USER MANUAL

DASP version 1.3

DASP: Distributive Analysis Stata Package

By

Abdelkrim Araar,
Jean-Yves Duclos

Université Laval
World Bank, PEP and CIRPÉE

June 2007

Table of contents

List of Figures

1 Introduction

The STATA software has become a very popular tool to transform and process data. It comes with a large number of basic data management modules that are highly efficient for transformation of large datasets. The flexibility of STATA also enables programmers to provide specialized ".ado" routines to add to the power of the software. This is indeed how *DASP* interacts with STATA. *DASP,* which stands for *Distributive Analysis STATA Package*, is mainly designed to assist researchers and policy analysts interested in conducting distributive analysis with STATA. In particular, *DASP* is built to:

- Estimate the most popular statistics (indices, curves) used for the analysis of poverty, inequality, social welfare, and equity;
- Estimate the differences in such statistics;
- Estimate standard errors and confidence intervals by taking full account of survey design;
- Support distributive analysis on more than one data base;
- Perform the most popular poverty and decomposition procedures;
- Check for the ethical robustness of distributive comparisons;
- Unify syntax and parameter use across various estimation procedures for distributive analysis.

For each *DASP* module, three types of files are provided:

*.ado: This file contains the program of the module
*.hlp : This file contains help material for the given module
*.dlg : This file allows the user to perform the estimation using the module's dialog box

The *.dlg files in particular makes the *DASP* package very user friendly and easy to learn. When these dialog boxes are used, the associated program syntax is also generated and showed in the **review window**. The user can save the contents of this window in a *.do file to be subsequently used in another session.

2 *DASP* and Stata versions

DASP requires

- STATA version 9.2 or higher
- ado files must be updated

To update the executable file (from 9.0 to 9.2) and the ado files, see:

http://www.stata.com/support/updates/

3 Installing and updating the *DASP* package

In general, the *.ado files are saved in the following main directories:

Priority	Directory	Sources
1	**UPDATES:**	**Official updates of STATA *.ado files**
2	**BASE:**	***.ado files that come with the installed STATA software**
3	**SITE:**	***.ado files downloaded from the net**
4	**PLUS:**	**..**
5	**PERSONAL:**	**Personal *.ado files**

3.1 installing DASP modules).

a. Unzip the file ***dasp.zip*** in the directory c:
b. Make sure that you have ***c:/dasp/dasp.pkg*** or ***c:/dasp/stata.toc***
c. In the Stata **command windows**, type the syntax
d.

    net from c:/dasp

Figure 1: Ouput of *net describe dasp*
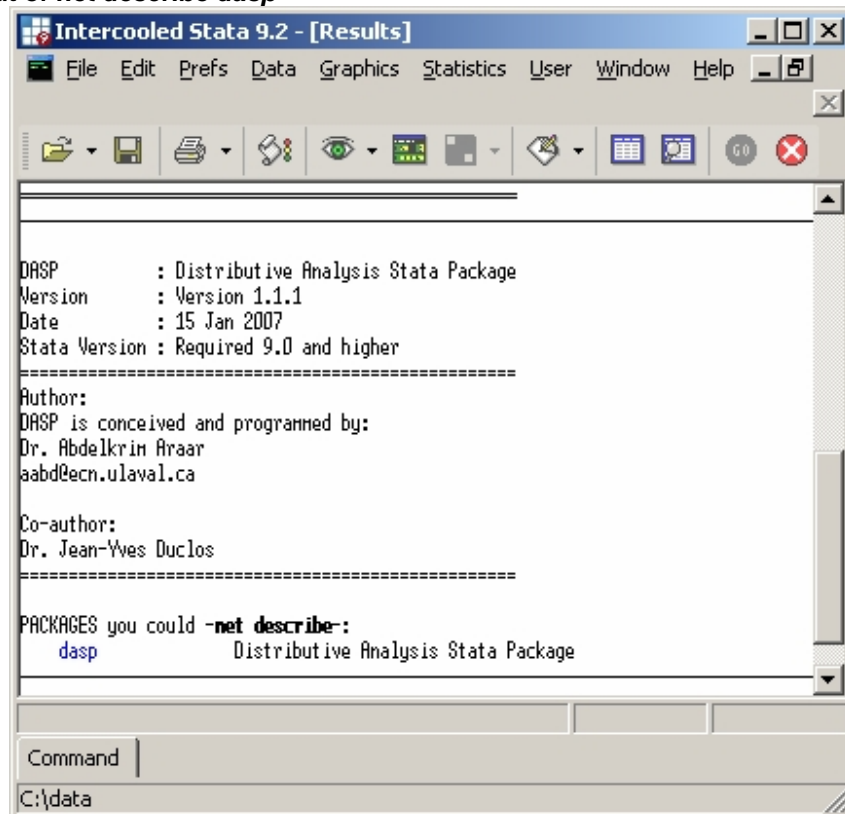
e. Type the syntax
    net install dasp_p1.pkg, force replace
    net install dasp_p2.pkg, force replace

3.2 Adding the *DASP* **submenu to STATA's main menu.**

With STATA 9, sub menus can be added to the menu item **User**.

Figure 2: *DASP* submenu
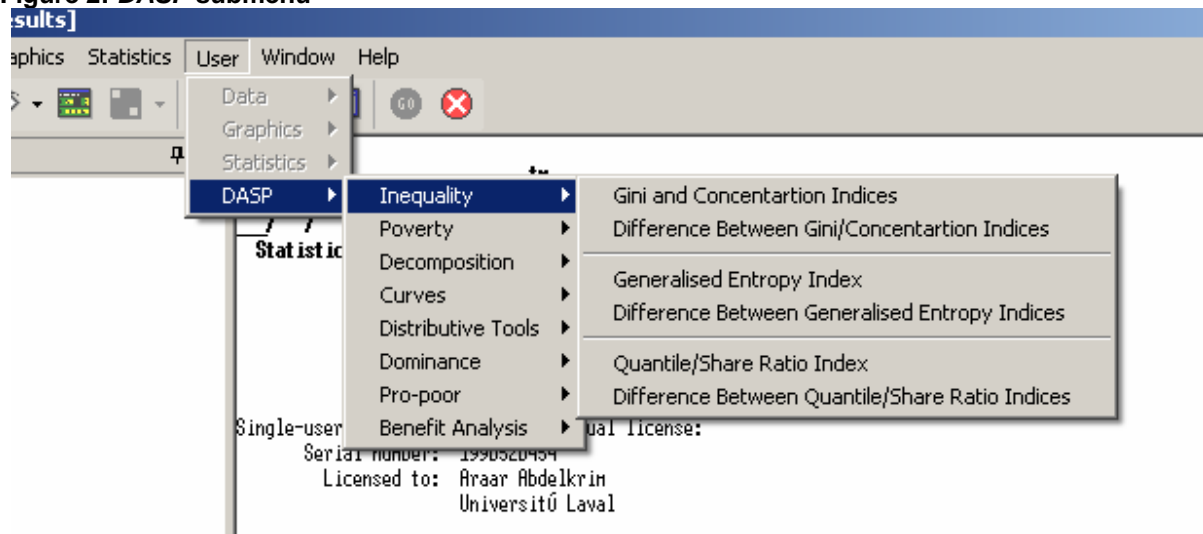
To add the *DASP* sub menus, the file ***profile.do*** (which is provided with the *DASP* package) must be copied into the ***PERSONAL*** directory. If the file ***profile.do*** already exists, add the contents of the *DASP* –provided ***profile.do*** file into that existing file and save it. To check if the file ***profile.do*** already exists, type the command: *findfile profile.do .*

4 *DASP* **and data files**

DASP makes it possible to use simultaneously more than one data file. The user should, however, "initialize" each data file before using it with *DASP*. This initialization is done by:

1. Labeling variables and values for categorical variables;
2. Initializing the sampling design with the command *svyset*;
3. Saving the initialized data file.

Users are recommended to consult appendices A, B and C,

5 **Main variables for distributive analysis**

VARIABLE OF INTEREST. This is the variable that usually captures living standards. It can represent, for instance, income per capita, expenditures per adult equivalent, calorie intake, normalized height-for-age scores for children, or household wealth.

SIZE VARIABLE. This refers to the "ethical" or physical size of the observation. For the computation of many statistics, we will indeed wish to take into account how many relevant individuals (or statistical units) are found in a given observation.

GROUP VARIABLE. (This should be used in combination with GROUP NUMBER.) It is often useful to focus one's analysis on some population subgroup. We might, for example, wish to estimate poverty within a country's rural area or within female-headed families. One way to do this is to force *DASP* to focus on a population subgroup defined as those for whom some GROUP VARIABLE (say, area of residence) equals a given GROUP NUMBER (say 2, for rural area).

SAMPLING WEIGHT. Sampling weights are the inverse of the sampling probability. This variable should be set upon the initialization of the dataset.

6   How can *DASP* commands be invoked?

STATA commands can be entered directly into a **command window**:
Figure 3: Using *DASP* with a command window
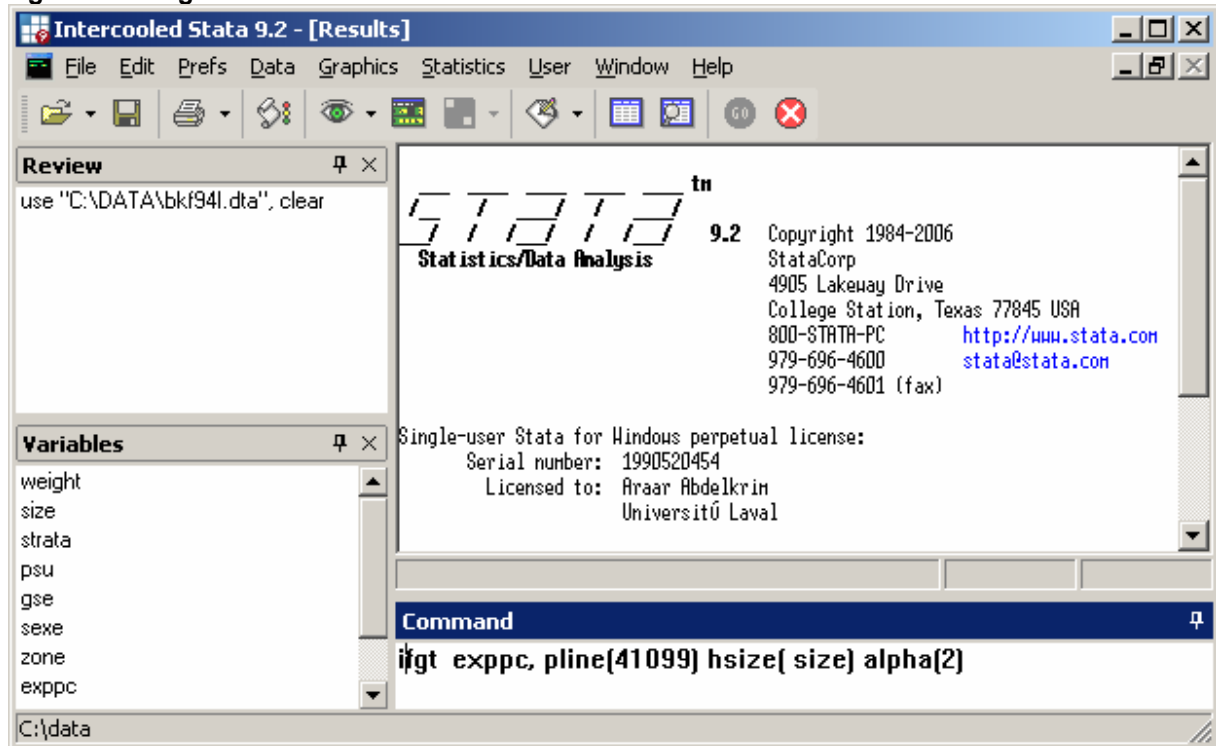
An alternative is to use dialog boxes. For this, the command *db* should be typed and followed by the name of the relevant *DASP* module.
Example:
db ifgt

7   How can help be accessed for a given *DASP* module?

Type the command *help* followed by the name of the relevant *DASP* module.

Example:
help ifgt

Figure 4: Accessing help on *DASP*
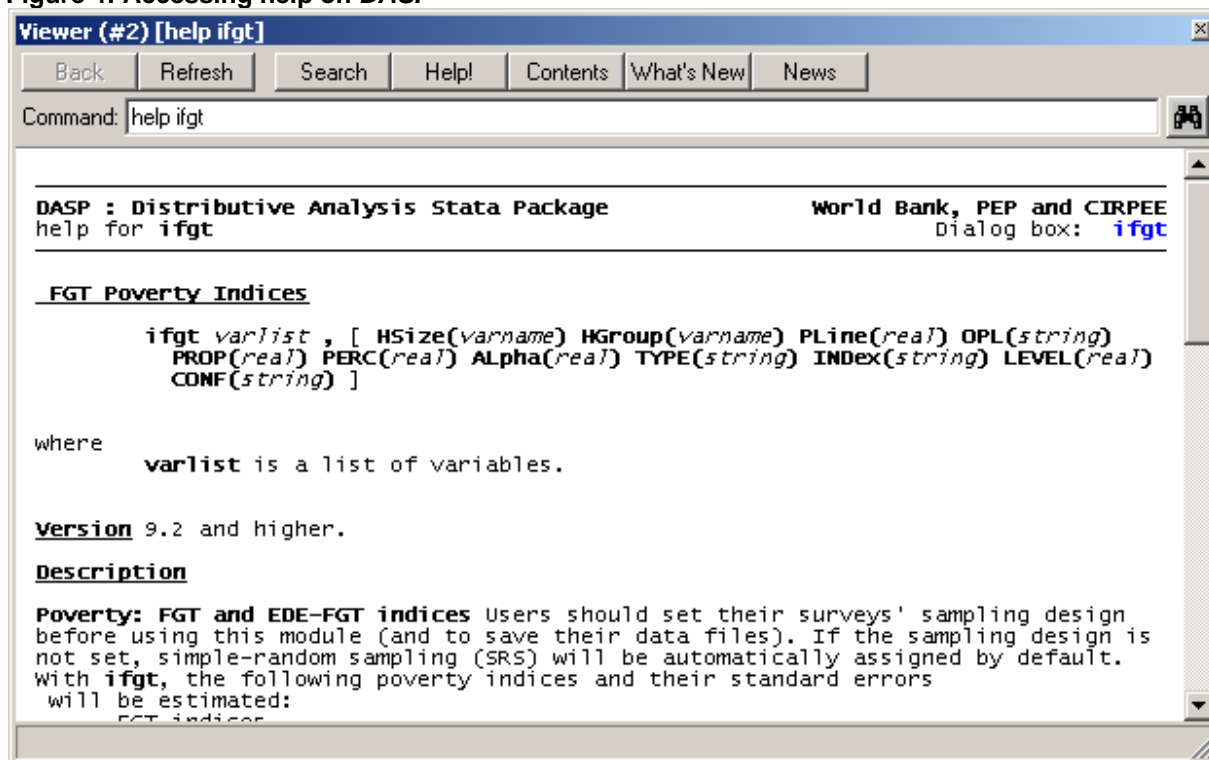
8 **Applications and files in *DASP***

Two main types of applications are provided in *DASP*. For the first one, the estimation procedures require only one data file. In such cases, the data file in memory is the one that is used (or "loaded"); it is from that file that the relevant variables must be specified by the user to perform the required estimation.

Figure 5: **Estimating FGT poverty with one distribution**
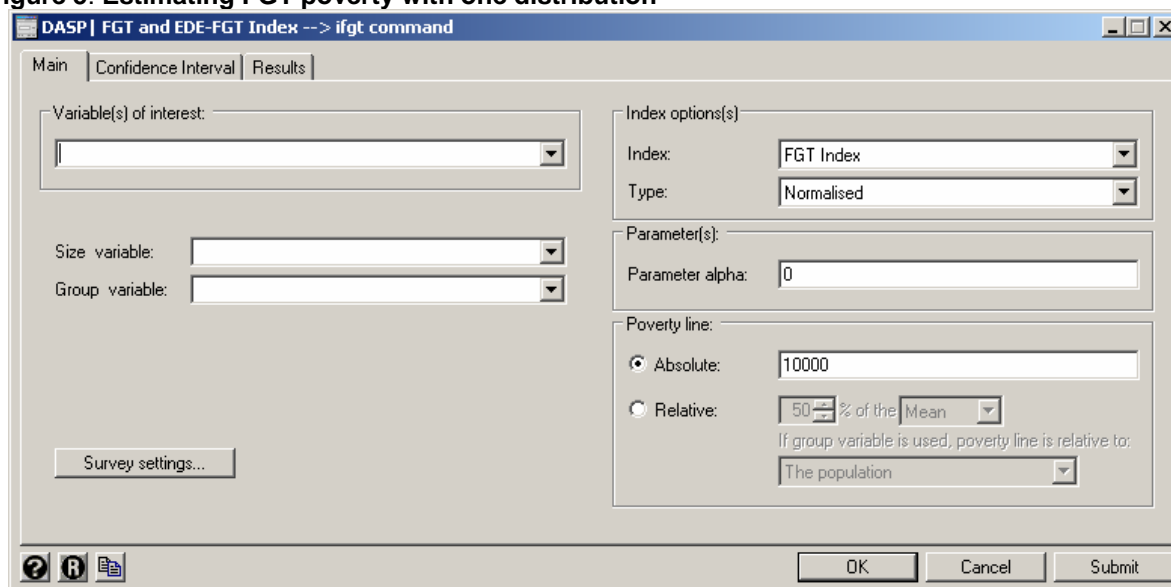
For the second type of applications, two distributions are needed. For each of these two distributions, the user can specify the currently-loaded data file (the one in memory) or one saved on disk.

