

## Stochastic Frontier Analysis for Healthcare, with Illustrations in R

### APPENDIX I3.1 R CODE FOR SYNTHETIC DATA-SET GENERATION

Due to confidentiality, we cannot provide the Queensland hospital data set used in our illustrations. As an alternative, we provide a synthetic data set to the interested readers to illustrate the SFMs introduced in this chapter with R.

The R code used to generate the synthetic panel data set is as noted in Box I3.1, in which we assume a scenario of 100 decision-making units (e.g., firms, factories, banks, and hospitals) in a period of four years. The production process contains three inputs, one output, and three binary environmental factors.

When using a replicable data generation process (i.e., setting the seed of the random number generator as 99 in the following code), the readers can obtain the inefficiency estimation as reported in Table I3.4 and Figure I3.2 from R.

TABLE I3.4 *Statistics of the estimated level of inefficiency (%) by different models with the synthetic data set*

Model	Mean	Std. dev.	Min	Q1	Median	Q3	Max
ALS77	44.98%	6.83%	30.21%	39.96%	44.74%	49.49%	67.23%
SS84	58.44%	20.76%	0.00%	41.93%	59.78%	70.44%	100.00%
PL81	39.66%	10.87%	18.27%	30.54%	38.91%	45.16%	64.35%
CSS90	52.75%	26.72%	0.00%	36.63%	56.69%	73.52%	100.00%
BC92	56.86%	9.47%	30.21%	49.06%	57.69%	62.86%	74.73%
G05	0.01%	0.00%	0.01%	0.01%	0.01%	0.01%	0.01%
KLH14T	0.50%	0.00%	0.49%	0.49%	0.50%	0.50%	0.50%
KLH14P	2.79%	0.05%	2.65%	2.75%	2.79%	2.82%	2.89%
BC95	21.24%	15.81%	0.00%	13.53%	23.47%	23.47%	52.05%
SVKZ	39.14%	25.60%	0.00%	0.00%	51.25%	59.26%	73.00%

```
1 rm(list=ls())
2 graphics.off()
3 # Generate synthetic panel data set
4 # -----
5 set.seed(99) # Set fixed seed for synthetic data
6
7 # Unit ID and Year:
8 id = sort(rep(1:100, times = 4))
9 Year = rep(1:4, times = 100)
10 t = Year
11 t2 = t^2
12
13 # Output
14 Y = rnorm(400, mean=-2, sd=1.6)
15
16 # Inputs
17 X1 = rnorm(400, mean=3.3, sd=1.3)
18 X2 = rnorm(400, mean=-1.6, sd=1.5)
19 X3 = rnorm(400, mean=13.6, sd=2)
20
21 # Environmental
22 #Assume 20% of units are 1 in Z1
23 Z1 = rep(c(1, 0), times= c(20,80))[order(runif(100))]
24 Z1 = rep(Z1, each = 4) # Replicate for a panel data set
25 Z2 = rep(c(1, 0), times= c(30,70))[order(runif(100))]
26 Z2 = rep(Z2, each = 4)
27 Z3 = rep(c(1, 0), times= c(80,20))[order(runif(100))]
28 Z3 = rep(Z3, each = 4)
29
30 # Composed data set
31 data = as.data.frame(cbind(id, Year, Y, X1, X2, X3, Z1, Z2, Z3, t, t2))
32 # Convert to panel data
33 library(plm)
34 paneldata<- pdata.frame(data, c("id","Year"))
35
36 # Define formulas for different models
37 form = Y ~ X1 + X2 + X3 # For ALS77/SS84/PL81/BC92/KLH14
38 formz = Y ~ X1 + X2 + X3 | Z1 + Z2 + Z3 # For BC95
39 formt = Y~ X1 + X2 + X3 + factor(id) # For G05
40 formc = Y~ X1 + X2 + X3 + t + t2 # For CSS90
41
42 fZ1 = as.factor(Z1)
43 fZ2 = as.factor(Z2)
44 fZ3 = as.factor(Z3)
45
46 forms = Y ~ X1 + X2 + X3 + fZ1 + fZ2 + fZ3 # For SVKZ
47 forms3 = ehat3 ~ X1 + X2 + X3 + fZ1 + fZ2 + fZ3 #For SVKZ
48
49 # The synthetic data set can be then applied with the SFM code as
      illustrated in Appendix B.
```

