



order α : non-parametric order- α Efficiency Analysis for Stata

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Introduction

- ▶ Countless empirical efficiency analyses in economics
 - ▶ **Decision Making Units/production units (DMUs)**: firms, states, universities, etc.
- ▶ Measuring distance to *Production Possibility Frontier*
 - ▶ alternatively: cost- or profit-frontier
- ▶ *Output-orientated* and *Input-orientated* efficiency
 1. How much additional output (y_1, \dots, y_L) can be produced leaving inputs (x_1, \dots, x_M) unchanged?
 2. How much input (x_1, \dots, x_M) can be saved leaving outputs (y_1, \dots, y_L) unchanged?
- ▶ Prime purpose: estimating efficiency scores θ_i for individual DMUs
 - ▶ (Input-oriented) efficiency score $(0, 1]$: factor by which input consumption can *proportionally* be reduced

Parametric Approaches: Stochastic Frontier Models

- ▶ Introduced by Aigner et al. (1977)
- stata command **frontier**
- ▶ Production- (cost-, profit-) function estimated
- ▶ Linear regression model augmented by additional non-positive (reps. non-negative) error term η_i
- ▶ Maximum-likelihood estimation
- ▶ Distributional assumptions for η_i and v_i required

$$\log(y_i) = \beta_0 + \sum_{m=1}^M \beta_m x_{mi} + v_i - \eta_i \quad \text{with} \quad \eta_i \geq 0$$

- ▶ (Output-oriented) technical efficiency score computed as:

$$\hat{\theta}_i^{SF} = E(\exp(-\eta_i) | v_i - \eta_i)$$

Non-Parametric Approaches

1. Data envelopment Analysis (DEA)

- ▶ Introduced by Charnes et al. (1978)
- stata ado-file **dea** (Ji & Lee, 2010)
- ▶ Linear programming approach
- ▶ Envelopes data by *piecewise-linear convex* hull
- ▶ Solution for θ (input-oriented) efficiency score $\hat{\theta}_i^{DEA}$:

$$\begin{array}{ll}
 \min_{\theta, \lambda} \theta & \text{subject to} \\
 \theta x_{mi} - \sum_{j=1}^N \lambda_j x_{mj} \geq 0 & m = 1, \dots, M \\
 \sum_{j=1}^N \lambda_j y_{lj} - y_{li} \geq 0 & l = 1, \dots, L \\
 \lambda_j \geq 0 & \forall j
 \end{array}$$

Non-Parametric Approaches II

2. Free Disposal Hull (FDH)

- ▶ Introduced by Deprins et al. (1984)
- ▶ Based on principle of weak dominance
 - ▶ DMU i compared to those DMUs that produce at least the same amount of any output
 - ▶ DMU with minimal input use serves as reference
- ▶ Envelopes data by piecewise linear *non-convex* (step) hull
- ▶ (Input-oriented) efficiency scores computed as:

$$\hat{\theta}_i^{FDH} = \min_{j=1, \dots, N | y_{ij} \geq y_{ii} \forall l} \left\{ \max_{m=1, \dots, M} \left\{ \frac{x_{mj}}{x_{mi}} \right\} \right\}$$

Parametric vs. Non-Parametric Approaches

1. Parametric Approach (shortcomings):

- ▶ Relies on distributional assumptions
- ▶ Functional form for production technology required
- ▶ Production function ill-suited regression model (endogeneity of inputs)
- ▶ Accommodates only single-output technologies

2. Non-Parametric Approach (shortcomings):

- ▶ No well-defined data generating process (→ recent advances)
 - ▶ Deterministic approach
- ⇒ Extremely vulnerable to outliers and measurement error

Partial Frontier Approaches

- ▶ Sensitivity to outliers reduced by allowing for super-efficient DMUs
- ▶ Super-efficient DMUs located beyond production-possibility frontier
- ▶ partial frontier envelopes just a sub-sample of the data
- ▶ Super-efficiency: (input-oriented) efficiency score > 1

Partial frontier approaches that generalize FDH:

1. Order- m efficiency (Cazals et al., 2002)
2. Order- α efficiency (Aragon et al., 2005)

Order- m Efficiency

- ▶ Adds a 'layer of randomness' to FDH
- ▶ Series of FDH analyses using an randomly drawn sup-samples of size m
- ▶ DMU may or may not serve as its own peer
- ▶ Re-sampling repeated D times
- ▶ $\hat{\theta}_i^{OM}$: average of D efficiency scores
- ▶ Shortcoming: very time consuming (for large data sets)
 - Rules virtually out statistical inference based on bootstrapping
 - Determining appropriate value for m may require trying numerous values