Figure 6: Estimating FGT poverty with two distributions

Notes:
1. *DASP* considers two distributions to be statistically dependent (for statistical inference purposes) if the same data set is used (the same loaded data or data with the same path and filename) for the two distributions.
2. If the option DATA IN FILE is chosen, the keyboard must be used to type the name of the required variables.

9   Basic Notation

The following table presents the basic notation used in *DASP*'s user manual.

Symbol	Indication
y	variable of interest
i	observation number
y_i	value of the variable of interest for observation i
hw	sampling weight
hw_i	sampling weight for observation i
hs	size variable
hs_i	size of observation i (for example the size of household i)
w_i	$hw_i * hs_i$
hg	group variable
hg_i	group of observation i.
w_i^k	$sw_i^k = sw_i$ if $hg_i = k$, and 0 otherwise.

n   sample size

For example, the mean of *y* is estimated by *DASP* as $\hat{\mu}$:

$$\hat{\mu} = \frac{\sum_{i=1}^{n} w_i \, y_i}{\sum_{i=1}^{n} w_i}$$

10 *DASP* and poverty indices

10.1 FGT and EDE-FGT poverty indices (ifgt).

The non-normalised Foster-Greer-Thorbecke or FGT index is estimated as

$$\widehat{P}(z;\alpha) = \frac{\sum_{i=1}^{n} w_i \, (z - y_i)_+^{\alpha}}{\sum_{i=1}^{n} w_i}$$

where *z* is the poverty line and $x_+ = \max(x, 0)$. The usual normalised FGT index is estimated as

$$\widehat{\overline{P}}(z;\alpha) = \widehat{P}(z;\alpha)/(z)^{\alpha}$$

The EDE-FGT index is estimated as:

$$\widehat{EDE}(P(z;\alpha)) = \left(\widehat{P}(z;\alpha)\right)^{1/\alpha} \quad for \quad \alpha > 0$$

- There exist three ways of fixing the poverty line:

  1- Setting a deterministic poverty line;
  2- Setting the poverty line to a proportion of the mean;
  3- Setting the poverty line to a proportion of a quantile *Q(p)*.

- The user can choose the value of parameter α.

- The user can select more than one variable of interest simultaneously. For example, one can estimate poverty by using simultaneously *per capita* consumption and *per capita* income.

- A group variable can be used to estimate poverty at the level of a categorical group. If a group variable is selected, only the first variable of interest is then used.

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 6 decimals; this can be changed.

Interested users are encouraged to consider the exercises that appear in Section **19.1**

10.2 Difference between FGT indices (difgt)

This module estimates differences between the FGT indices of two distributions.

For each of the two distributions:

- There exist three ways of fixing the poverty line:

  1- Setting a deterministic poverty line;
  2- Setting the poverty line to a proportion of the mean;
  3- Setting the poverty line to a proportion of a quantile *Q(p)*.

- One variable of interest should be selected.

- Conditions can be specified to focus on specific population subgroups.

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 6 decimals; this can be changed.

- A level for the parameter α can be chosen for each of the two distributions.

Interested users are encouraged to consider the exercises that appear in Section **19.2.**

10.3 DASP and multidimensional poverty indices (imdpov)

The general form of an additive multidimensional poverty index is:

$$P(X,Z) = \frac{\sum_{i=1}^{n} w_i p(X_i, Z)}{\sum_{i=1}^{n} w_i}$$

where $p(X_i, Z)$ is individual *I*'s poverty function (with vector of attributes $X_i = \left(x_{i,1}, ..., x_{i,J}\right)$ and vector of poverty lines $Z = \left(z_1, ..., z_J\right)$), determining *I*'s contribution to total poverty $P(X,Z)$.

[1] Chakravarty et al (1998) index

$$p(X_i, Z) = \sum_{j=1}^{J} a_j \left(\frac{z_j - x_{i,j}}{z_j}\right)_+^{\alpha}$$

[2] Extended Watts index

$$p(X_i, Z) = \sum_{j=1}^{J} a_j \ln\left(\frac{z_j}{\min(z_j; x_{i,j})}\right)$$

[3] Multiplicative extended FGT index

$$p(X_i, Z) = \prod_{j=1}^{J} \left(\frac{z_j - x_{i,j}}{z_j}\right)_+^{\alpha_j}$$

[4] Tsui (2002) index

$$p(X_i, Z) = \prod_{j=1}^{J} \left(\frac{z_j}{\min(z_j; x_{i,j})}\right)^{b_j} - 1$$

[5] Intersection headcount index

$$p(X_i, Z) = \prod_{j=1}^{J} I\left(z_j > x_{i,j}\right)$$

[6] Union headcount index

$$p(X_i, Z) = 1 - \prod_{j=1}^{J} I\left(z_j < x_{i,j}\right)$$

[7] Bourguignon and Chakravarty bi-dimensional (2003) index

$$p(X_i, Z) = \left[C_1 + \beta^{\gamma/\alpha} C_2\right]^{\alpha/\gamma}$$

where:

$$C_1 = \left(\frac{z_1 - x_{i,1}}{z_1}\right)_+^{\gamma} \quad \text{and} \quad C_2 = \left(\frac{z_2 - x_{i,2}}{z_2}\right)_+^{\gamma}$$

imdpov estimates the above multidimensional poverty indices as well as their standard errors.

- The user can select among the seven multidimensional poverty indices.

- The number of dimensions can be selected (1 to 6).

- If applicable, the user can choose parameter values relevant to a chosen index.

- A group variable can be used to estimate the selected index at the level of a categorical group.

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 3 decimals; this can be also changed.

Users are encouraged to consider the exercises that appear in Section **19.3**

11 *DASP* and inequality indices

11.1 Gini and concentration indices (igini)

The Gini index is estimated as

$$\hat{I} = 1 - \frac{\hat{\xi}}{\hat{\mu}}$$

where

$$\hat{\xi} = \sum_{i=1}^{n} \left[\frac{(V_i)^2 - (V_{i+1})^2}{[V_1]^2} \right] y_i \quad \text{and} \quad V_i = \sum_{h=i}^{n} w_h \quad \text{and} \quad y_1 \geq y_2 \geq \cdots y_{n-1} \geq y_n \ .$$

The concentration index for the variable *T* when the ranking variable is *Y* is estimated as

$$\widehat{IC}_T = 1 - \frac{\hat{\xi}_T}{\hat{\mu}_T}$$

where $\hat{\mu}_T$ is the average of variable *T*,

$$\hat{\xi}_T = \sum_{i=1}^{n} \left[\frac{(V_i)^2 - (V_{i+1})^2}{[V_1]^2} \right] t_i$$

and where $V_i = \sum_{h=i}^{n} w_h$ and $y_1 \geq y_2 \geq \cdots y_{n-1} \geq y_n$.

- The user can select more than one variable of interest simultaneously. For example, one can estimate inequality, for instance by using simultaneously *per capita* consumption and *per capita* income.

- To estimate a concentration index, the user must select a ranking variable..

- A group variable can be used to estimate inequality at the level of a categorical group. If a group variable is selected, only the first variable of interest is then used.

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 6 decimals; this can be changed.

Interested users are encouraged to consider the exercises that appear in Section **19.9**

11.2 Difference between Gini/concentration indices (digini)

This module estimates differences between the Gini/concentration indices of two distributions.

For each of the two distributions:

- One variable of interest should be selected;

- To estimate a concentration index, a ranking variable must be selected;

- Conditions can be specified to focus on specific population subgroups;

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 6 decimals; this can be changed.

11.3 Generalised entropy index (ientropy)

The generalized entropy index is estimated as

$$
\hat{I}(\theta) = \begin{cases}
\dfrac{1}{\theta(\theta-1)\sum_{i=1}^{n} w_i} \sum_i w_i \left[\left(\dfrac{y_i}{\hat{\mu}} \right)^{\theta} - 1 \right] & \text{if } \theta \neq 0,1 \\[3ex]
\dfrac{1}{\sum_{i=1}^{n} w_i} \sum_i w_i \log\left(\dfrac{\hat{\mu}}{y_i} \right) & \text{if } \theta = 0 \\[3ex]
\dfrac{1}{\sum_{i=1}^{n} w_i} \sum_i \dfrac{w_i y_i}{\hat{\mu}} \log\left(\dfrac{y_i}{\hat{\mu}} \right) & \text{if } \theta = 1
\end{cases}
$$

- The user can select more than one variable of interest simultaneously. For example, one can estimate inequality simultaneously for *per capita* consumption and for *per capita* income.

- A group variable can be used to estimate inequality at the level of a categorical group. If a group variable is selected, only the first variable of interest is then used.

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 6 decimals; this can be changed.

11.4 Difference between generalized entropy indices (diengtropy)

This module estimates differences between the generalized entropy indices of two distributions.

For each of the two distributions:

- One variable of interest should be selected;

- Conditions can be specified to focus on specific population subgroups;

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 6 decimals; this can be changed.

11.5 Quantile/share ratio indices of inequality (inineq)

The quantile ratio is estimated as

$$\widehat{QR}(p_1, p_2) = \frac{\hat{Q}(p_1)}{\hat{Q}(p_2)}$$

where $Q(p)$ denotes a p-quantile and p_1 and p_2 are percentiles.

The share ratio is estimated as

$$\widehat{SR}(p1, p2, p3, p4) = \frac{\widehat{GL}(p2) - \widehat{GL}(p1)}{\widehat{GL}(p4) - \widehat{GL}(p3)}$$

where $GL(p)$ is the Generalised Lorenz curve and p_1, p_2, p_3 and p_4 are percentiles.

- The user can select more than one variable of interest simultaneously. For example, one can estimate inequality simultaneously for *per capita* consumption and for *per capita* income.

- A group variable can be used to estimate inequality at the level of a categorical group. If a group variable is selected, only the first variable of interest is then used.

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 6 decimals; this can be changed.

11.6 *Difference between Quantile/Share indices (dinineq)*

This module estimates differences between the Quantile/Share indices of two distributions.

For each of the two distributions:

- One variable of interest should be selected;

- Conditions can be specified to focus on specific population subgroups;

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed;

- The results are displayed with 6 decimals; this can be changed.

12 *DASP* and decompositions.

12.1 *FGT Poverty: decomposition by population subgroups (dfgtg)*

The **dgfgt** module decomposes the FGT poverty index by population subgroups. This decomposition takes the form

$$\widehat{P}(z;\alpha) = \sum_{g=1}^{G} \widehat{\phi}(g)\widehat{P}(z;\alpha;g)$$

where G is the number of population subgroups. The results show:

- The estimated FGT index of subgroup g : $\widehat{P}(z;\alpha;g)$

- The estimated population share of subgroup g : $\widehat{\phi}(g)$

- The estimated absolute contribution of subgroup g to total poverty: $\widehat{\phi}(g)\widehat{P}(z;\alpha;g)$

- The estimated relative contribution of subgroup g to total poverty: $\left(\widehat{\phi}(g)\widehat{P}(z;\alpha;g)\right)/\widehat{P}(z;\alpha)$

An asymptotic standard error is provided for each of these statistics.

To open the dialog box for module **dfgtg**, type *db dfgtg* in the **command window**.

Figure 7: Decomposition of the FGT index by groups

Interested users are encouraged to consider the exercises that appear in Section **19.7**

12.2 Inequality: decomposition by income sources (diginis)

The **diginis** module decomposes the Gini index by income sources. The three available approaches are:
- Rao's approach (1969)
- Lerman and Yitzhaki's approach (1985)
- Araar's approach (2006)

Reference(s)
- Lerman, R. I., and S. Yitzhaki. "Income Inequality Effects by Income Source: A New Approach and Applications to the United States." Review of Economics and Statistics 67 (1985): 151-56.
- Araar Abdelkrim (2006). On the Decomposition of the Gini Coefficient: an Exact Approach, with an Illustration Using Cameroonian Data, W.P 02-06 CIRPEE University.

12.3 Generalized entropy indices of inequality: decomposition by population subgroups (dentropyg).

The Generalised Entropy indices of inequality can be decomposed as follows:

$$\hat{I}(\theta) = \sum_{k=1}^{K} \hat{\phi}(k) \left(\frac{\hat{\mu}(k)}{\hat{\mu}} \right)^{\theta} . \hat{I}(k;\theta) + \hat{\bar{I}}(\theta)$$

where:

$\phi(k)$ is the proportion of the population found in subgroup k.

$\mu(k)$ is the mean income of group k.

$I(k;\theta)$ is inequality within group k.

$\bar{I}(\theta)$ is population inequality if each individual in subgroup k is given the mean income of subgroup k, $\mu(k)$.

12.4 Decomposition of the variation in FGT indices into growth and redistribution components (dfgtgr).

Datt and Ravallion (1992) decomposes the change in the FGT index between two periods, t1 and t2, into growth and redistribution components as follows:

$$\underbrace{P_2 - P_1}_{var\,iation} = \underbrace{\left[P(\mu^{t2},\pi^{t1}) - P(\mu^{t1},\pi^{t1}) \right]}_{C1} + \underbrace{\left[P(\mu^{t1},\pi^{t2}) - P(\mu^{t1},\pi^{t1}) \right]}_{C2} + R \quad / \,ref = 1$$

$$\underbrace{P_2 - P_1}_{var\,iation} = \underbrace{\left[P(\mu^{t2},\pi^{t2}) - P(\mu^{t1},\pi^{t2}) \right]}_{C1} + \underbrace{\left[P(\mu^{t2},\pi^{t2}) - P(\mu^{t2},\pi^{t1}) \right]}_{C2} + R \quad / \,ref = 2$$

where

variation = difference in poverty between t1 and t2;
C1   = growth component;
C2   = redistribution component;
R   = residual;
Ref   = period of reference.

$P(\mu^{t1},\pi^{t1})$: the FGT index of the first period

$P(\mu^{t1},\pi^{t1})$: the FGT index of the second period

$P(\mu^{t2},\pi^{t1})$: the FGT index of the first period when all incomes y_i^{t1} of the first period are multiplied by μ^{t2}/μ^{t1}

$P(\mu^{t1},\pi^{t2})$: the FGT index of the second period when all incomes y_i^{t2} of the second period are multiplied by μ^{t1}/μ^{t2}

The Shapley value decomposes the variation in the FGT Index between two periods, t1 and t2, into growth and redistribution components as follows:

$$\underbrace{P_2 - P_1}_{\text{Variation}} = C_1 + C_2$$

$$C_1 = \frac{1}{2}\left(\left[P(\mu^{t2}, \pi^{t1}) - P(\mu^{t1}, \pi^{t1})\right] + \left[P(\mu^{t2}, \pi^{t2}) - P(\mu^{t1}, \pi^{t2})\right]\right)$$

$$C_2 = \frac{1}{2}\left(\left[P(\mu^{t1}, \pi^{t2}) - P(\mu^{t1}, \pi^{t1})\right] + \left[P(\mu^{t2}, \pi^{t2}) - P(\mu^{t2}, \pi^{t1})\right]\right)$$

13 *DASP* and curves.

13.1 FGT CURVES (cfgt).

FGT curves are useful distributive tools that can *inter alia* be used to:
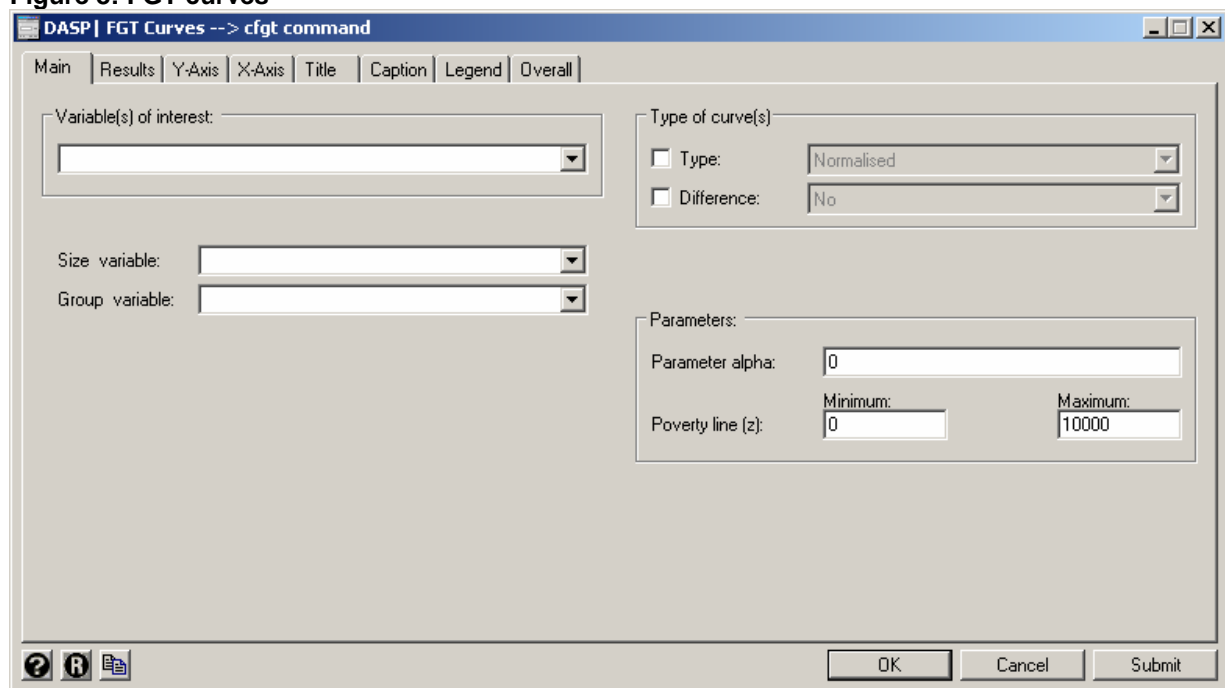
1. Show how the level of poverty varies with different poverty lines;
2. Test for poverty dominance between two distributions;
3. Test pro-poor growth conditions.