BOX 13.10: *Synthetic data set generation process*

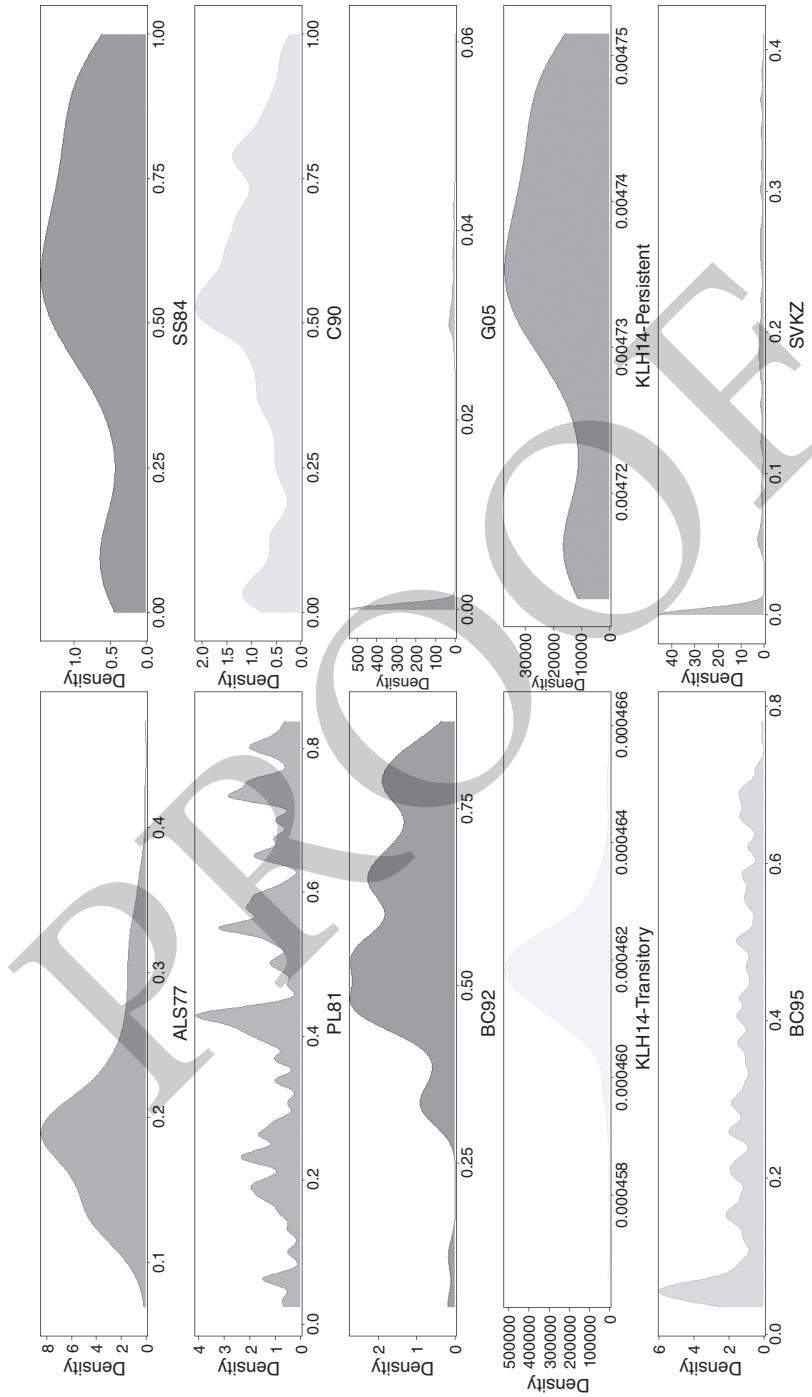


FIGURE 13.2 Estimated kernel densities of the inefficiency level of different SFMs with the synthetic data set

```
> summary(als77)
Error Components Frontier (see Battese & Coelli 1992)
Inefficiency decreases the endogenous variable (as in a production function)
The dependent variable is logged
Iterative ML estimation terminated after 8 iterations:
log likelihood values and parameters of two successive iterations
are within the tolerance limit

final maximum likelihood estimates
  Estimate Std. Error z value Pr(>|z|)
(Intercept) -1.751327  1.184760 -1.4782  0.13935
X1          0.032920  0.063208  0.5208  0.60249
X2          0.012927  0.053784  0.2403  0.81006
X3          0.010792  0.043463  0.2483  0.80391
sigmasq     3.057531  1.393798  2.1937  0.02826 *
gamma       0.188451  0.623408  0.3023  0.76243
---
signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1
log likelihood value: -765.5158

cross-sectional data
total number of observations = 400

mean efficiency: 0.5973101
```

FIGURE 13.3 Model estimation of ALS77 with the synthetic data set

Moreover, taking the output of the ALS77 model as an example, the R output with our synthetic data set is as shown in Figure 13.3. The production function is estimated after eight iterations with a final log likelihood value of  $-765.5158$ . The parameters estimated are as listed in the second column, which however, are all insignificant in this case. The estimated variances of the two error terms (i.e.,  $\sigma_u^2$  and  $\sigma_v^2$ ) are reported afterward, where  $\sigma^2 = \sigma_u^2 + \sigma_v^2$  and  $\gamma = \sigma_u^2/\sigma^2$ . Moreover, the mean efficiency, which as noted in Section 13.2.1 as  $E[\exp(-u)]$  is also reported at the end as 59.73 %, whereas the individual inefficiency level can be estimated with the subsequent commands.

## APPENDIX I3.2 FULL R CODE FOR THE EMPIRICAL ILLUSTRATIONS

```
I 1 rm(list=ls())
2 graphics.off()
3
4 # Data Process for the Queensland hospital data for illustrations
5 # -----
6 # Read data
7 data <- read.csv("QLD.csv")
8 names(data)[names(data)=="HOSID"] <- "id"
9 names(data)[names(data)=="Yeardummy"] <- "Year"
10
11 # Convert to panel data
12 library(plm)
13 paneldata<- pdata.frame(data, c("id","Year"))
14 # Define formulas for different models
15 form = lAggout ~ lBEDS + lAgglabours + lSUPP
16 formz = lAggout ~ lBEDS + lAgglabours + lSUPP | TEACH + Remote + Small
17 formt = lAggout ~ lBEDS + lAgglabours + lSUPP + factor(id)
18 t = data$Year
19 t2 = t^2
20 formc = lAggout ~ lBEDS + lAgglabours + lSUPP + t + t2
21
22 # Define factor variables and formulas for SVKZ