Order- α Efficiency

- ▶ Chooses $[100 - \alpha]^{th}$ (α^{th}) percentile, with $0 \leq \alpha \leq 100$, rather than a minimum (maximum) as an efficiency benchmark
 - ▶ Not **min.** input consumption, but $[100 - \alpha]^{th}$ **percentile** (among peer-DMUs) serves as reference
 - ▶ FDH represents a special case of order- α (for $\alpha = 100$)
 - ▶ No re-sampling \rightarrow less time consuming than order- m
- \rightarrow New stata ado-file **orderalpha**

Order- α Input-Oriented Efficiency:

$$\hat{\theta}_i^{OA_{input}} = P_{100-\alpha} \left\{ \max_{m=1, \dots, M} \left\{ \frac{x_{mj}}{x_{mi}} \right\} \right\}_{j=1, \dots, N | y_{lj} \geq y_{li} \forall l}$$

Order- α Output-Oriented Efficiency:

$$\hat{\theta}_i^{OA_{output}} = P_{\alpha} \left\{ \min_{l=1, \dots, L} \left\{ \frac{y_{lj}}{y_{li}} \right\} \right\}_{j=1, \dots, N | x_{mj} \leq x_{mi} \forall m}$$

Illustration for Single-Input Single-Output case

(districts-level health production in Germany)

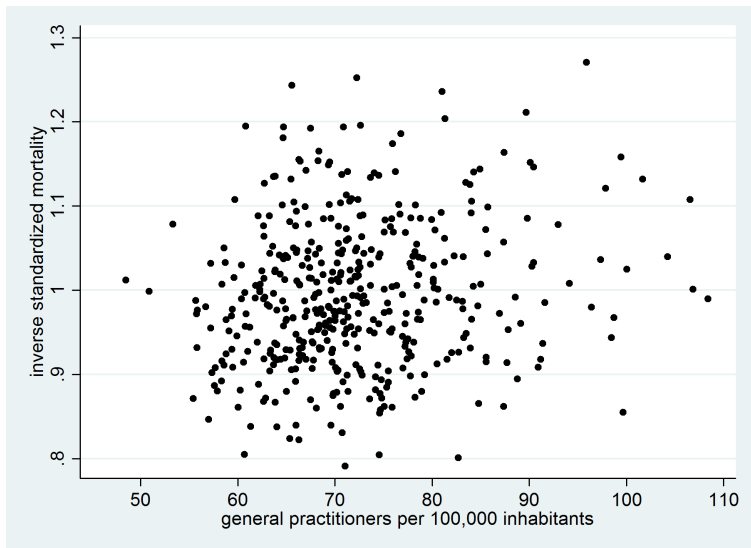


Illustration for Single-Input Single-Output case

(districts-level health production in Germany)

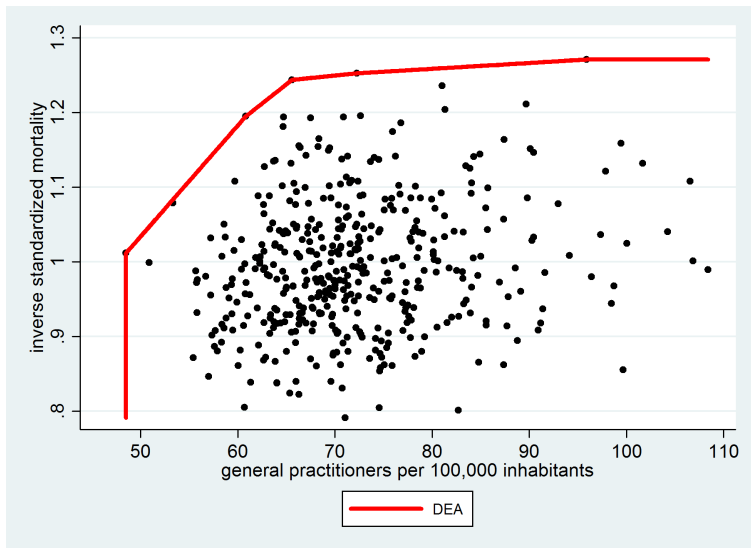


Illustration for Single-Input Single-Output case

(districts-level health production in Germany)