FGT curves are also called primal dominance curves. The **cfgt** module draws such curves easily. The module can:

- draw more than one FGT curve simultaneously whenever more than one variable of interest is selected;
- draw FGT curves for different population subgroups whenever a group variable is selected;
- draw FGT curves that are not normalized by the poverty lines;
- draw differences between FGT curves;
- list or save the coordinates of the curves;
- save the graphs in different formats:
    - o *.gph : STATA format;
    - o *.wmf : typically recommended to insert graphs in Word documents;
    - o *.eps : typically recommended to insert graphs in Tex/Latex documents.
- Many graphical options are available to change the appearance of the graphs.

To open the dialog box of the module **cfgt**, type the command *db dfgt* in the **command window**.

Figure 8: FGT curves

Interested users are encouraged to consider the exercises that appear in Section **19.4**.

13.2 FGT CURVE with confidence interval (cfgts).

The **cfgts** module draws an FGT curve and its confidence interval by taking into account sampling design. The module can:

- draw an FGT curve and two-sided, lower-bounded or upper-bounded confidence intervals around that curve;
- condition the estimation on a population subgroup;
- draw a FGT curve that is not normalized by the poverty lines;
- list or save the coordinates of the curve and of its confidence interval;
- save the graphs in different formats:
    - *.gph : STATA format;
    - *.wmf : typically recommended to insert graphs in Word documents;
    - *.eps : typically recommended to insert graphs in Tex/Latex documents.
- Many graphical options are available to change the appearance of the graphs.

Interested users are encouraged to consider the exercises that appear in Section **19.5**.

13.3 Difference between FGT CURVES with confidence interval (cfgts2d).

The **cfgts2d** module draws differences between FGT curves and their associated confidence interval by taking into account sampling design. The module can:

- draw differences between FGT curves and two-sided, lower-bounded or upper-bounded confidence intervals around these differences;
- normalize or not the FGT curves by the poverty lines;
- list or save the coordinates of the differences between the curves as well as the confidence intervals;
- save the graphs in different formats:
    - *.gph : STATA format;
    - *.wmf : typically recommended to insert graphs in Word documents;
    - *.eps : typically recommended to insert graphs in Tex/Latex documents.
- Many graphical options are available to change the appearance of the graphs.

Interested users are encouraged to consider the exercises that appear in Section **19.5**.

13.4 Lorenz and concentration CURVES (clorenz).

Lorenz and concentration curves are useful distributive tools that can *inter alia* be used to:

1. show the level of inequality;
2. test for inequality dominance between two distributions;
3. test for welfare dominance between two distributions;
4. test for progressivity.

The **clorenz** module draws Lorenz and concentration curves simultaneously. The module can:

- draw more than one Lorenz or concentration curve simultaneously whenever more than one variable of interest is selected;

- draw more than one generalized or absolute Lorenz or concentration curve simultaneously whenever more than one variable of interest is selected;
- draw more than one deficit share curve;
- draw Lorenz and concentration curves for different population subgroups whenever a group variable is selected;
- draw differences between Lorenz and concentration curves;
- list or save the coordinates of the curves;
- save the graphs in different formats:
  - *.gph : STATA format;
  - *.wmf : typically recommended to insert graphs in Word documents;
  - *.eps : typically recommended to insert graphs in Tex/Latex documents.
- Many graphical options are available to change the appearance of the graphs.

To open the dialog box of the module **clorenz**, type the command *db clorenz* in the **command window**.

Figure 9: Lorenz and concentration curves

Interested users are encouraged to consider the exercises that appear in Section **19.8**.

13.5 Lorenz/concentration curves with confidence intervals (clorenzs).

The **clorenzs** module draws a Lorenz/concentration curve and its confidence interval by taking sampling design into account. The module can:

- draw a Lorenz/concentration curve and two-sided, lower-bounded or upper-bounded confidence intervals;
- condition the estimation on a population subgroup;
- draw Lorenz/concentration curves and generalized Lorenz/concentration curves;

- list or save the coordinates of the curves and their confidence interval;
- save the graphs in different formats:
    - o *.gph : STATA format;
    - o *.wmf : typically recommended to insert graphs in Word documents;
    - o *.eps : typically recommended to insert graphs in Tex/Latex documents.
- Many graphical options are available to change the appearance of the graphs.

13.6 Differences between Lorenz/concentration curves with confidence interval (clorenzs2d)

The **clorenz2d** module draws differences between Lorenz/concentration curves and their associated confidence intervals by taking sampling design into account. The module can:

- draw differences between Lorenz/concentration curves and associated two-sided, lower-bounded or upper-bounded confidence intervals;
- list or save the coordinates of the differences and their confidence intervals;
- save the graphs in different formats:
    - o *.gph : STATA format;
    - o *.wmf : typically recommended to insert graphs in Word documents;
    - o *.eps : typically recommended to insert graphs in Tex/Latex documents.
- Many graphical options are available to change the appearance of the graphs.

13.7 Poverty curves (cpoverty)

The **cpoverty** module draws the poverty gap or the cumulative poverty gap curves.

- o The poverty gap at a percentile p is: $G(p;z) = (z - Q(p))_+$
- o The cumulative poverty gap at a percentile p, noted by $CPG(p;z)$, is given by:

$$CPG(p;z) = \frac{\sum\limits_{i=1}^{n} w_i \, (z - y_i)_+ I(y_i \leq Q(p))}{\sum\limits_{i=1}^{n} w_i}$$

The module can thus:

- draw more than one poverty gap or cumulative poverty gap curves simultaneously whenever more than one variable of interest is selected;
- draw poverty gap or cumulative poverty gap curves for different population subgroups whenever a group variable is selected;
- draw differences between poverty gap or cumulative poverty gap curves;
- list or save the coordinates of the curves;
- save the graphs in different formats:
    - o *.gph : STATA format;
    - o *.wmf : typically recommended to insert graphs in Word documents;
    - o *.eps : typically recommended to insert graphs in Tex/Latex documents.
- Many graphical options are available to change the appearance of the graphs.

14 Dominance

14.1 Poverty dominance (dompov)

Distribution 1 dominates distribution 2 at order s over the range $\left[z^{-}, z^{+}\right]$ if only if:

$$P_1\ (\zeta;\alpha) < P_2\ (\zeta;\alpha) \quad \forall \quad \zeta \in \left[z^{-}, z^{+}\right] \text{ for } \alpha = s - 1.$$

This involves comparing stochastic dominance curves at order *s* or FGT curves with $\alpha = s - 1$. This application estimates the points at which there is a reversal of the ranking of the curves. Said differently, it provides the crossing points of the dominance curves, that is, the values of ζ and $P_1\ (\zeta;\alpha)$ for which

$P_1\ (\zeta;\alpha) = P_2\ (\zeta;\alpha)$ when:

$sign(P_1\ (\zeta - \eta;\alpha) - P_2\ (\zeta - \eta;\alpha)) = sign(P_2\ (\zeta + \eta;\alpha) - P_1\ (\zeta + \eta;\alpha))$ for a small η. The crossing points ζ can also be referred to as "critical poverty lines".

The **dompov** module can be used to check for poverty dominance and to compute critical values. This module is mostly based on Araar (2006):

Araar, Abdelkrim, (2006), Poverty, Inequality and Stochastic Dominance, Theory and Practice: Illustration with Burkina Faso Surveys, **Working Paper: 06-34**. CIRPEE, Department of Economics, Université Laval.

Interested users are encouraged to consider the exercises that appear in Section **19.6.**

14.2 Inequality dominance (domineq)

Distribution 1 inequality-dominates distribution 2 at the second order if and only if:

$$L_1(p) \le L_2(p) \quad \forall \quad p \in [0,1]$$

The module **domineq** can be used to check for such inequality dominance. It is based mainly on Araar (2006):

Araar, Abdelkrim, (2006), Poverty, Inequality and Stochastic Dominance, Theory and Practice: Illustration with Burkina Faso Surveys, **Working Paper: 06-34**. CIRPEE, Department of Economics, Université Laval.

Intersections between curves can be estimated with this module. It can also used to check for tax and transfer progressivity by comparing Lorenz and concentration curves.

14.3 DASP and bi-dimensional poverty dominance (dombdpov)

Let two dimensions of well-being be denoted by $k = 1, 2$. The intersection bi-dimensional FGT index for distribution D is estimated as

$$\hat{P}_D(Z;A) = \frac{\sum\limits_{i=1}^{n} w_i \left[\prod\limits_{k=1}^{2} (z^k - y_i^k)_+^{\alpha_k} \right]}{\sum\limits_{i=1}^{n} w_i}$$

where $Z = (z_1, z_2)$ *and* $A = (\alpha_1, \alpha_2)$ *are* vectors of poverty lines and parameters α respectively, and $x_+ = \max(x, 0)$.

Distribution 1 dominates distribution 2 at orders (s_1, s_2) over the range $\left[0, Z^+ \right]$ if and only if:

$$P_1(Z; A = s - 1) < P_2(Z; A = s - 1) \quad \forall \quad Z \in \left[0, z_1^+ \right] \times \left[0, z_2^+ \right] \text{ and for } \alpha_1 = s_1 - 1, \quad \alpha_2 = s_2 - 1.$$

The DASP **dombdpov** module can be used to check for such dominance.

For each of the two distributions:

- The two variables of interest (dimensions) should be selected;
- Conditions can be specified to focus on specific population subgroups;
- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.
- Surfaces showing the difference, the lower bound and the upper bound of the confidence surfaces are plotted interactively with the GnuPlot tool.
- Coordinates can be listed.
- Coordinates can be saved in Stata or GnuPlot-ASCII format.

Interested users are encouraged to consider the exercises that appear in Section **19.12.**

15 Distributive tools

15.1 Quantile curves (c_quantile)

The quantile at a percentile *p* of a continuous population is given by:

$Q(p) = F^{-1}(p)$ where $p = F(y)$ is the cumulative distribution function at *y*.

For a discrete distribution, let *n* observations of living standards be ordered such that $y_1 \leq y_2 \leq \cdots \leq y_i \leq y_{i+1} \leq \cdots \leq y_n$. If $F(y_i) < p \leq F(y_{i+1})$, we define $Q(p) = y_{i+1}$. The normalised quantile is defined as $\overline{Q}(p) = Q(p)/\mu$.

Interested users are encouraged to consider the exercises that appear in Section **19.10.**

15.2 Density curves (cdensity)

The Gaussian kernel estimator of a density function $f(x)$ is defined by

$$\hat{f}(x) = \frac{\sum_i w_i K_i(x)}{\sum_{i=1}^{n} w_i} \quad and \quad K_i(x) = \frac{1}{h\sqrt{2\pi}} \exp\left(-0.5 \, \lambda_i(x)^2\right) \quad and \quad \lambda_i(x) = \frac{x - x_i}{h}$$

where h is a bandwidth that acts as a "smoothing" parameter.

Interested users are encouraged to consider the exercises that appear in Section **19.10.**

15.3 Non-parametric regression curves (cnpe)

Non-parametric regression is useful to show the link between two variables without specifying beforehand a functional form. It can also be used to estimate the local derivative of the first variable with respect to the second without having to specify the functional form linking them. Regressions with the **cnpe** module can be performed with one of the following two approaches:

15.3.1 Nadaraya-Watson approach

A Gaussian kernel regression of y on x is given by:

$$E\left(y|x\right) = \Phi(y \mid x) = \frac{\sum_i w_i K_i(x) y_i}{\sum_i w_i K_i(x)}$$

From this, the derivative of $\Phi(y \mid x)$ with respect to x is given by

$$E\left(\frac{dy}{dx}|x\right) = \frac{\partial \Phi(y \mid x)}{\partial x}$$

15.3.2 Local linear approach

The local linear approach is based on a local OLS estimation of the following functional form:

$$K_i(x)^{\frac{1}{2}} y_i = \mu(x) K_i(x)^{\frac{1}{2}} + \mu'(x) K_i(x)^{\frac{1}{2}} (x_i - x) + v$$

or, alternatively, of:

$$K_i(x)^{\frac{1}{2}} y_i = \alpha K_i(x)^{\frac{1}{2}} + \beta K_i(x)^{\frac{1}{2}} (x_i - x) + v_i$$

Estimates are then given by:

$$E\left(y|x\right) = \alpha, \; E\left(\frac{dy}{dx}|x\right) = \beta$$

Interested users are encouraged to consider the exercises that appear in Section **19.10.**

15.4 DASP and joint density functions.

The module **sjdensity** can be used to draw a joint density surface. The Gaussian kernel estimator of the joint density function $f(x, y)$ is defined as:

$$\hat{f}(\overline{x}, \overline{y}) = \frac{1}{2\pi h_x h_y \sum\limits_{i=1}^{n} w_i} \sum\limits_{i=1}^{n} w_i \ \exp\left(-\left(\frac{1}{2}\right)\left(\left(\frac{\overline{x} - x_i}{h_x}\right)^2 + \left(\frac{\overline{y} - y_i}{h_y}\right)^2\right)\right)$$

With this module:

- The two variables of interest (dimensions) should be selected;
- specific population subgroup can be selected;
- surfaces showing the joint density function are plotted interactively with the GnuPlot tool;
- coordinates can be listed;c
- coordinates can be saved in Stata or GnuPlot-ASCII format.

Interested users are encouraged to consider the exercises that appear in Section **19.11???**

15.5 DASP and joint distribution functions

The module **sjdistrub** can be used to draw joint distribution surfaces. The joint distribution function $F(\overline{x}, \overline{y})$ is defined as:

$$\hat{F}(\overline{x}, \overline{y}) = \frac{\sum\limits_{i=1}^{n} w_i \ I(x_i \leq \overline{x}) I(y_i \leq \overline{y})}{\sum\limits_{i=1}^{n} w_i}$$

With this module:

- The two variables of interest (dimensions) should be selected;
- specific population subgroups can be selected;
- surfaces showing the joint distribution function are plotted interactively with the GnuPlot tool;
- coordinates can be listed;
- coordinates can be saved in Stata or GnuPlot-ASCII format.

Interested users are encouraged to consider the exercises that appear in Section **19.11**

16 *DASP* and pro-poor growth

16.1 DASP and pro-poor indices

The module **ipropoor** estimates simultaneously the three following pro-poor indices:

1. **The Chen and Ravallion pro-poor index (2003):**

$$Index = \frac{W_1(z) - W_2(z)}{F_1(z)}$$

where $W_D(z)$ is the Watts index for distribution $D \in [1,2]$ and $F_1(z)$ is the headcount for index for the first distribution, both with poverty lines *z*.

2. **The Kakwani and Pernia pro-poor index (2000):**

$$Index = \frac{P_1(z,\alpha) - P_2(z\alpha)}{P_1(z,\alpha) - P_1(z(\mu_1/\mu_2),\alpha)}$$

3. **The Kakwani, Khandker and Son pro-poor index (2003):**

$$Index\_1 = g\frac{P_1(z,\alpha) - P_2(z\alpha)}{P_1(z,\alpha) - P_1(z(\mu_1/\mu_2),\alpha)}$$

where the average growth is $g(\mu_2 - \mu_1)/\mu_1$ and where a second index is given by:

$$Index\_2 = Index\_1 - g$$

- One variable of interest should be selected for each distribution.

- Conditions can be specified to focus on specific population subgroups.

- Standard errors and confidence intervals with a confidence level of 95% are provided. Both the type of confidence intervals provided and the level of confidence used can be changed.

- The results are displayed with 6 decimals; this can be changed.

- A level for the parameter α can be chosen for each of the two distributions.

16.2 DASP and pro-poor curves

Pro-poor curves can be drawn using either the primal or the dual approach. The former uses income levels. The latter is based on percentiles.

16.2.1 Primal pro-poor curves

The change in the distribution from state 1 to state 2 is s-order absolutely pro-poor with standard *cons* if:

$$\Delta(z,s) = \left(P_2(z+cons,\alpha = s-1) - P_1(z,\alpha = s-1)\right) < 0 \; \forall \; z \in \left[0,z^+\right]$$

The change in the distribution from state 1 to state 2 is s-order relatively pro-poor if:

$$\Delta(z,s) = z\left(P_2(z\frac{\mu_2}{\mu_1},\alpha = s-1) - P_1\left(z,\alpha = s-1\right)\right) < 0 \; \forall \; z \in \left[0,z^+\right]$$

The module **cpropoorp** can be used to draw these primal pro-poor curves and their associated confidence interval by taking into account sampling design. The module can:

- draw pro-poor curves and their two-sided, lower-bounded or upper-bounded confidence intervals;
- list or save the coordinates of the differences between the curves as well as those of the confidence intervals;
- save the graphs in different formats:
    - *.gph : STATA format;
    - *.wmf : typically recommended to insert graphs in Word documents;
    - *.eps : typically recommended to insert graphs in Tex/Latex documents.
    Many graphical options are available to change the appearance of the graphs.