23 attach (data)
24 fTEACH = as.factor(TEACH)
25 fRemote = as.factor(Remote)
26 fSmall = as.factor(Small)
27 forms = lAggout ~ lBEDS + lAgglabours + lSUPP + fTEACH + fRemote +
28 fSmall
29 forms3 = ehat3 ~ lBEDS + lAgglabours + lSUPP + fTEACH + fRemote +
30 fSmall
31 # Define output variable
32 Y = lAggout
33 ##### The following code can be applied with the synthetic data #####
34 ### The following code can be applied with the synthetic data #####
35 ### or the Queensland hospital data processed above. #####
36 #####
37
38 # Code for SFMs -----
39 library(frontier)
40 ##### ALS77 #####
41 attach(data)
42 # Model estimation
43 als77 <- sfa (form, data = data, ineffDecrease = T, truncNorm = F,
44 timeEffect = F)
45 summary(als77)
46 # Individual inefficiency by JLMS
47 # Coefficients
48 als77coef <- coef(als77, which = "mle", extraPar = T)
49 # epsilon from fitted values
50 fals77 <- fitted(als77, asInData = T)
ei = Y - fals77
```

```

51 # E(ui|ei)
52 us2 = (als77$coef[["sigmaU"]])^2
53 vs2 = (als77$coef[["sigmaV"]])^2
54 sigmistar = sqrt((vs2*us2)/(vs2+us2))
55 ustari = (-us2*ei)/(vs2+us2)
56 uals77 = ((sigmistar*dnorm(ustari/sigmistar))/(pnorm(ustari/sigmistar
      )))+ustari
57 inals77 = 1 - exp(-uals77) # Inefficiency estimation of individual
      unit
58 summary(inals77)
59
60 ##### SS84 #####
61 attach(paneldata)
62 # Fixed effects for ui
63 ss84 <- plm(form, data = paneldata,
      model = "within", index = c("id", "Year"), effect =
      "individual")
64 summary(ss84)
65 # ai and uihat
66 ai = as.numeric(unname(fixef(ss84)))
67 uss84 = max(ai) - ai
68 inss84 = (uss84 - min(uss84))/(max(uss84)-min(uss84)) # Relative
      inefficiency of individual unit
69 summary(inss84)
70
71 ##### PL81 #####
72 attach(paneldata)
73 # Fixed effects
74 pl81f <- plm(form, data = paneldata,
      model = "random", index = c("id", "Year"), effect =
      "individual")
75 summary(pl81f)
76 # Extract initial values
77 sigmau2 = as.numeric(ercomp(pl81f)[["sigma2"]][2])
78 sigmav2 = as.numeric(ercomp(pl81f)[["sigma2"]][1])
79 sigmasq = sigmau2+sigmav2
80 gamma = sigmau2/sigmasq
81
82 init_para = c(as.numeric(pl81f$coefficients), sigmasq, gamma)
83 # MLE with initial values
84 pl81 <- sfa(form, data = paneldata, ineffDecrease = T, truncNorm = F,
      timeEffect = F,
      startVal = init_para)
85 summary(pl81)
86
87 # Estimate individual inefficiency
88 inpl81 <- 1-efficiencies(pl81, asInData = T, logDepVar = T, minusU =
      T)
89 summary(inpl81)
90
91 ##### CSS90 #####
92 attach(paneldata)
93 # Fixed effects for ui
94 c90ori <- plm(formc, data = paneldata,
      model = "within", index = c("HOSID", "Yeardummy"),
      effect = "twoways")
95 summary(c90ori)