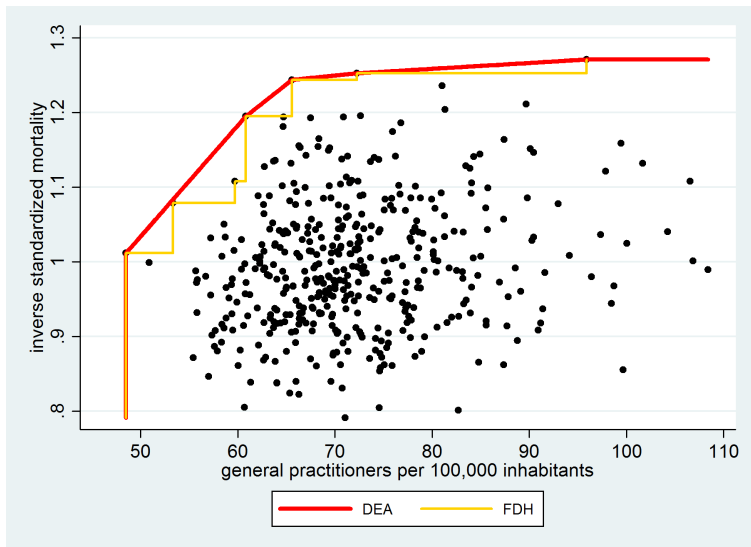


Illustration for Single-Input Single-Output case

(districts-level health production in Germany)

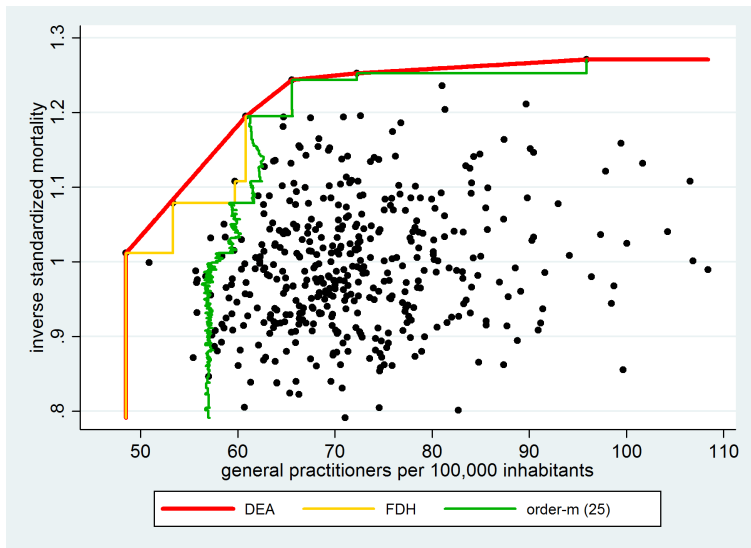


Illustration for Single-Input Single-Output case

(districts-level health production in Germany)

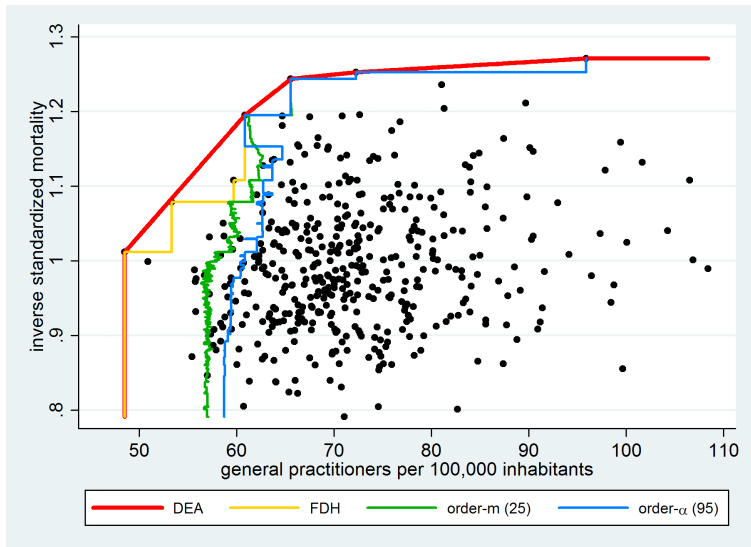
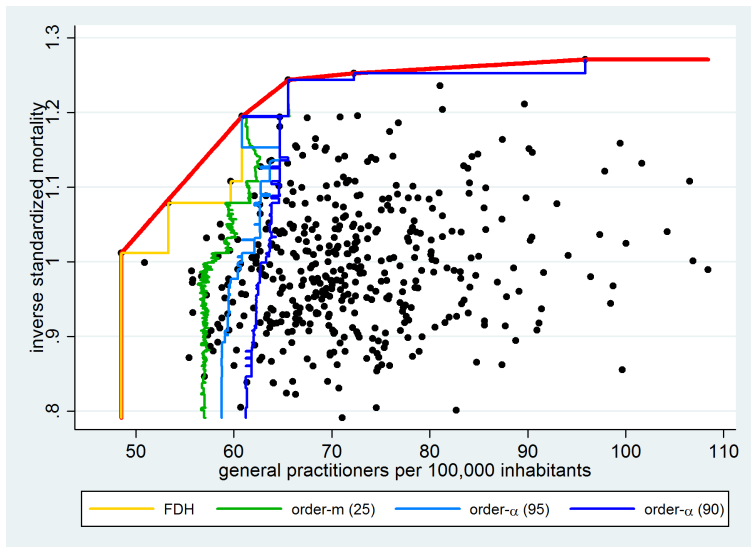