Interested users are encouraged to consider the exercises that appear in Section **19.13.**

16.2.2   Dual pro-poor curves

Let: :

$Q(p)$: quantile at percentile p .
$GL(p)$: Generalised Lorenz curve at percentile p .
μ : average living standards.
The change in the distribution from state 1 to state 2 is first-order absolutely pro-poor with standard *cons=0* if:

$$\Delta(z,s) = Q_2(p) - Q_1(p) > 0 \; \forall \; p \in \left[0, \mathrm{p}^+ = F(z^+)\right]$$

or equivalently if:

$$\Delta(z,s) = \frac{Q_2(p) - Q_1(p)}{Q_1(p)} > 0 \; \forall \; p \in \left[0, \mathrm{p}^+ = F(z^+)\right]$$

The change in the distribution from state 1 to state 2 is first-order relatively pro-poor if:

$$\Delta(z,s) = \frac{Q_2(p)}{Q_1(p)} - \frac{\mu_2}{\mu_1} > 0 \; \forall \; p \in \left[0, \mathrm{p}^+ = F(z^+)\right]$$

The change in the distribution from state 1 to state 2 is second-order absolutely pro-poor if:

$$\Delta(z,s) = GL_2(p) - GL_1(p) > 0 \ \forall \ p \in \left[0, \mathrm{p}^+ = F(z^+) \right]$$

or equivalently if:

$$\Delta(z,s) = \frac{GL_2(p) - GL_1(p)}{GL_1(p)} > 0 \ \forall \ p \in \left[0, \mathrm{p}^+ = F(z^+) \right]$$

The change in the distribution from state 1 to state 2 is first-order relatively pro-poor if:

$$\Delta(z,s) = \frac{GL_2(p)}{GL_1(p)} - \frac{\mu_2}{\mu_1} > 0 \ \forall \ p \in \left[0, \mathrm{p}^+ = F(z^+) \right]$$

The module **cpropoord** can be used to draw these dual pro-poor curves and their associated confidence interval by taking into account sampling design. The module can:

- draw pro-poor curves and their two-sided, lower-bounded or upper-bounded confidence intervals;
- list or save the coordinates of the differences between the curves as well as those of the confidence intervals;
- save the graphs in different formats:
  - *.gph : STATA format;
  - *.wmf : typically recommended to insert graphs in Word documents;
  - *.eps : typically recommended to insert graphs in Tex/Latex documents.

Many graphical options are available to change the appearance of the graphs.

Interested users are encouraged to consider the exercises that appear in Section **19.13**

17 *DASP* and Benefit Incidence Analysis

17.1 Benefit incidence analysis

The main objective of using a benefit incidence approach is to analyse the distribution of benefits from the use of public services according to the distribution of living standards.

Two main sources of information are used. The first informs on the access of household members to public services. This information can be found in the usual household surveys. The second deals with the amount of total public expenditures on each public service. This information is usually available at the national level and sometimes in a more disaggregated format, such as at the regional level. The benefit incidence approach combines the use of these two sources of information to analyse the distribution of public benefits and its progressivity.

Formally, let

w_i be the sampling weight of observation i;

y_i be the living standard of members belonging to observation i (i.e., *per capita* income);

e_i^s be the number of "eligible" members of observation i, i.e., members that "need" the public service provided by sector s. There are S sectors;

f_i^s be the number of members of observation i that effectively use the public service provided by sector s;

g_i be the socio-economic group of eligible members of observation i (typically classified by income percentiles);

c_i be a subgroup indicator for observation i (e.g., 1 for a rural resident, and 2 for an urban resident). Eligible members can thus be grouped into population exclusive subgroups;

E_r^s be total public expenditures on sector s in area r. There are R areas (the area here refers to the geographical division which one can have reliable information on total public expenditures on the studied public service);

E^s be total public expenditures on sector s $\left(E^s = \sum\limits_{r=1}^{R} E_r^s \right)$.

Here are some of the statistics that can be computed.

1. The share of a g in sector s is defined as follows:

$$SH_g^s = \frac{\sum\limits_{i=1}^{n} w_i f_i^s I(i \in g)}{\sum\limits_{i=1}^{n} w_i f_i^s}$$

Note that: $\sum\limits_{g=1}^{G} SH_g^s = 1$.

2. The rate of participation of a group g in sector s is defined as follows:

$$CR_g^s = \frac{\sum\limits_{i=1}^{n} w_i f_i^s I(i \in g)}{\sum\limits_{i=1}^{n} w_i e_i^s I(i \in g)}$$

This rate cannot exceed 100% since $f_i^s \le e_i^s \ \forall i$.

3. The unit cost of a benefit in sector s for observation j, which refers to the household members that live in area r :

$$UC_j^s = \frac{E_r^s}{\sum\limits_{j=1}^{n_r} w_j f_j^s}$$

where n_r is the number of sampled households in area r.

4. The benefit of observation i from the use of public sector s is:

$$B_i^s = f_i^s UC_i^s$$

5. The benefit of observation i from the use of the S public sectors is:

$$B_i = \sum_{s=1}^{S} B_i^s$$

6. The average benefit at the level of those eligible to a service from sector s and for those observations that belong to a group g, is defined as:

$$ABE_g^s = \frac{\sum_{i=1}^{n} w_i B_i^s I(i \in g)}{\sum_{i=1}^{n} w_i e_i^s I(i \in g)}$$

7. The average benefit for those that use the service s and belong to a group g is defined as:

$$ABF_g^s = \frac{\sum_{i=1}^{n} w_i B_i^s I(i \in g)}{\sum_{i=1}^{n} w_i f_i^s I(i \in g)}$$

8. The proportion of benefits from the service from sector s that accrues to observations that belong to a group g is defined as:

$$PB_g^s = \frac{B_g^s}{E^s}$$

where $B_g^s = \sum_{i=1}^{n} w_i B_i^s I(i \in g)$.

These statistics can be restricted to specific socio-demographic groups (e.g.,. rural/urban) by replacing $I(i \in g)$ by $I(i \in c)$.

.
The **bian.ado** module allows the computation of these different statistics.

Some characteristics of the module:

- o Possibility of selecting between one and six sectors.
- o Possibility of using frequency data approach when information about the level of total public expenditures is not available.

- Generation of benefit variables by the type of public services (ex: primary, secondary and tertiary education levels) and by sector.
- Generation of unit cost variables for each sector.
- Possibility of computing statistics according to groups of observations.
- Generation of statistics according to social-demographic groups, such as quartiles, quintiles or deciles.

Generally, public expenditures on a given service can vary from one geographical or administrative area to another. When the information about public expenditures is available at the level of areas, this information can be used with the *bian* module to estimate unit cost more accurately.

Example 1

Observation i	HH size	Eligible HH members	Frequency	Area indicator	Total level of regional public expenditures
1	7	3	2	1	14000
2	4	2	2	1	14000
3	5	5	3	1	14000
4	6	3	2	2	12000
5	4	2	1	2	12000

In this example, the first observation contains information on household 1.
- This household contains 7 individuals;
- Three individuals in this household are eligible to the public service;
- Only 2 among the 3 eligible individuals benefit from the public service;
- This household lives in area 1. In this area, the government spends a total of 14000 to provide the public service for the 7 users of this area (2+2+3).

The unit cost in area 1 equals: 14000/7=2000
The unit cost in area 2 equals: 12000/3=4000

By default, the area indicator is set to 1 for all households. When this default is used, the variable Regional public expenditures (the fifth column that appears in the dialog box) should be set to total public expenditures at the national level. This would occur when the information on public expenditures is only available at the national level.

Example 2

Observation i	HH size	Eligible members	Frequency	Area indicator	Regional public expenditures
1	7	3	2	1	28000
2	4	2	2	1	28000
3	5	5	3	1	28000
4	6	3	2	1	28000
5	4	2	1	1	28000

The unit cost benefit (at the national level) equals: 28000/10=2800

Interested users are encouraged to consider the exercises that appear in Section **19.14**

18 Appendices

18.1 Appendix A: illustrative household surveys

18.1.1   The 1994 Burkina Faso survey of household expenditures (bkf94I.dta)

This is a nationally representative survey, with sample selection using two-stage stratified random sampling. Seven strata were formed. Five of these strata were rural and two were urban. Primary sampling units were sampled from a list drawn from the 1985 census. The last sampling units were households.

List of variables
strata   **Stratum in which a household lives**

psu   **Primary sampling unit**

weight   **Sampling weight**

size   **Household size**

exp   **Total household expenditures**

expeq   **Total household expenditures per adult equivalent**

expcp   **Total household expenditures per capita**

gse   **Socio-economic group of the household head**
    1 wage-earner (public sector)
    2 wage-earner (private sector)
    3 Artisan or trader
    4 Other type of earner
    5 Crop farmer
    6 Subsistence farmer
    7 Inactive

sex   **Sex of household head**
    1 Male
    2 Female
zone   **Residential area**
    1 Rural
    2 Urban

18.1.2   The 1998 Burkina Faso survey of household expenditures (bkf98l.dta)

This survey is similar to the 1994 one, although ten strata were used instead of seven for 1994. To express 1998 data in 1994 prices, two alternative procedures have been used. First, 1998 expenditure data were multiplied by the ratio of the 1994 official poverty line to the 1998 official poverty line: z_1994/z_1998. Second, 1998 expenditure data were multiplied by the ratio of the 1994 consumer price index to the 1998 consumer price index: ipc_1994/ipc_1998.

List of new variables

expcpz   Total household expenditures per capita deflated by (z_1994/z_1998)

expcpi   Total expenditures per capita deflated by (ipc_1994/ipc_1998)

18.1.3   Canadian Survey of Consumer Finance (a sub sample of 1000 observations – can6.dta)

List of variables

X   Yearly gross income per adult equivalent.

T   Income taxes per adult equivalent.

B1   Transfer 1 per adult equivalent.

B2   Transfer 2 per adult equivalent.

B3   Transfer 3 per adult equivalent.

B   Sum of transfers *B1*, *B2* and *B3*

N   Yearly net income per adult equivalent (*X* minus *T* plus *B*)

18.1.4   Peru LSMS survey 1994 (A sample of 3623 household observations - PEREDE94l.dta)

List of variables

exppc   Total expenditures, per capita (constant June 1994 soles per year).

weight   Sampling weight

size   Household size

npubprim   Number of household members in public primary school

npubsec   Number of household members in public secondary school

npubuniv   Number of household members in public post-secondary school

18.1.5 Peru LSMS survey 1994 (A sample of 3623 household observations – PERU_A_I.dta)

List of variables

hhid   Household Id.

exppc   Total expenditures, per capita (constant June 1994 soles per year).

size   Household size

literate   Number of literate household members

pliterate   literate/size

18.1.6 The 1995 Colombia DHS survey (columbial.dta)

This sample is a part of the Data from the Demographic and Health Surveys (Colombia_1995) witch contains the following information for children aged 0-59 months

List of variables

hid   Household id
haz   height-for-age
waz   weight-for-age
whz   weight-for-height
sprob   survival probability
wght   sampling weight
Asset   asset index

18.1.7 The 1996 Dominican Republic DHS survey (Dominican_republic1996l.dta)

This sample is a part of the Data from the Demographic and Health Surveys (Republic Dominican_1996) witch contains the following information for children aged 0-59 months

List of variables

hid   Household id
haz   height-for-age
waz   weight-for-age

whz	weight-for-height
sprob	survival probability
wght	sampling weight
Asset	asset index

18.2 Appendix B: labelling variables and values

- The following .do file can be used to set labels for the variables in ***bkf94.dta.***
- For more details on the use of *label* command, type *help label* in the **command window**.

```
===============================lab_bkf94.do ===============================

# delim ;

/*   To drop all label values   */

label drop _all;

/* To assign labels */
label var strata    "Stratum in which a household lives";

label var psu       "Primary sampling unit";
label var weight    "Sampling weight";
label var size      "Household size";
label var totexp    "Total household expenditures";
label var exppc     "Total household expenditures per capita";
label var expeq     "Total household expenditures per adult equivalent";
label var gse       "Socio-economic group of the household head";

/* To define the label values that will be assigned to the categorical
variable gse */

label define lvgse

1 "wage-earner (public sector)"
2 "wage-earner (private sector)"
3 "Artisan or trader"
4 "Other type of earner"
5 "Crop farmer"
6 "Subsistence farmer"
7 "Inactive"
;

/*To assign the label values "lvgse" to the variable gse */
label val gse lvgse;

label var sex       "Sex of household head";
label def lvsex
1 Male
2 Female
;

label val sex lvsex;

label var zone      "Residential area";
```

```
label def lvzone
1 Rural
2 Urban
;

label val zone lvzone;

===================================End===================================
```

18.3 Appendix C: setting the sampling design

To set the sampling design for the data file **bkf94.dta**, open the dialog box for the command *svyset* by typing the syntax *db svyset* in the **command window**. In the **Main** panel, set STRATA and SAMPLING UNITS as follows:

Figure 10: Survey data settings
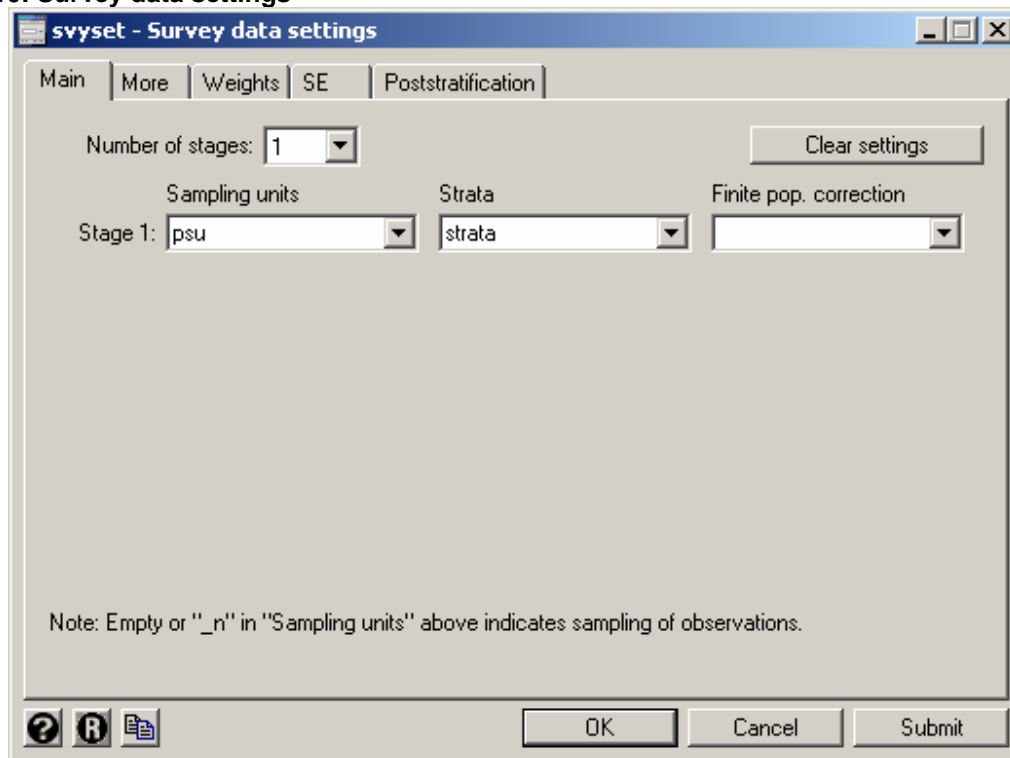
In the **Weights** panel, set SAMPLING WEIGHT VARIABLE as follows:

Figure 11: Setting sampling weights

Click on OK and save the data file.

To check if the sampling design has been well set, type the command *svydes*. The following will be displayed:

Survey: Describing stage 1 sampling units

```
    pweight: weight
        VCE: linearized
   Strata 1: strata
       SU 1: psu
      FPC 1: <zero>
```

			#Obs per Unit		
Stratum	#Units	#Obs	min	mean	max
1	42	838	19	20.0	20
2	37	733	17	19.8	20
3	98	1959	19	20.0	20
4	55	1093	19	19.9	20
5	66	1286	13	19.5	20
6	41	779	1	19.0	20
7	97	1937	19	20.0	20
7	436	8625	1	19.8	20

19 Examples and exercises

19.1 Estimation of FGT poverty indices

"How poor was Burkina Faso in 1994?"

1. Open the **bkf94.dta** file and label variables and values using the information of Section 18.1.1. Type the *describe* command and then *label list* to list labels.
2. Use the information of Section 18.1.1. to set the sampling design and then save the file.
3. Estimate the headcount index using variables of interest *expcc* and *expeq*.
   a. You should set SIZE to household size in order to estimate poverty over the population of individuals.
   b. Use the so-called 1994 official poverty line of 41099 Francs CFA per year.
4. Estimate the headcount index using the same procedure as above except that the poverty line is now set to 60% of the median.
5. Using the official poverty line, how does the headcount index for male- and female-headed households compare?
6. Can you draw a 99% confidence interval around the previous comparison? Also, set the number of decimals to 4.

Answer
Q.1
If **bkf94.dta** is saved in the directory **c:/data**, type the following command to open it:
use "C:\data\bkf94.dta", clear

If **lab_bkf94.do** is saved in the directory **c:/do_files**, type the following command to label variables and labels:

do "C:\do_files\lab_bkf94.do"

Typing the command *describe*, we obtain:

```
obs:      8,625
vars:     9                    31 Oct 2006 13:48
size:     285,087 (99.6% of      memory free)
storage  display  value
variable name    type     format        label          variable label

weight           float    %9.0g                         Sampling weight
size             byte     %8.0g                         Household size
strata           byte     %8.0g                         Stratum in which a household lives
psu              byte     %8.0g                         Primary sampling unit
gse              byte     %29.0g        gse             Socio-economic group of the household head
sex              byte     %8.0g         sex             Sex of household head
zone             byte     %8.0g         zone            Residential area
exp              double   %10.0g                        Total household expenditures
expeq            double   %10.0g                        Total household expenditures per adult equivalent
exppc            float    %9.0g                         Total household expenditures per capita
```

Typing *label list*, we find:
```
zone:
        1       Rural
        2       Urban
sex:
        1       Male
        2       Female
gse:
```

1	wage-earner (public sector)
2	wage-earner (private sector)
3	Artisan or trader
4	Other type of earner
5	Crop farmer
6	Food farmer
7	Inactive

Q.2

You can set the sampling design with a dialog box, as indicated in Section 18.3, or simply by typing

svyset psu [pweight=weight], strata(strata) vce(linearized)

Typing *svydes*, we obtain

```
Survey: Describing stage 1 sampling units

    pweight: weight
        VCE: linearized
   Strata 1: strata
      SU 1: psu
     FPC 1: <zero>

                                    #Obs per Unit

Stratum     #Units      #Obs     min       mean      max

      1         42       838      19      20.0        20
      2         37       733      17      19.8        20
      3         98      1959      19      20.0        20
      4         55      1093      19      19.9        20
      5         66      1286      13      19.5        20
      6         41       779       1      19.0        20
      7         97      1937      19      20.0        20

      7        436      8625       1      19.8        20
```

Q.3

Type *bd ifgt* to open the dialog box for the FGT poverty index and choose variables and parameters as indicated in the following window. Click on SUBMIT.