```

*Stochastic Frontier Analysis for Healthcare, with Illustrations in R*

```

100 ait = lAggout - lBEDS*as.numeric(c90ori[["coefficients"]]) [1] -
101   lAgglabours*as.numeric(c90ori[["coefficients"]]) [2] - lSUPP*as.
102   numeric(c90ori[["coefficients"]]) [3]
103
104   # Calculate ajt
105   ti = c(1, 2, 3, 4)
106   ti2 = ti^2
107   aith = numeric()
108
109   for (i in 1:length(unique(id))){ 
110     aith = c(aith, as.numeric(unname(
111       lm(ait[(4*(i-1)+1):(4*(i-1)+4)] ~ ti2 + ti)$fitted.values)))
112   }
113
114   c90 = cbind(aith, t)
115   # athat for each year
116   ajtyear = aggregate(aith ~ t, data = c90, max)
117   # Compute uithat
118   c90 = as.data.frame(merge(c90, ajtyear, by = "t"))
119   colnames(c90) <- c("t","ait","ajt")
120   uc90 = c90$ajt - c90$ait
121   summary(uc90)
122   inc90 = (uc90 - min(uc90))/(max(uc90)-min(uc90)) # Relative
123   inefficiency of individual unit
124   summary(inc90)
125
126 ##### BC92 #####
127 attach(paneldata)
128 # Model estimation
129 bc92 <- sfa (form, data = paneldata, ineffDecrease = T, truncNorm = T,
130   timeEffect = T)
131 summary(bc92, effic = F, logDepVar = T, effMinusU = T)
132
133 # Estimate individual inefficiency
134 inbc92 <- 1-efficiencies (bc92, asInData = T, logDepVar = T, minusU =
135   T)
136 summary(inbc92)
137
138 ##### G05 #####
139 attach(paneldata)
140 # Model estimation
141 g05 <- sfa (formt, data = paneldata, ineffDecrease = T, truncNorm = F,
142   timeEffect = T)
143 summary(g05, effic = F, logDepVar = T, effMinusU = T)
144
145 # Estimate individual inefficiency
146 ing05 <- 1-efficiencies (g05, asInData = T, logDepVar = T, minusU = T
147   )
148 summary(ing05)
149
150 ##### KLH14 #####
151 attach(paneldata)
152 # Fixed effects for alphai and epsilonit
153 klh14ori <- plm(form, data = paneldata,
154   model = "within", index = c("id","Year"), effect = "
155   individual")
156 summary(klh14ori)
157 # alphai