Illustration for Single-Input Single-Output case

(districts-level health production in Germany)



orderalpha: Syntax

Viewer (#5) [help orderalpha]

help orderalpha

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help orderalpha

Title

orderalpha — order-alpha efficiency analysis

Syntax

orderalpha *varname* [*if*] [*in*], **inputs**(*varlist1*) **outputs**(*varlist2*) [*options*]

<i>technology_definition</i>	description
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Model

<i>varname</i>	identifier
inputs (<i>varlist1</i>)	list of input variables
outputs (<i>varlist2</i>)	list of output variables

options description

Main

ort (<i>string</i>)	<i>string</i> may be output or input ; default is ort(input)
alpha (<i>#</i>)	set benchmark percentile; default is alpha(100)

SE/Bootstrap

bootstrap	perform bootstrap using 100 replications
reps (<i>#</i>)	set (temporary) number <i>#</i> of bootstrap replications and perform bootstrap
tune (<i>#</i>)	set tuning parameter for subsampling bootstrap; values within the [0.5,1] interval are allowed

Reporting

level (<i>#</i>)	set confidence level; default is level(95)
table (<i>string</i>)	display table of results; <i>string</i> may be scores or full , no table is the default
dots (<i>#</i>)	display replication/loop dots; <i>#</i> may be 1 or 2 , no dots is the default
invert	report reciprocal of output-oriented efficiency scores

Generate

generate (<i>newvarlist</i>)	supply names for new variables, containing efficiency scores, ranks, and reference <i>dmus</i>
replace	replace existing variables in <i>newvarlist</i>
nogenerate	do not create new variables containing results

weights are not allowed; see [weight](#).
bootstrap, **by**, and **svy** are not allowed; see [prefix](#).

Description

RWI

Application: Regional Health Production in Bavaria

- ▶ Decision making units:
 - ▶ Districts ('kreis') in Bavaria
 - ▶ Cross section: year 2004
 - ▶ # of observations: 96
- ▶ Inputs:
 - i. Resident medical specialists per 100 000 inhabitants ('specialists')
 - ii. General practitioners per 100 000 inhabitants ('gps')
 - iii. Hospital beds per 10 000 inhabitants ('beds')
- ▶ Outputs:
 - i. Inverse normalized (to district demographics) *mortality* ('survival')

orderalpha: Screen Shot (regional health production in Bavaria)

```
. orderalpha kreis if jahr == 2004 & regio >= 9000 & regio < 10000, inputs(specialists
> gps beds) outputs(survival) alpha(95) replace table(full) reps(200) gen(escore erank
> refdmu)
```

order-alpha(95) input-oriented efficiency scores estimated (variable escore)

```
Number of dmus          = 96
Number of inputs        = 3
Number of outputs       = 1
Mean efficiency         = .9229
Median efficiency       = .9023
Share of efficient dmus = .2083
Share of super-efficient dmus = .0833
```