Figure 12: Estimating FGT indices

The following results should then be displayed:

Q.4

Select RELATIVE for the poverty line and set the other parameters as above.

Figure 13: Estimating FGT indices with relative poverty lines

After clicking on SUBMIT, the following results should be displayed:

```
. ifgt exppc, alpha(0) hsize(size) opl(median) prop(60)

    Poverty Index   :  FGT Index
    Household size  :  size
    Sampling weight :  weight
    Parameter alpha :  0.00
```

Variable	Estimate	STD	LB	UB	P. Line
exppc	**0.185243**	**0.008576**	**0.168386**	**0.202099**	27046.71

Q.5

Set the group variable to *sex*.

Figure 14: FGT indices differentiated by gender

Clicking on SUBMIT, the following should appear:

Q.6

Using the panel CONFIDENCE INTERVAL, set the confidence level to 99 % and set the number of decimals to 4 in the RESULTS panel.

```
. ifgt exppc, alpha(0) hsize(size) hgroup(sex) dec(4) level(99) pline(41099)
```

```
Poverty Index    : FGT Index
Household size   : size
Sampling weight  : weight
Group variable   : sex
Parameter alpha  : 0.00
```

Group	Estimate	STD	LB	UB	P. Line
1: Male	**0.4522**	**0.0166**	**0.4091**	**0.4952**	41099.00
2: Female	**0.2819**	**0.0282**	**0.2089**	**0.3548**	41099.00
POPULATION	0.4446	0.0161	0.4028	0.4863	41099.00

19.2 Estimating differences between FGT indices.

"Has poverty Burkina Faso decreased between 1994 and 1998?"

1. Open the dialog box for the difference between FGT indices.
2. Estimate the difference between headcount indices when
    a. Distribution 1 is year 1998 and distribution 2 is year 1994;
    b. The variable of interest is **exppc** for 1994 and **exppcz** for 1998.
    c. You should set size to household size in order to estimate poverty over the population of individuals.
    d. Use 41099 Francs CFA per year as the poverty line for both distributions.
3. Estimate the difference between headcount indices when
    a. Distribution 1 is rural residents in year 1998 and distribution 2 is rural residents in year 1994;
    b. The variable of interest is **exppc** for 1994 and **exppcz** for 1998.
    c. You should set size to household size in order to estimate poverty over the population of individuals.
    d. Use 41099 Francs CFA per year as the poverty line for both distributions.
4. Redo the last exercise for urban residents.
5. Redo the last exercise only for members of male-headed households.
6. Test if the estimated difference in the last exercise is significantly different from zero. Thus, test:

$$H_0 : \Delta P(z = 41099, \alpha = 0) = 0 \quad against \quad H_1 : \Delta P(z = 41099, \alpha = 0) \neq 0$$

Set the significance level to 5% and assume that the test statistics follows a normal distribution.

Answers

Q.1

Open the dialog box by typing

db difgt

Q.2
- For distribution 1, choose the option DATA IN FILE instead of DATA IN MEMORY and click on BROWSE to specify the location of the file ***bkf98I.dta***.
- Follow the same procedure for distribution 2 to specify the location of ***bkf94I.dta***.
- Choose variables and parameters as follows:

Figure 15: Estimating differences between FGT indices

After clicking on SUBMIT, the following should be displayed:

```
. difgt exppcz exppc, alpha(0) file1(C:\DATA\bkf98I.dta) hsize1(size) file2(C:\DATA\bkf94I.dta) hsize2(size) pline1(41099) pline2(41099)
```

```
Poverty Index  :  FGT Index
Parameter alpha :     0.00
```

	Estimate	STD	LB	UB	P. Line
Distribution_1	0.452677	0.010927	0.431199	0.474156	41099.00
Distribution_2	0.444565	0.016124	0.412873	0.476256	41099.00
Difference	0.008113	0.019477	-0.030062	0.046288	---

Q.3

- ■ Restrict the estimation to rural residents as follows:
  - o Select the option Condition(s)
  - o Write ZONE in the field next to CONDITION (1) and type 1 in the next field.

Figure 16: Estimating differences in FGT indices

After clicking on SUBMIT, we should see:

```
Poverty Index   :   FGT Index
Parameter alpha :       0.00
```

	Estimate	STD	LB	UB	P. Line
Distribution_1	0.510344	0.011601	0.487539	0.533149	41099.00
Distribution_2	0.510497	0.019975	0.471236	0.549758	41099.00
Difference	-0.000153	0.023100	-0.045427	0.045121	---

Q.4

```
Poverty Index   :   FGT Index
Parameter alpha :       0.00
```

	Estimate	STD	LB	UB	P. Line
Distribution_1	0.164573	0.016297	0.132538	0.196608	41099.00
Distribution_2	0.103684	0.013419	0.077309	0.130059	41099.00
Difference	0.060889	0.021111	0.019513	0.102265	---

50

One can see that the change in poverty was significant only for urban residents. **Q.5**

Restrict the estimation to male-headed urban residents as follows:
- o Set the number of Condition(s) to 2;
- o Set **sex** in the field next to **Condition (2)** and type **1** in the next field.

Figure 17: FGT differences across years by gender and zone

After clicking on SUBMIT, the following should be displayed:

```
Poverty Index   :   FGT Index
Parameter alpha :      0.00
```

	Estimate	STD	LB	UB	P. Line
Distribution_1	0.172384	0.017701	0.137590	0.207179	41099.00
Distribution_2	0.105997	0.013945	0.078588	0.133405	41099.00
Difference	0.066388	0.022534	0.022222	0.110553	---

Q.6

We have that:

Lower Bound: = 0.0222
Upper Bound: = 0.1105

The null hypothesis is rejected since the lower bound of the 95% confidence interval is above zero.

19.3 Estimating multidimensional poverty indices

"How much is bi-dimensional poverty (total expenditures and literacy) in Peru in 1994?"

Using the **peru94l.dta** file,

1. Estimate the *Chakravarty et al (1998)* index with parameter alpha = 1 and

	Var. of interest	Pov. line	a_j
Dimension 1	exppc	400	1
Dimension 2	pliterate	0.90	1

2. Estimate the *Bourguignon and Chakravarty (2003)* index with parameters alpha=beta=gamma= 1 and

	Var. of interest	Pov. line
Dimension 1	exppc	400
Dimension 2	literate	0.90

Q.1

Steps:
- Type
  use "C:\data\peru94l.dta", clear

- To open the relevant dialog box, type
  db imdpov

- Choose variables and parameters as in

Figure 18: Estimating multidimensional poverty indices (A)

After clicking SUBMIT, the following results appear.


```
imdpov exppc pliterate, hsize(size) index(1) alpha(0) a1(1) pl1(400) a2(1) pl2(0.9)

    M.D. Poverty index :  Chakravarty et al (1998)
    Household size     :  size
```

	Estimate	STD	LB	UB
Population	0.418	0.009	0.403	0.433

Q.2

Steps:

- Choose variables and parameters as in

Figure 19: Estimating multidimensional poverty indices (B)

After clicking SUBMIT, the following results appear.

19.4 Estimating FGT curves.

"How sensitive to the choice of a poverty line is the rural-urban difference in poverty?"

1. Open **bkf94I.dta**
2. Open the FGT curves dialog box.
3. Draw FGT curves for variables of interest *exppc* and *expeq* with
    a. parameter $\alpha = 0$;
    b. poverty line between 0 and 100,000 Franc CFA;
    c. size variable set to *size*;
    d. subtitle of the figure set to "Burkina 1994".
4. Draw FGT curves for urban and rural residents with
    a. variable of interest set to *expcap*;
    b. parameter $\alpha = 0$;
    c. poverty line between 0 and 100,000 Franc CFA;
    d. size variable set to *size*.
5. Draw the difference between these two curves and
    a. save the graph in *.gph format to be plotted in Stata and in *.wmf format to be inserted in a Word document.
    b. List the coordinates of the graph.
6. Redo the last graph with $\alpha = 1$.

Answers

Q.1

Open the file with

use "C:\data\bkf94I.dta", clear

Q.2

Open the dialog box by typing

db difgt

Q.3

Choose variables and parameters as follows:

Figure 20: Drawing FGT curves

To change the subtitle, select the **Title** panel and write the subtitle.

Figure 21: Editing FGT curves

After clicking SUBMIT, the following graph appears:

Figure 22: Graph of FGT curves
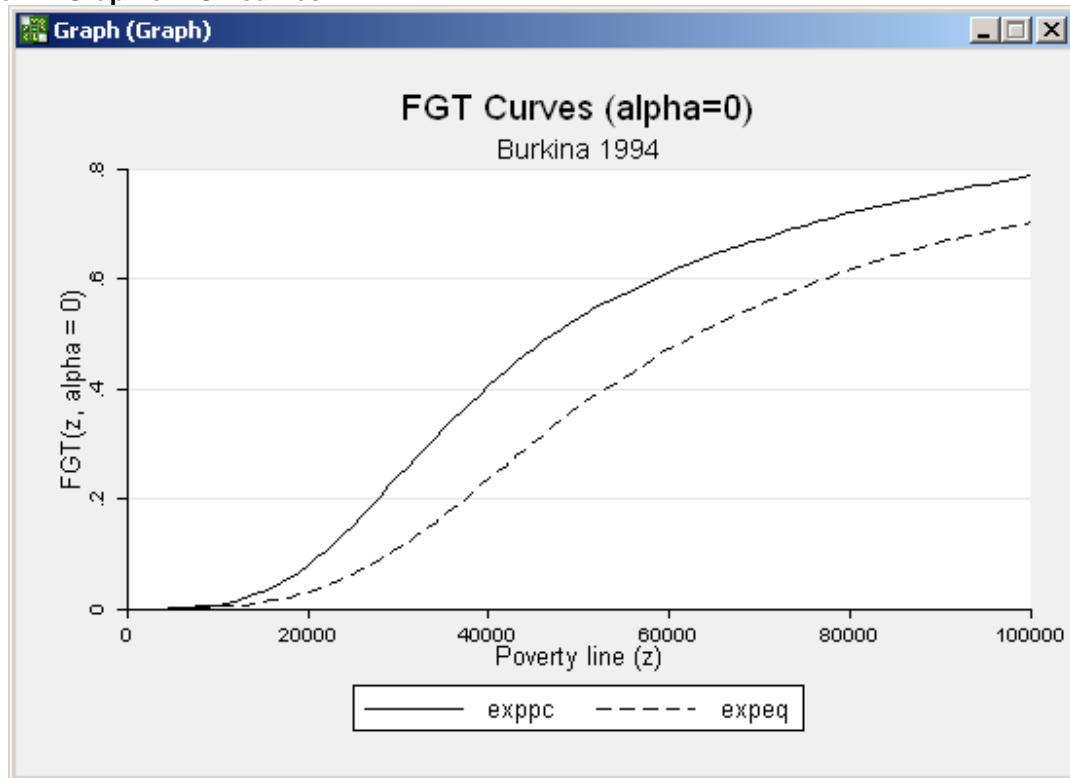
FGT Curves (alpha=0)
Burkina 1994

Choose variables and parameters as in the following window:

Figure 23: FGT curves by zone

After clicking SUBMIT, the following graph appears:

Figure 24: Graph of FGT curves by zone
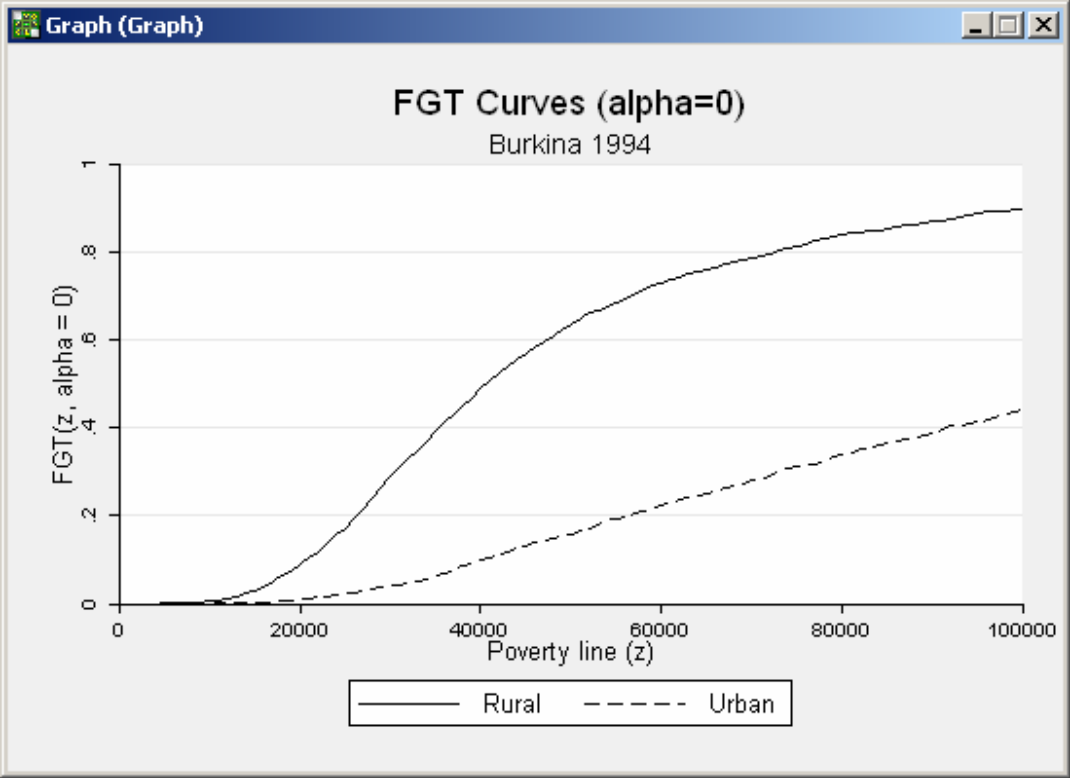
- Choose the option DIFFERENCE and select: WITH THE FIRST CURVE;
- Indicate that the group variable is *zone*;
- Select the **Results** panel and choose the option LIST in the COORDINATES quadrant.
- In the GRAPH quadrant, select the directory in which to save the graph in gph format and to export the graph in wmf format.