```

```
150 | alphai = as.numeric(unname(fixef(klh14ori)))
151 | # epsilonit
152 | epsiloni = as.numeric(klh14ori[["residuals"]])
153 | summary(alphai)
154 | summary(epsiloni)
155 | # Constant for sfa estimation
156 | constant = 1
157 | # Persistent inefficiency
158 | klh14per = as.data.frame(cbind(alphai,constant))
159 | klh14p <- sfa(alphai~constant-1, data = klh14per, ineffDecrease = T,
160 |   truncNorm = F, timeEffect = F)
161 | summary(klh14p, effic = F, logDepVar = T, effMinusU = T)
162 | # Estimate individual persistent inefficiency
163 | ink1h14p <- 1-efficiencies (klh14p, asInData = T, logDepVar = T,
164 |   minusU = T)
165 | summary(ink1h14p)
166 | # Transitory inefficiency
167 | klh14tran = as.data.frame(cbind(epsiloni,constant))
168 | klh14t <- sfa(epsiloni~constant-1, data = klh14tran, ineffDecrease = T,
169 |   truncNorm = F, timeEffect = T)
170 | summary(klh14t, effic = F, logDepVar = T, effMinusU = T)
171 | # Estimate individual transitory inefficiency
172 | ink1h14t <- 1-efficiencies (klh14t, asInData = T, logDepVar = T,
173 |   minusU = T)
174 | summary(ink1h14t)
175 | ##### BC95 #####
176 | attach(paneldata)
177 | # Model estimation
178 | bc95 <- sfa(formz, data = paneldata, ineffDecrease = T, truncNorm = T,
179 |   timeEffect = F)
180 | summary(bc95, effic = F, logDepVar = T, effMinusU = T)
181 | # Estimate individual inefficiency
182 | inbc95 <- 1-efficiencies (bc95, asInData = T, logDepVar = T, minusU =
183 |   T)
184 | ##### SVKZ #####
185 | library(np)
186 |
187 | # Bandwidths selection
188 | bws.r1 <- npregbw(forms, regtype="ll", data=data, bwmethod = "cv.ls",
189 |   ckertype = "epanechnikov")
190 | # Estimate conditional mean and extract fitted values and residuals
191 | r1.est <- npreg(bws=bws.r1, gradients=TRUE)
192 | r1hat <- fitted(r1.est)
193 | elhat <- residuals(r1.est)
194 | # Generate for r3
195 | ehat3 <- elhat^3
196 | # Bandwidths selection for r3 (skewness measures)
197 | bws.r3 <- npregbw(forms3, data=data, regtype="ll", bwmethod = "cv.ls",
198 |   ckertype = "epanechnikov")
199 | # Estimate conditional mean and extract fitted values and residuals
  r3.est <- npreg(bws=bws.r3, gradients=TRUE)
```

```

200 r3hat <- fitted(r3.est)
201 e3hat <- residuals(r3.est)
202
203 # Estimate individual inefficiency
204 sigu3.hat <- sqrt(pi/2)*(pi/(pi-4))*r3hat
205 sigu.hat <- apply(cbind(0,sigu3.hat),1,FUN=max)^(1/3)
206 muhat <- sqrt(2/pi)*sigu.hat
207 summary(muhat)
208 insvkz = 1 - exp(-muhat) # Inefficiency estimation of individual unit
209 summary(insvkz)
210
211 # Statistics of results -----
212 library(qpcR)
213 # Combine inefficiency vectors of different length
214 ineffs <- qpcR:::cbind.na(inals77, inss84, inpl81, inc90, inbc92,
215   ing05, inklh14t, inklh14p, inbc95, insvkz)
216 # Apply statistical analysis to each model
217 stat = list("mean" = apply(ineffs, 2, mean, na.rm = T),
218   "sd" = apply(ineffs, 2, sd, na.rm = T),
219   "min" = apply(ineffs, 2, min, na.rm = T),
220   "Q1" = apply(ineffs, 2, quantile, probs=0.25, na.rm = T),
221   "Median" = apply(ineffs, 2, quantile, probs=0.5, na.rm =
222     T),
223   "Q3" = apply(ineffs, 2, quantile, probs=0.75, na.rm = T),
224   "max" = apply(ineffs, 2, max, na.rm = T))
225
226 write.csv(stat, file = 'Stats of inefficiency.csv')
227
228 # Correlation between estimates and confidence interval of mean
229 # inefficiency
230 cor(ineffs, method = c("spearman"))
231 library(tidyverse)
232 ci <- function(x) {
233   t.test(x, conf.level = 0.95)$conf.int
234 }
235 apply(ineffs, 2, ci)
236
237 # Rank of individual unit -----
238 # Ranks of panel data setting
239 Rankp = cbind(data$id, data$Year,
240   rank(inals77),
241   as.integer(as.factor(rank(inpl81, ties.method =
242     "average"))), #Tune for duplicate values
243   as.integer(as.factor(rank(inc90, ties.method = "average
244     "))),
245   rank(inbc92),
246   rank(inbc95),
247   rank(ing05),
248   rank(inklh14t),
249   as.integer(as.factor(rank(insvkz, ties.method =
250     "average"))))

# Rename for Rankp
Name <- c("id", "Year", "ALS77", "PL81", "CSS90", "BC92", "BC95", "G05",
      "KLH14-transitory", "SVKZ")
colnames(Rankp) <- Name
# Ranks of cross-sectional data setting

