=GAN(data)
print(result)

# Alpha: 0.3999998
# Delta: 0
# Mu: 442.1667
# Epsilon_b: 263.3333
# Epsilon_s: 424.9
# Likelihood Value: 44371.84
# PIN: 0.2044464

# Parameter estimates using the EHO factorization of Easley et. al. (2010)
# with the algorithm of Gan et. al. (2015)

result=GAN(data, likelihood="EHO")
print(result)

# Alpha: 0.3230001
# Delta: 0.4780001
# Mu: 481.3526
# Epsilon_b: 356.6359
# Epsilon_s: 313.136
# Likelihood Value: Inf
# PIN: 0.1884001
```

Description

The function calculates the likelihood factorization of Lin and Ke (2011) and computes paramaters for estimation of PIN value.

Usage

```
LK(data, fixed = c(FALSE, FALSE, FALSE, FALSE, FALSE))
```

Arguments

`data`	Data frame with 2 variables
`fixed`	Initial values for parameters in the following order: alpha, delta, mu, epsilon_b, epsilon_s

Details

In order to use LK's return in optimization functions, please **omit** second argument. With this way, LK will return a function instead of a value. Moreover, argument for data must be a data frame with 2 columns that contain numbers. Not any other type.

Value

`LK_out`	Returns an optim() object including parameter estimates for the likelihood factorization of Lin and Ke (2011)

Warning

This function does not handle NA values. Therefore the datasets should not contain any missing value

Author(s)

Duygu Celik and Murat Tinic

References

Lin, H.W.W. and Ke, W.C. A computing bias in estimating the probability of informed trading. Journal of Financial Markets, 14(4), pp.625-640, 2011.

Examples

```
# Sample Data
#   Buy Sell
#1  350  382
#2  250  500
#3  500  463
#4  552  550
#5  163  200
#6  345  323
#7  847  456
```

```
#8  923  342
#9  123  578
#10 349  455

Buy<-c(350,250,500,552,163,345,847,923,123,349)
Sell<-c(382,500,463,550,200,323,456,342,578,455)
data=cbind(Buy,Sell)

# Initial parameter values
# par0 = (alpha, delta, mu, epsilon_b, epsilon_s)
par0 = c(0.5,0.5,300,400,500)

# Call LK function
LK_out = LK(data)
model = optim(par0, LK_out, gr = NULL, method = c("Nelder-Mead"), hessian = FALSE)

## Parameter Estimates
model$par[1] # Estimate for alpha
# [1] 0.480277
model$par[2] # Estimate for delta
# [1] 0.830850
model$par[3] # Estimate for mu
# [1] 315.259805
model$par[4] # Estimate for eb
# [1] 296.862318
model$par[5] # Estimate for es
# [1] 434.3046

## Estimate for PIN
(model$par[1]*model$par[3])/((model$par[1]*model$par[3])+model$par[4]+model$par[5])
# [1] 0.178391
####
```

YZ *Yan and Zhang (2012) Grid-Search based PIN Estimates*

Description

It estimates PIN using Yan and Zhang (2012) algorithm.

Usage

```
YZ(data, likelihood = c("LK", "EHO"))
## S3 method for class 'YZ_class'
print(obj)
```

Arguments

data	Data frame with 2 variables
likelihood	Character strings for likelihood algorithm. Default is "LK".
obj	object variable

Details

Argument for data must be a data frame with 2 columns that only contain numbers. Not any other type. You do not have to give names to the columns. We will assign first one as "Buy" and second as "Sell", therefore you should put order numbers with respect to this order.

Value

Returns a list of parameter estimates (output)

`alpha`	A Number
`delta`	A Number
`mu`	A Number
`eb`	A Number
`es`	A Number
`LikVal`	A Number
`PIN`	A Number

Warning

This function does not handle NA values. Therefore the datasets should not contain any missing value

Author(s)

Duygu Celik and Murat Tinic

References

Y. Yan and S. Zhang. An improved estimation method and empirical properties of the probability of informed trading. Journal of Banking & Finance, 36(2):454-467, 2012.

Examples

```
# Sample Data
#   Buy Sell
#1  350  382
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#10 349  455

Buy<-c(350,250,500,552,163,345,847,923,123,349)
Sell<-c(382,500,463,550,200,323,456,342,578,455)
data<-cbind(Buy,Sell)
```

```
# Parameter estimates using the LK factorization of Lin and Ke (2011)
# with the algorithm of Yan and Zhang (2012).
# Default factorization is set to be "LK"

result=YZ(data)
print(result)

# Alpha: 0.3999999
# Delta: 0
# Mu: 442.1667
# Epsilon_b: 263.3333
# Epsilon_s: 424.9
# Likelihood Value: 44371.84
# PIN: 0.2004457

# Parameter estimates using the EHO factorization of Easley et. al. (2010)
# with the algorithm of Yan and Zhang (2012).

result=YZ(data,likelihood="EHO")
print(result)

# Alpha: 0.9000001
# Delta: 0.9000001
# Mu: 489.1111
# Epsilon_b: 396.1803
# Epsilon_s: 28.72002
# Likelihood Value: Inf
# PIN: 0.3321033
```

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