Figure 25: Differences of FGT curves

Figure 26: Listing coordinates

After clicking SUBMIT, the following appears:

Figure 27: Differences between FGT curves

Difference Between FGT Curves (alpha=0)
Burkina 1994
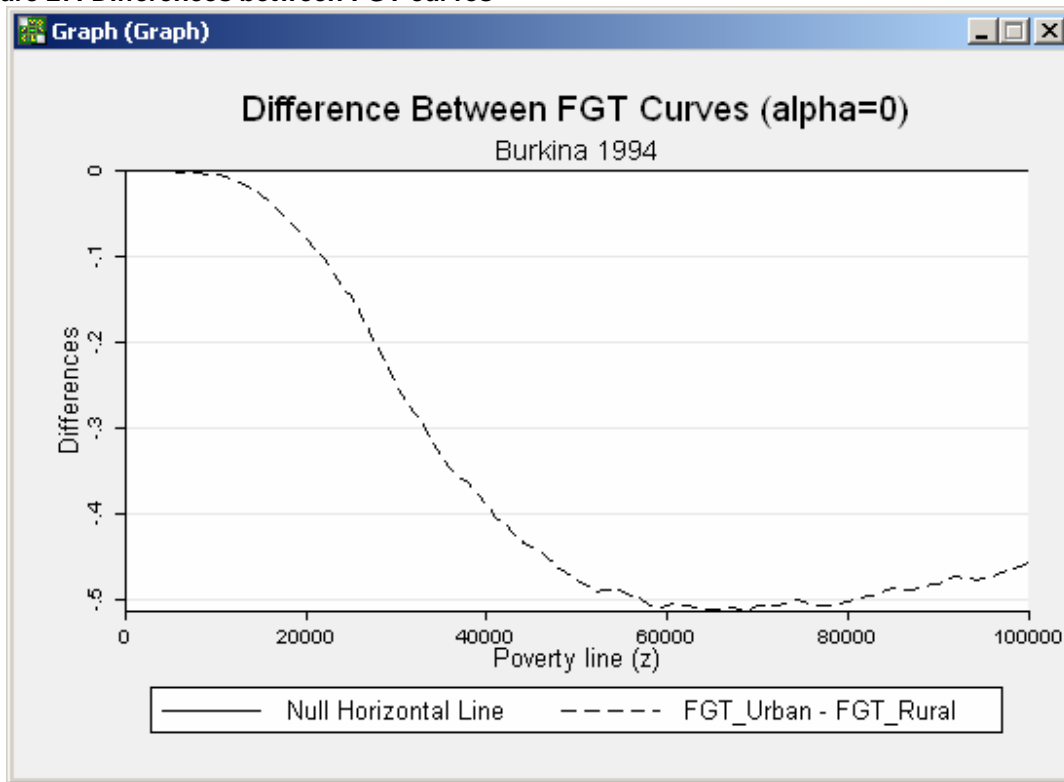
Q.6

Figure 28: Differences between FGT curves
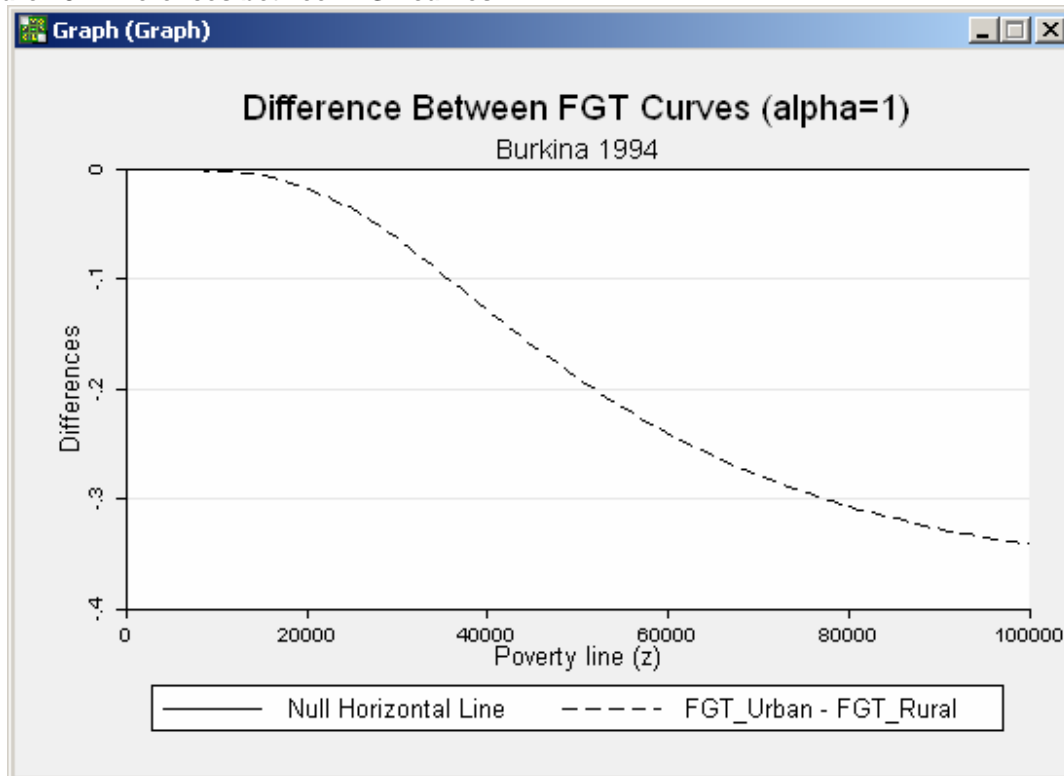
19.5 Estimating FGT curves and differences between FGT curves with confidence intervals

"Is the poverty increase between 1994 and 1998 in Burkina Faso statistically significant?"

1) Using the file **bkf94l.dta**, draw the FGT curve and its confidence interval for the variable of interest *exppc* with:
   a) parameter $\alpha = 0$;
   b) poverty line between 0 and 100,000 Franc CFA;
   c) size variable set to *size*.

2) Using simultaneously the files **bkf94l.dta** and **bkf98l.dta**, draw the difference between FGT curves and associated confidence intervals with:
   a) The variable of interest *exppc* for 1994 and *exppcz* for 1998.
   b) parameter $\alpha = 0$;
   c) poverty line between 0 and 100,000 Franc CFA;
   d) size variable set to *size*.

3) Redo 2) with parameter $\alpha = 1$.

Answers

Q.1

Steps:
   ▪ Type

use "C:\data\bkf94I.dta", clear

- To open the relevant dialog box, type
  db cfgts

- Choose variables and parameters as in

Figure 29: Drawing FGT curves with confidence interval

After clicking SUBMIT, the following appears:

Figure 30: FGT curves with confidence interval

FGT curve (alpha = 0)
Burkina Faso

Confidence interval (95 %) ——— Estimate

Q.2
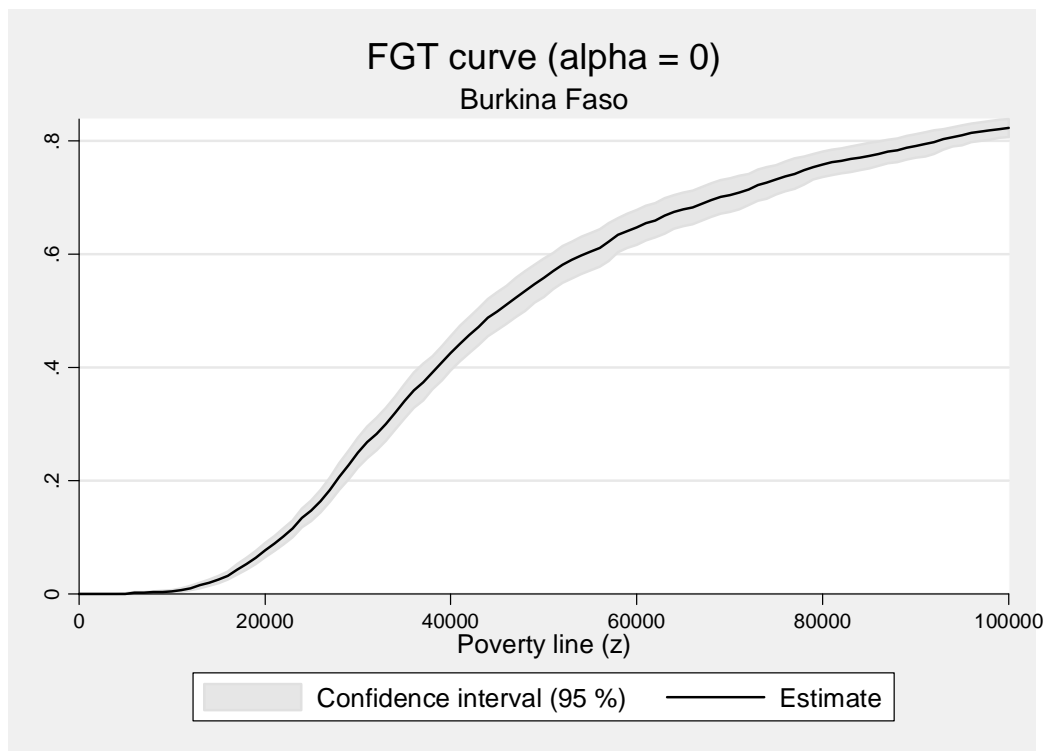
Steps:

- To open the relevant dialog box, type
  db cfgtsd2

- Choose variables and parameters as in

Figure 31: Drawing the difference between FGT curves with confidence interval

Figure 32: Difference between FGT curves with confidence interval $(\alpha = 0)$
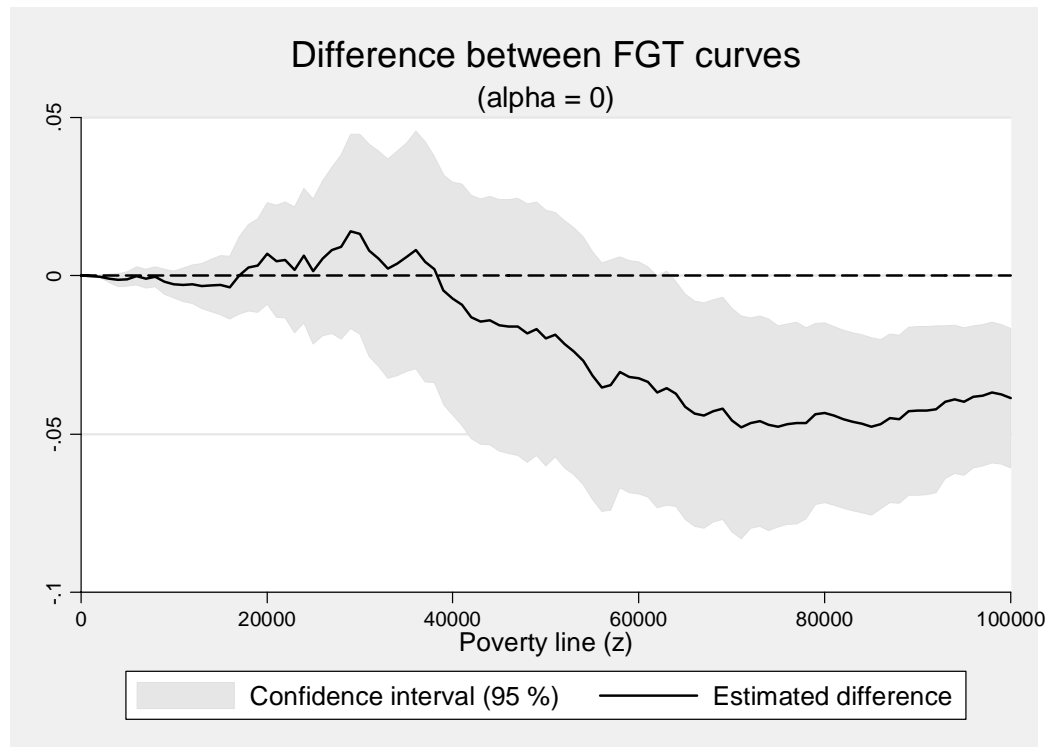
Figure 33: Difference between FGT curves with confidence interval $(\alpha = 1)$
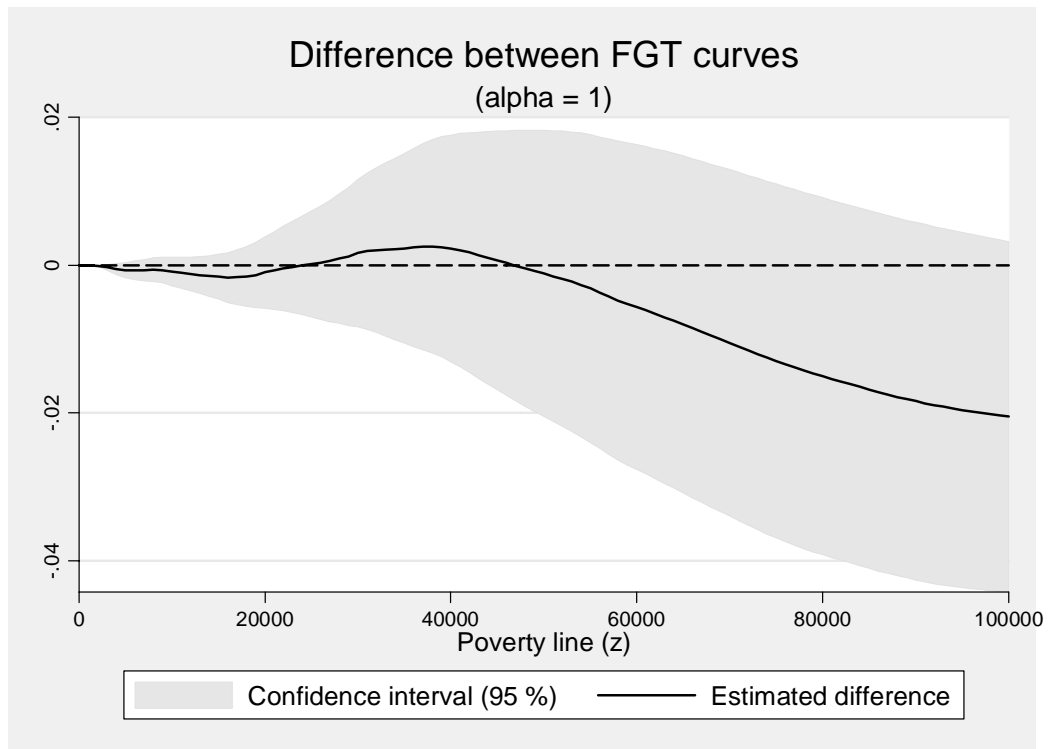
19.6 Testing poverty dominance and estimating critical values.

"Has the poverty increase in Burkina Faso between 1994 and 1998 been statistically significant?"

1) Using simultaneously files *bkf94l.dta* and *bkf98l.dta*, check for second-order poverty dominance and estimate the values of the poverty line at which the two FGT curves cross.
   a) The variable of interest is *exppc* for 1994 and *exppcz* for 1998;
   b) The poverty line should vary between 0 and 100,000 Franc CFA;
   c) The size variable should be set to *size*.

Answers

Q.1

Steps:

- To open the relevant dialog box, type
  db dompov

- Choose variables and parameters as in

Figure 34: Testing for poverty dominance

After clicking SUBMIT, the following results appear:


```
Number of      Critical     Min. range of   Max. range of           Case
intersection   pov. line    pov. lines       pov. line

        1      24262.871          .               .                   A
        2      46775.652          .               .                   B

Notes :
_case A: Before this intersection, distribution 2 dominates distribution 1.
_case B: Before this intersection, distribution 1 dominates distribution 2.
_case C: No dominance before this intersection.
```

19.7 Decomposing FGT indices.

"What is the contribution of different types of earners to total poverty in Burkina Faso?"

1. Open **bkf94I.dta** and decompose the average poverty gap
   a. with variable of interest *exppc*;
   b. with size variable set to *size*;
   c. at the official poverty line of 41099 Francs CFA;
   d. and using the group variable *gse* (Socio-economic groups).

2. Do the above exercise without standard errors and with the number of decimals set to 4.

Answers

Steps:

- Type
  use "C:\data\bkf94I.dta", clear

- To open the relevant dialog box, type
  db dfgtg

- Choose variables and parameters as in

Figure 35: Decomposing FGT indices by groups

After clicking SUBMIT, the following information is provided:


```
dfgtg exppc, hgroup(gse) hsize(size) alpha(1) pline(41099) type(nor)
```

FGT Index: Decomposition by Groups

Group	FGT Index	Population Share	Absolute Contribution	Relative Contribution
1: wage-earning (public sector)	0.004237	0.042971	0.000182	0.001308
	0.002571	0.003790	0.000117	0.000840
2: wage-earning (private sector)	0.022176	0.026598	0.000590	0.004237
	0.010678	0.002164	0.000291	0.002083
3: Artisan or trading	0.027741	0.062640	0.001738	0.012484
	0.004653	0.004288	0.000325	0.002371
4: Others activities	0.063853	0.006650	0.000425	0.003050
	0.025805	0.001308	0.000170	0.001203
5: Farmers (crop)	0.137525	0.104402	0.014358	0.103148
	0.011808	0.014896	0.002459	0.016980
6: Farmers (food)	0.162894	0.680885	0.110912	0.796800
	0.008643	0.016403	0.005823	0.019015
7: Inactive	0.144916	0.075856	0.010993	0.078973
	0.014994	0.004839	0.001332	0.008520
POPULATION	0.139197	1.000000	0.139197	1.000000
	0.006553	0.000000	0.006553	0.000000

Q.2

Using the RESULTS panel, change the number of decimals and unselect the option DISPLAY STANDARD ERRORS.

After clicking SUBMIT, the following information is obtained:

```
. dfgtg exppc, hgroup(gse) hsize(size) alpha(1) pline(41099) dstd(0) type(nor) dec(4)
```

FGT Index: Decomposition by Groups

Group	FGT Index	Population Share	Absolute Contribution	Relative Contribution
1: wage-earning (public sector)	0.0042	0.0430	0.0002	0.0013
2: wage-earning (private sector)	0.0222	0.0266	0.0006	0.0042
3: Artisan or trading	0.0277	0.0626	0.0017	0.0125
4: Others activities	0.0639	0.0066	0.0004	0.0031
5: Farmers (crop)	0.1375	0.1044	0.0144	0.1031
6: Farmers (food)	0.1629	0.6809	0.1109	0.7968
7: Inactive	0.1449	0.0759	0.0110	0.0790
POPULATION	0.1392	1.0000	0.1392	1.0000

19.8 Estimating Lorenz and concentration curves.

"How much do taxes and transfers affect inequality in Canada?"

By using the ***can6.dta*** file,
1. Draw the Lorenz curves for gross income X and net income N. How can you see the redistribution of income?
2. Draw Lorenz curves for gross income X and concentration curves for each of the three transfers *B1*, *B2* and *B3* and the tax *T*. What can you say about the progressivity of these elements of the tax and transfer system?

"What is the extent of inequality among Burkina Faso rural and urban households in 1994?"

By using the ***bkf94I.dta*** file,

3. Draw Lorenz curves for rural and urban households

      a. with variable of interest *exppc*;
      b. with size variable set to *size*;
      c. and using the group variable *zone* (as residential area).