dmu (kreis)	Eff. Score	Std. Err.	z Stat.	Eff. Rank	Ref. DMU
Aichach-Friedbe	1	.1322505	0	9	Aichach-Fr
Altötting, Land	.8259127	.0254798	6.832376	65	Eichstätt,
Amberg, krsfr.	.650349	.0157609	22.18474	89	Aschaffenb
Amberg-Sulzbach	1.104511	.3570595	.2926985	5	Landshut,
Ansbach, Landkr	1	.1559544	0	9	Ansbach, L
Ansbach, krsfr.	.7476639	.0246253	10.24702	77	Freising,
Aschaffenburg,	2.423852	.7841469	1.815798	2	Aichach-Fr
Aschaffenburg,	.8616958	.035989	3.842959	57	Aschaffenb
Augsburg, Landk	.9280609	.1040334	.6914996	43	Regensburg
Augsburg, krsfr	.7404743	.0212208	12.2298	79	Landshut,
Bad Kissingen,	.8131074	.0247055	7.564818	70	Straubing-
Bad Tölz-wolfra	1	.5837663	0	9	Bad Tölz-w
Bamberg, Landkr	1.082539	.3349955	.2463872	6	Coburg, La
Bamberg, krsfr.	.5726839	.0222016	19.24707	96	Erlangen-H
Bayreuth, Landk	1.435832	.5069627	.8596922	3	Bamberg, L
Bayreuth, krsfr	.7540368	.0182821	13.45376	76	Aschaffenb
Berchtesgaden	.787752	.056467	3.7588	73	Erding, La

orderalpha: Saved Results

Viewer (#5) [help orderalpha]

help orderalpha

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Saved results

orderalpha saves the following in **e()**:

Scalars

e(N)	number of dmus
e(alpha)	value of alpha (#)
e(inputs)	number of inputs
e(outputs)	number of outputs
e(efficient)	share of efficient dmus
e(super)	share of super-efficient dmus
e(mean_e)	mean estimated efficiency
e(med_e)	median estimated efficiency
e(level)	confidence level

Macros

e(cmd)	orderalpha
e(cmdline)	command as typed
e(title)	Order-alpha efficiency analysis
e(dmuid)	<i>varname</i> (name of dmu-identifier)
e(mode)	either order-alpha or FDH
e(saved)	names of variables saved (not saved for option nogenerate)
e(table)	scores , full , or no
e(invert)	either inverted or notinverted (not saved for ort(input))
e(ort)	either input or output
e(properties)	either b or b v
e(depvar)	dmu

Matrices

e(b)	vector of estimated efficiency scores (<i>colnames</i> are of the form <i>varname:value_of_varname</i>)
e(ranks)	vector of efficiency ranks (<i>colnames</i> are of the form <i>varname:value_of_varname</i>)
e(reference)	matrix of names of reference dums (not saved if <i>varname</i> is a string variable)

Functions

e(sample)	marks estimation sample
------------------	-------------------------

orderalpha, boot reps(#) additionally saves the following in **e()**:

Scalars

e(N_reps)	number of bootstrap repetitions
e(tune)	value of tuning parameter
e(N_bs)	size of bootstrap samples

Macros

e(vce)	bootstrap
e(vctype)	Bootstrap

Matrices

e(v)	bootstrap variance-covariance matrix for estimated efficiency scores
e(bias)	estimated biases
e(reps)	number of nonmissing results
e(b_bs)	bootstrap estimates

Order- α based outlier-detection

Idee proposed by Daraio & Simar (2007):

- ▶ Increasing the value of α reduces number of DMUs classified as “super-efficient”
- ▶ In the absence of outliers: share of super-efficient DMUs should decrease smoothly
- ▶ Discontinuity points at presence of outliers
- ⇒ DMUs still classified “super-efficient” for $\alpha \geq \alpha^{disc}$ (point of discontinuity) most likely outliers
- ⇒ To be excluded from efficiency analysis applying FDH or DEA

Implementing approach of Daraio & Simar (2007)

oaoutlier:

- ▶ Carries out series of order- α analyses
- ▶ Plots share of super-efficient units against α
- ▶ Suggests two local and one global rules for detecting discontinuities:
 1. α for which the twice differenced series takes minimum value (following a non-negative one)
 2. Values of α for which negative values persist after repeatedly smoothing twice differenced series by running odd-spaced median smoothers (\rightarrow **smooth**)
 3. α that minimizes BIC for splitting the series into two parts and fitting linear (quadratic) functions to each

oaooutlier: Syntax

Viewer (#6) [help oaooutlier]

help oaooutlier

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help oaooutlier

Title

oaooutlier — outlier detection based on order-alpha efficiency analysis

Syntax

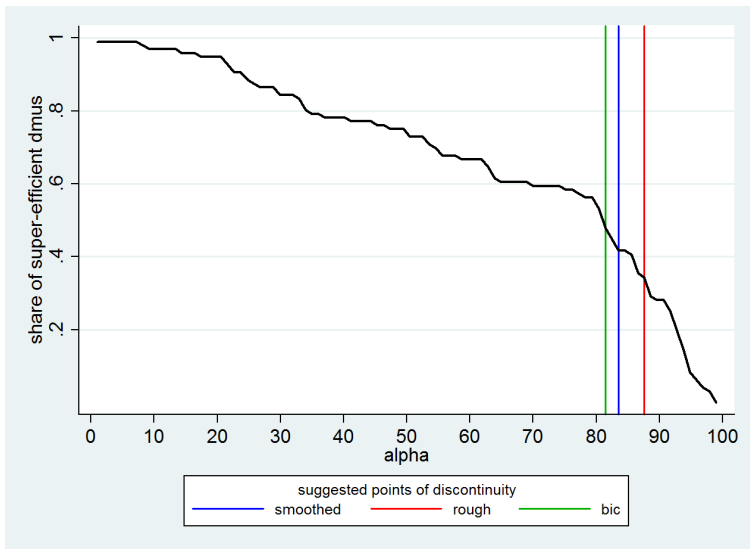
oaooutlier *varname* [*if*] [*fn*], **inputs**(*varlist1*) **outputs**(*varlist2*) [*options*]

<i>technology_definition</i>	<i>description</i>
Model	
<i>varname</i>	identifier
inputs (<i>varlist1</i>)	list of input variables
outputs (<i>varlist2</i>)	list of output variables
<i>options</i>	description
Main	
ort (<i>string</i>)	<i>string</i> may be output or input ; default is ort(input)
nalpha (<i>#</i>)	try <i>#</i> values for alpha; the maximum allowed value is <i>N</i> which is also the default
Detection	
nobic	do not suggest discontinuities based on BIC
norough	do not suggest discontinuities based on rough series of difference in differences
nosmooth	do not suggest discontinuities based on smoothed series of differences in differences
smoother (<i>string</i>)	use smoother <i>string</i> for smoothing series of differences in differences
Reporting	
noplot	suppress plotting series of share of super-efficient dmus
dots	display loop dots

weights are not allowed; see [weight](#).
bootstrap, **by**, and **svy** are not allowed; see [prefix](#).

Description

Graph. output oaoutlier: (regional health production in Bavaria)



oaoutlier: Saved Results

Viewer (#6) [help oaoutlier]

Help oaoutlier

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Examples

- . oaoutlier firm, inputs(capital labor energy) outputs(durables perishables) ort(output) dots
Here *oaoutlier* and *orderalpha* precede *dea* and already respect *deas* requirements for naming variables:
- . oaoutlier dmu, inputs(i_capital i_labor i_energy) outputs(o_durables o_perishables) ort(output)
- . orderalpha dmu, inputs(i_capital i_labor i_energy) outputs(o_durables o_perishables) ort(output) invert alpha('r(asmooth#)') gen(oaeffi)
- . dea if oaeffi <= 1, rts(vrs) ort(o)

Saved results

orderalpha saves the following in **r()**:

Scalars

r(asmooth#)	point of discontinuity suggested by local rule using smoothed series (# may take values up to 3)
r(arough#)	point of discontinuity suggested by local rule using rough series (# may take values up to 3)
r(abicl)	point of discontinuity suggested by global rule
r(ssmooth#)	share of super-efficients dmus that corresponds to r(asmooth#)
r(srough#)	share of super-efficients dmus that corresponds to r(arough#)
r(sbicl)	share of super-efficients dmus that corresponds to r(abicl)

Macros

r(cmd)	oaoutlier
r(cmdline)	command as typed
r(title)	Order-alpha based outlier detection
r(ort)	either input or output

Matrices

r(oareult)	Mx2 matrix of values tried for alpha and corresponding shares of super-efficient dmus
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References

Daraio, C. and L. Simar (2007). *Advanced robust and nonparametric methods in efficiency analysis: Methodology and applications*. Springer, New York.

Also see

Manual: [\[R\] smooth](#)

Help: [\[R\] smooth](#)

online: [dea](#), [orderalpha](#)

References

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