Q.1
Steps:
- Type
  use "C:\data\can6.dta", clear

- To open the relevant dialog box, type
  db clorenz

- Choose variables and parameters as in

Figure 36: Lorenz and concentration curves

After clicking SUBMIT, the following appears:

Figure 37: Lorenz curves
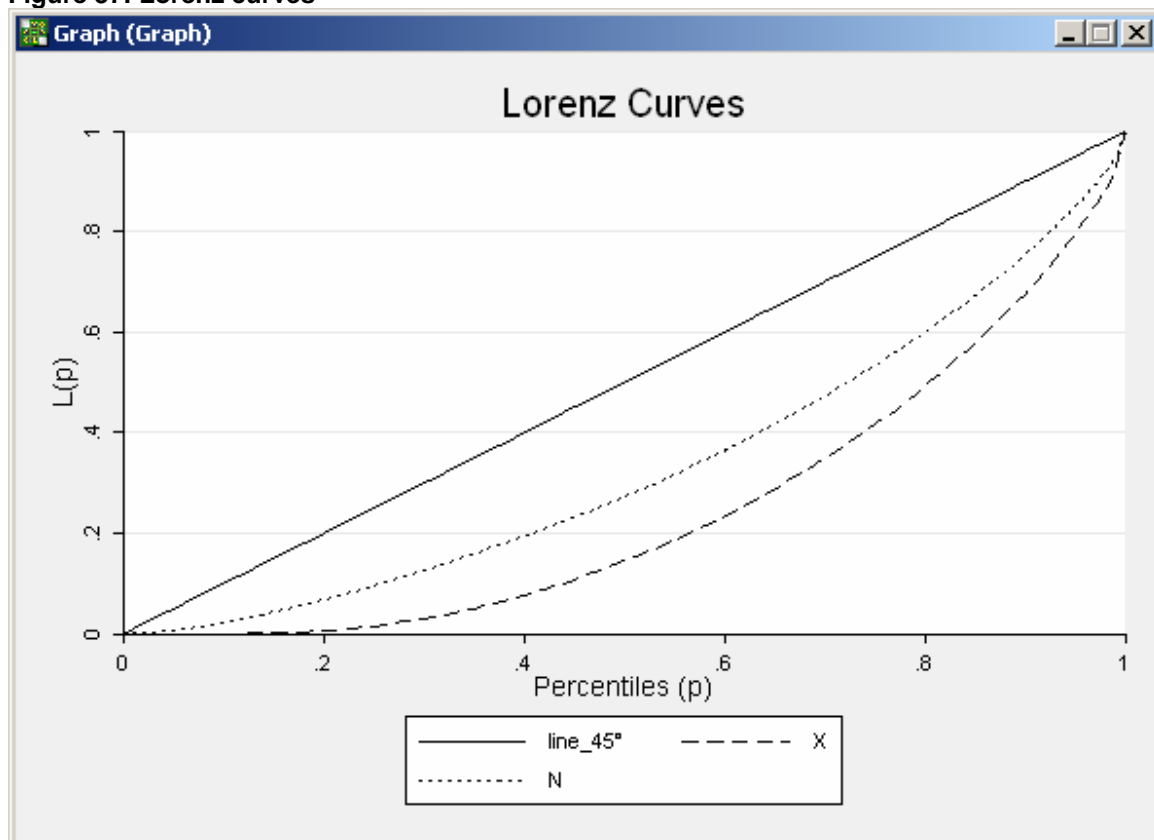
Steps:

- Choose variables and parameters as in

Figure 38: Drawing concentration curves

After clicking on SUBMIT, the following appears:

Figure 39: Lorenz and concentration curves
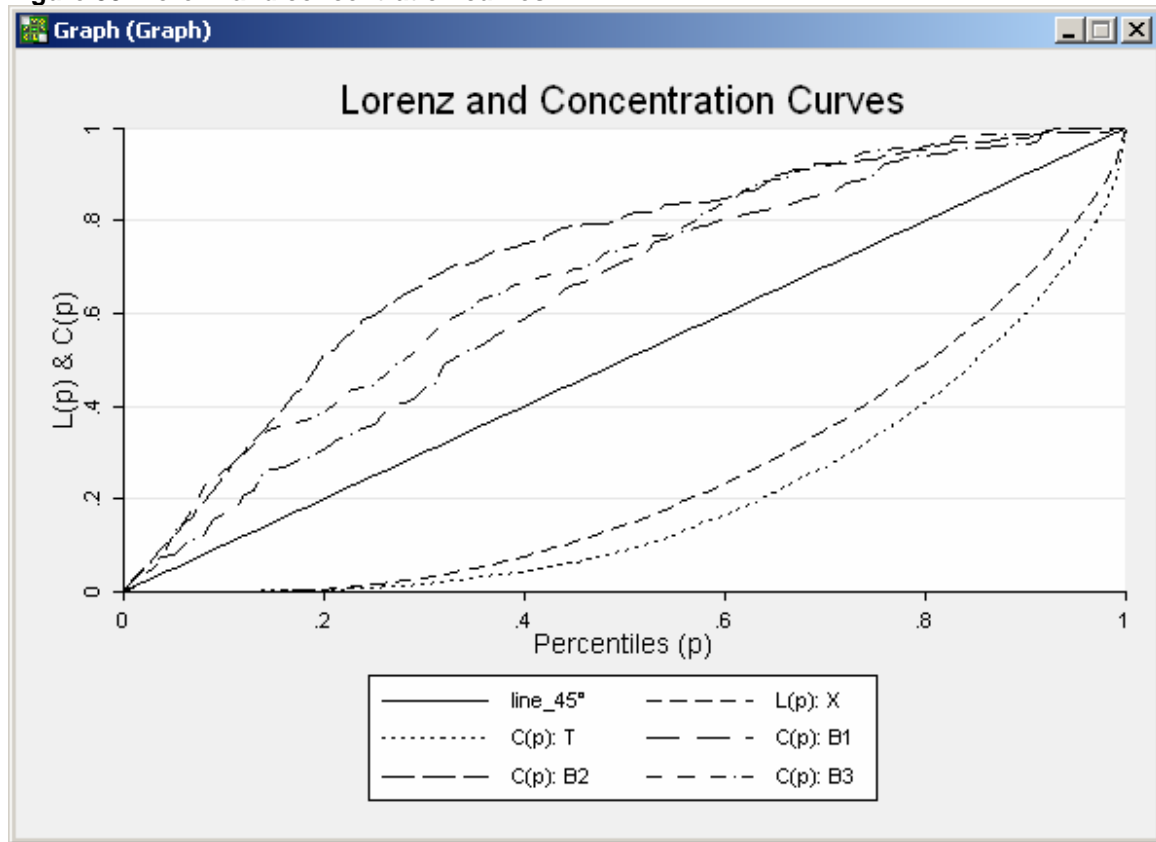
Q.3
Steps:
- Type
  use "C:\data\bkf94I.dta", clear

- Choose variables and parameters as in

Figure 40: Drawing Lorenz curves

Figure 41: Lorenz curves
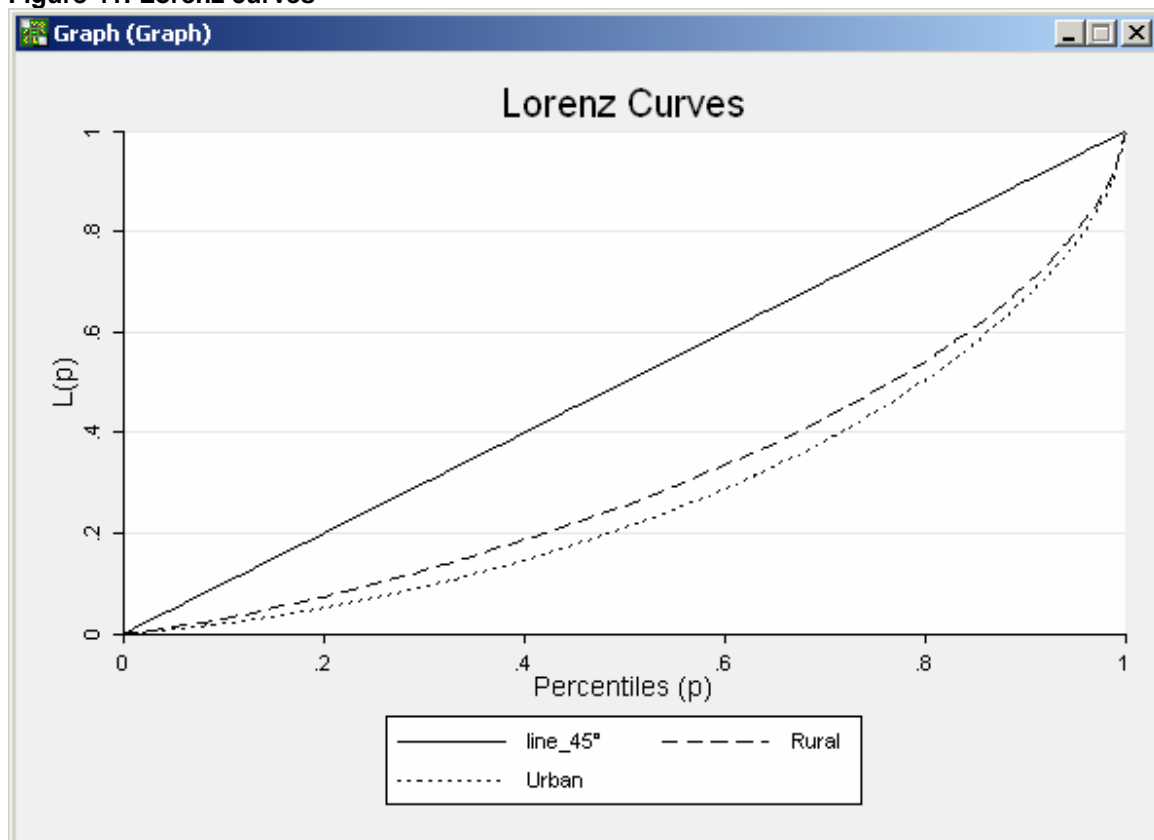
19.9 Estimating Gini and concentration curves

"By how much do taxes and transfers affect inequality in Canada?"

Using the **can6.dta** file,

1. Estimate the Gini indices for gross income *X* and net income *N*.
2. Estimate the concentration indices for variables *T* and *N* when the ranking variable is gross income *X.*

"By how much has inequality changed in Burkina Faso between 1994 and 1998?"

Using the **bkf94I.dta** file,

3. Estimate the difference in Burkina Faso's Gini index between 1998 and 1994

   a. with variable of interest *expeqz* for 1998 and *expeq* for 1994;
   b. with size variable set to *size.*

Q.1

Steps:

- Type
  use "C:\data\can6.dta", clear

- To open the relevant dialog box, type
  db igini

- Choose variables and parameters as in

Figure 42: Estimating Gini and concentration indices
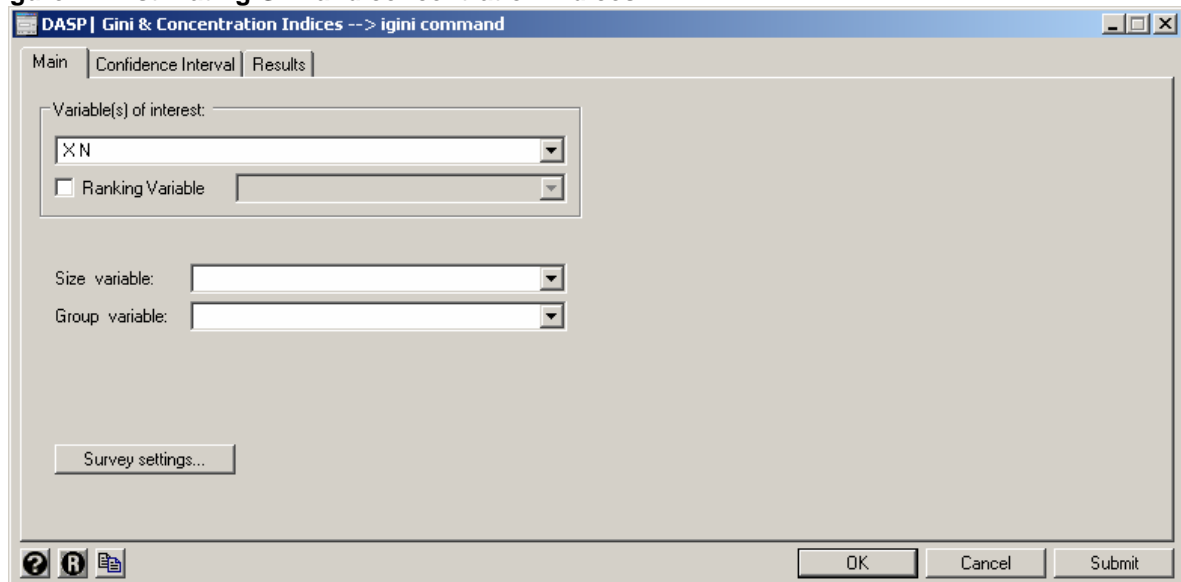
After clicking SUBMIT, the following results are obtained:

Variable	Estimate	STD	LB	UB
1: GINI_X	0.508456	0.016234	0.476599	0.540313
2: GINI_N	0.332355	0.012758	0.307318	0.357391

Q.2

Steps:

- Choose variables and parameters as in

Figure 43: Estimating concentration indices

After clicking SUBMIT, the following results are obtained:

Variable	Estimate	STD	LB	UB
1: CONC_T	0.595339	0.022931	0.550340	0.640338
2: CONC_N	0.306050	0.013268	0.280014	0.332087

Q.3

Steps:

- To open the relevant dialog box, type
  db digini

- Choose variables and parameters as in

Figure 44: Estimating differences in Gini and concentration indices

After clicking SUBMIT, the following information is obtained:

```
. digini expeqz expeq, file1(C:\data\bkf98I.dta) hsize1(size) file2(C:\data\bkf94I.dta) hsize2(size)
```

	Estimate	STD	LB	UB
Distribution_1:(GINI)	0.444563	0.012816	0.419371	0.469755
Distribution_2:(GINI)	0.450055	0.008618	0.433116	0.466994
Difference	-0.005492	0.015444	-0.035762	0.024778

19.10 Using basic distributive tools

"What does the distribution of gross and net incomes look like in Canada?"

Using the **can6.dta** file,

1. Draw the density for gross income *X* and net income *N*.
   - The range for the x axis should be [0, 60 000].
2. Draw the quantile curves for gross income *X* and net income *N*.
   - The range of percentiles should be [0, 0.8]
3. Draw the expected tax/benefit according to gross income *X.*
   - The range for the x axis should be [0, 60 000]
   - Use a *local linear estimation* approach.
4. Estimate marginal rates for taxes and benefits according to gross income *X.*
   - The range for the x axis should be [0, 60 000]
   - Use a *local linear estimation* approach.

Q.1
Steps:
- Type
  use "C:\data\can6.dta", clear

- To open the relevant dialog box, type
  db cdensity

- Choose variables and parameters as in

Figure 45: Drawing densities

After clicking SUBMIT, the following appears:

Figure 46: Density curves

Density Curves

Q.2
Steps:
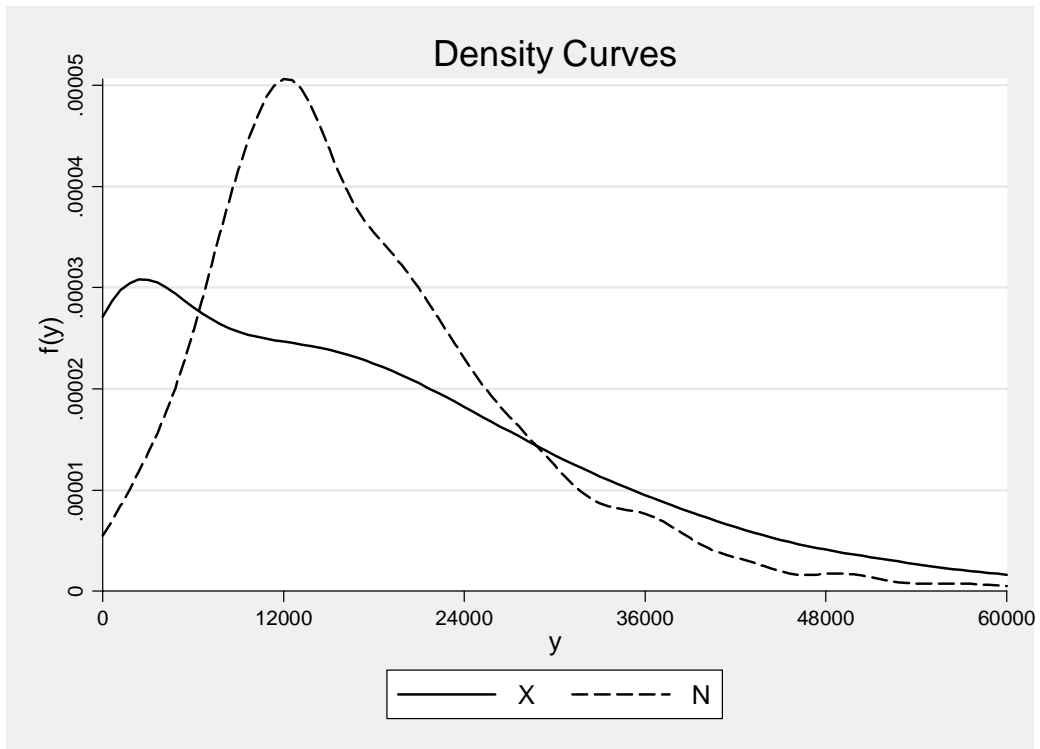
- To open the relevant dialog box, type
  db c_quantile

- Choose variables and parameters as in

79

Figure 47: Drawing quantile curves

After clicking SUBMIT, the following appears:

Figure 48: Quantile curves
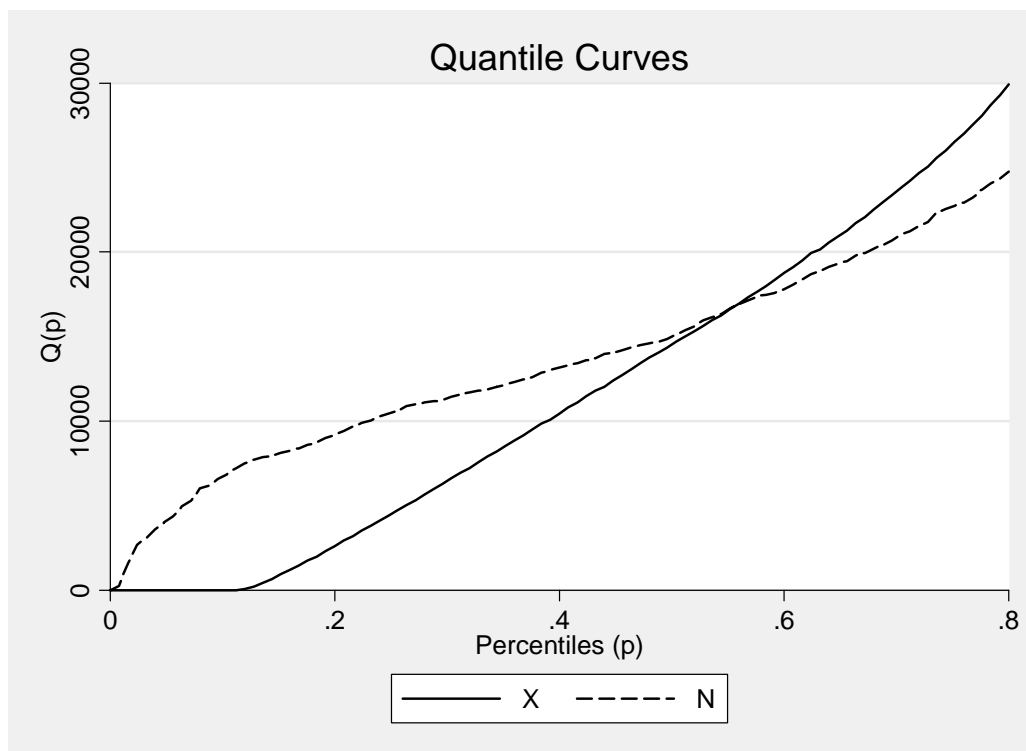
Steps:

- To open the relevant dialog box, type
  db cnpe

- Choose variables and parameters as in

Figure 49: Drawing non-parametric regression curves

After clicking SUBMIT, the following appears:

81

Figure 50: Non-parametric regression curves

Non parametric regression

(Linear Locally Estimation Approach | Bandwidth = 3699.26)
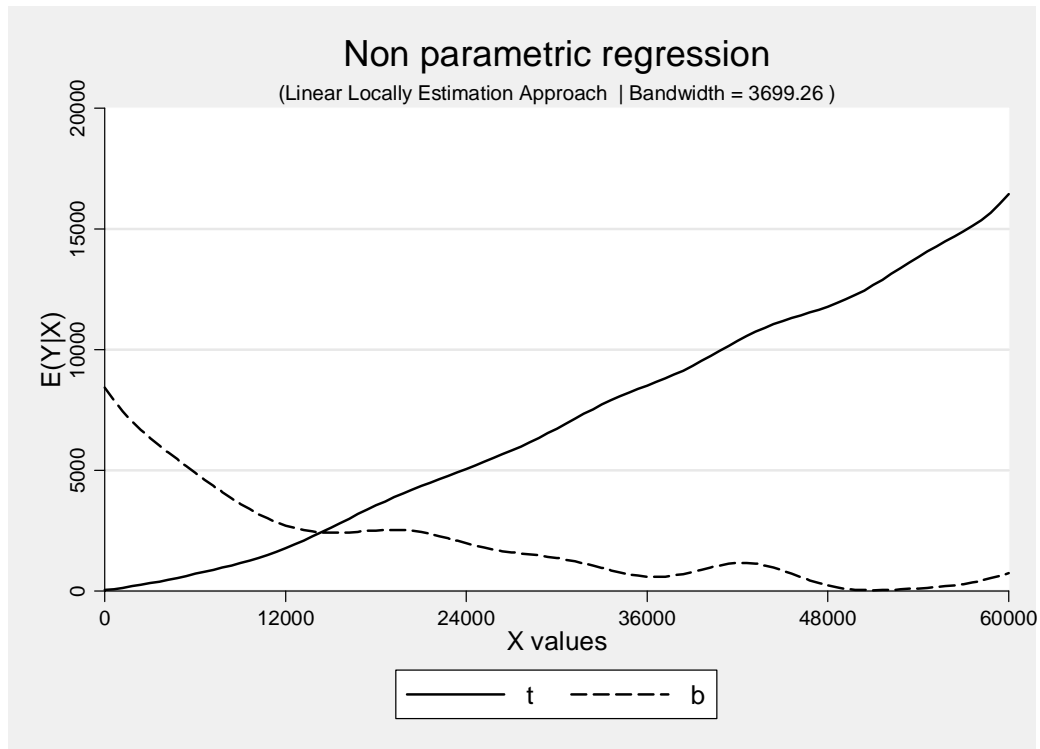
Q.4
Steps:

- Choose variables and parameters as in

Figure 51: Drawing derivatives of non-parametric regression curves

After clicking SUBMIT, the following appears:

Figure 52: Derivatives of non-parametric regression curves
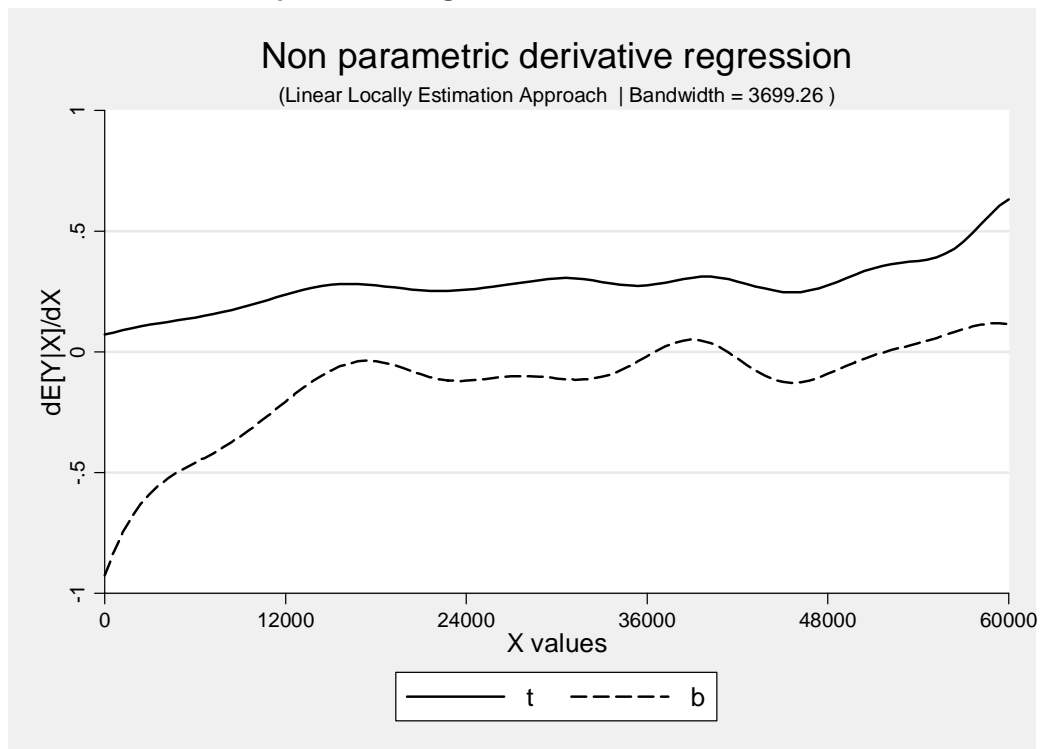
19.11 Plotting the joint density and joint distribution function

"What does the joint distribution of gross and net incomes look like in Canada?"

Using the **can6.dta** file,

4. Estimate the joint density function for gross income *X* and net income *N*.
   - o X range : [0,60000]
   - o N range : [0,60000]

5. Estimate the joint distribution function for gross income *X* and net income *N*.
   - o X range : [0,60000]
   - o N range : [0,60000]

Q.1

Steps:
- Type
  use "C:\data\can6.dta", clear

- To open the relevant dialog box, type
  db sjdensity

- Choose variables and parameters as in

Figure 53: Plotting joint density function

After clicking SUBMIT, the following graph is plotted interactively with Gnu Plot 4.2:

Joint Density Function

Q.2

Steps:

- To open the relevant dialog box, type
  db sjdistrub

- Choose variables and parameters as in

Figure 54: Plotting joint distribution function

After clicking SUBMIT, the following graph is plotted interactively with Gnu Plot 4.2:

Joint Distribution Function

19.12 Testing the bi-dimensional poverty dominance

Using the **columbia95l.dta** (distribution_1) and the **dominican_republic95l.dta** (distribution_2) files,

1. Draw the difference between the bi-dimensional multiplicative FGT surfaces and the confidence interval of that difference when

	Var. of interest	Range	alpha_j
Dimension 1	*haz* : height-for-age	-3.0 / 6.0	0
Dimension 2	*sprob* : survival probability	0.7 / 1.0	0

2. Test for bi-dimensional poverty using the information above.

Answer:

Q.1

Steps:

- To open the relevant dialog box, type
  db dombdpov

- Choose variables and parameters as in

Figure 55: Testing for bi-dimensional poverty dominance

After clicking SUBMIT, the following graph is plotted interactively with Gnu Plot 4.2:

Bi-dimensional poverty dominance

To make a simple test of multidimensional dominance, one should check if the lower-bounded confidence interval surface is always above zero for all combinations of relevant poverty lines – or conversely.
  o   For this, click on the panel "Confidence interval" and select the option lower-bounded.
  o   Click again on the button Submit.

After clicking SUBMIT, the following graph is plotted interactively with Gnu Plot 4.2:

Bi-dimensional poverty dominance
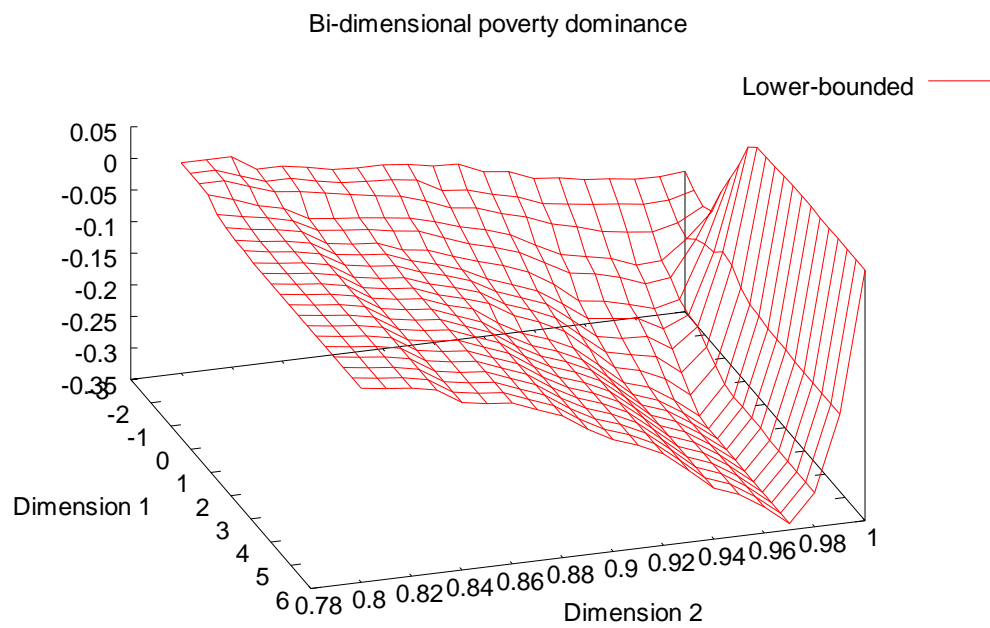
19.13   *Testing for pro-poorness of growth in Mexico*

The three sub-samples used in these exercises are sub-samples of 2000 observations drawn randomly from the three ENIGH Mexican household surveys for 1992, 1998 and 2004. Each of these three sub-samples contains the following variables:

strata   The stratum
psu   The primary sampling unit
weight   Sampling weight
inc   Income
hhsz   Household size

1. Using the files mex_92_2ml.dta and mex_98_2ml.dta, test for first-order relative pro-poorness of growth when:
   - The primal approach is used.
   - The range of poverty lines is [0, 3000].

2. Repeat with the dual approach.

3. By using the files mex_98_2ml.dta and mex_04_2ml.dta, test for absolute second-order pro-poorness with the dual approach.

4. Using mex_98_2ml.dta and mex_04_2ml.dta, estimate the pro-poor indices of module ipropoor.
   - Parameter alpha set to 1.
   - Poverty line equal to 600.

Answer:

Q.1

Steps:

- To open the relevant dialog box, type
db cpropoorp

- ▪ Choose variables and parameters as in (select the upper-bounded option for the confidence interval):

Figure 56: Testing the pro-poor growth (primal approach)

After clicking SUBMIT, the following graph appears

Relative propoor curve
(Order : s=1 | Dif. = P_2((m2/m1)z, a=s-1) - P_1(z,a=s-1))

Q.2

Steps:

- To open the relevant dialog box, type

db cpropoord

- Choose variables and parameters as in (with the lower-bounded option for the confidence interval):

Figure 57: Testing the pro-poor growth (dual approach)- A

After clicking SUBMIT, the following graph appears

Absolute propoor curves

(Order : s=1 | Dif. = Q_2(p) /Q_1(p) - mu_2/mu_1)

Difference
Lower bound of 95% confidence interval
Null horizontal line

Percentiles (p)

Q.2

Steps:

- To open the relevant dialog box, type
db cpropoord

- Choose variables and parameters as in (with the lower-bounded option for the confidence interval):

Figure 58: Testing the pro-poor growth (dual approach) – B

After clicking SUBMIT, the following graph appears

Absolute propoor curves

(Order : s=2 | Dif. = (GL_2(p) - GL_1(p)) / GL_2(p))

Difference   Lower bound of 95% confidence interval
Null horizontal line

Q.4

Steps:

▪ To open the relevant dialog box, type
db ipropoor

- Choose variables and parameters as.

After clicking SUBMIT, the following results appear:

```
Poverty line    :       600.00
Parameter alpha :         1.00
```

Pro-poor indices	Estimate	STD	LB	UB
Growth rate(g)	0.582359	0.125512	0.336361	0.828357
Chen & Ravallion (2003) index	0.712285	1.009337	-1.265979	2.690549
Kakwani & Pernia (2000) index	1.325436	0.107047	1.115627	1.535244
PEGR index	0.771879	0.137331	0.502716	1.041042
PEGR - g	0.189520	0.049357	0.092783	0.286257

19.14 Benefit incidence analysis of public spending on education in Peru (1994).

1. Using the peredu94I.dta file, estimate participation and coverage rates of two types of public spending on education when:
   - The standard of living is **exppc**
   - The number of household members that benefit from education is **fr_prim** for the primary sector and **fr_sec** for the secondary one.

- The number of eligible household members is **el_prim** for the primary sector and **el_sec** for the secondary one.
- Social groups are quintiles.

Type **db bian** in the windows command and set variables and options as follows:

Figure 59: Benefit incidence analysis

After clicking on **Submit,** the following appears:

```
Benefit Incidence Analysis: Education
```

Share by Quintile Groups.

Groups	Primary	Secondary
Quintile 1	0.218	0.155
Quintile 2	0.226	0.216
Quintile 3	0.220	0.224
Quintile 4	0.197	0.231
Quintile 5	0.139	0.173
All	1.000	1.000

Rate of Participation by Quintile Groups.

Groups	Primary	Secondary
Quintile 1	0.797	0.458
Quintile 2	0.825	0.641
Quintile 3	0.802	0.663
Quintile 4	0.723	0.687
Quintile 5	0.506	0.511
All	0.730	0.592

2. To estimate total public expenditures on education by sector at the national level, the following macro information was used:
    - Pre-primary and primary public education expenditure (as % of all levels), 1995: 35.2%
    - Secondary public education expenditure (as % of all levels), 1995: 21.2%
    - Tertiary public education expenditure (as % of all levels), 1995: 16%
    - Public education expenditure (as % of GNP), 1995 = 3%
    - GDP per capita: about 3 800.

    Using this information, the following variables are generated

    ```
    cap drop _var1;
    gen _var1 = size*weight*3800;
    qui sum _var1;
    qui gen pri_pub_exp=0.03*0.352*`r(sum)';
    qui gen sec_pub_exp=0.03*0.212*`r(sum)';
    qui gen uni_pub_exp=0.03*0.160*`r(sum)';
    cap drop _var1;
    ```

    - Total public expenditures on primary sector   : **pri_pub_exp**
    - Total public expenditures on secondary sector : **sec_sec_exp**
    - Total public expenditures on university sector : **uni_pub_exp**

    Estimate the average benefits per quintile and generate the benefit variables.

Answer:

Set variables and options as follows:

Figure 60: Benefit Incidence Analysis (unit cost approach)

After clicking on **Submit,** the following appears:

Average Benefits by Quintile Groups: (at the level of eligible members)

Groups	Sector_1	Sector_2
Quintile 1	248.662	128.548
Quintile 2	257.483	179.816
Quintile 3	250.395	186.119
Quintile 4	225.527	192.840
Quintile 5	157.982	143.327
All	227.961	166.095

Average Benefits by Quintile Groups: (at the level of members that use the public service)

Groups	Sector_1	Sector_2
Quintile 1	312.084	280.540
Quintile 2	312.084	280.540
Quintile 3	312.084	280.540
Quintile 4	312.084	280.540
Quintile 5	312.084	280.540
All	312.084	280.540

Proportion of Benefits by Quintile Groups and by Sectors.

Groups	Sector_1	Sector_2
Quintile 1	0.136	0.058
Quintile 2	0.141	0.081
Quintile 3	0.137	0.084
Quintile 4	0.123	0.087
Quintile 5	0.087	0.065
All	0.624	0.376