```

10 *Stochastic Frontier Analysis for Healthcare, with Illustrations in R*

```
251 Rankc = cbind(unique(data$id), #Extract id for individual unit
252                 rank(inss84),
253                 rank(inklh14p))
254
255 # Rename for Rankc
256 Name <- c("id", "SS84", "KLH14-persistent")
257 colnames(Rankc) <- Name
258
259 cor(Rankp, method = c("spearman"))
260 cor(Rankc, method = c("spearman"))
261
262 # Kernel densities of estimations -----
263 require("ggplot2")
264 .df <- na.omit(data.frame(x = inals77))
265 .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
266 .als<-ggplot(data = .df, aes(x = x, y = ..density..)) +
267   # Epanechnikov kernel and CV bandwidth
268   geom_density(
269     kernel = "gaussian",
270     bw = "ucv",
271     alpha = 0.5,
272     #Here for single group: color and fill without aes()
273     color = "brown1", fill = "brown1",
274     #shut the legend
275     show.legend = FALSE
276   ) +
277   scale_y_continuous(expand = c(0.01, 0)) +
278   xlab("ALS77") +
279   ylab("Density") +
280
281   RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
282                                         "sans")
283 print(.als)
284 rm(.df, .nbins)
285
286 require("ggplot2")
287 .df <- na.omit(data.frame(x = inss84))
288 .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
289 .ss84<-ggplot(data = .df, aes(x = x, y = ..density..)) +
290   # Epanechnikov kernel and CV bandwidth
291   geom_density(
292     kernel = "gaussian",
293     bw = "ucv",
294     alpha = 0.5,
295     #Here for single group: color and fill without aes()
296     color = "darkgreen", fill = "darkgreen",
297     #shut the legend
298     show.legend = FALSE
299   ) +
300   scale_y_continuous(expand = c(0.01, 0)) +
301   xlab("SS84") +
302   ylab("Density") +
303
304   RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
305                                         "sans")
306 print(.ss84)
307 rm(.df, .nbins)
```

```

307 require("ggplot2")
308 .df <- na.omit(data.frame(x = inpl81))
309 .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
310 .pl81<-ggplot(data = .df, aes(x = x, y = ..density..)) +
311   # Epanechnikov kernel and CV bandwidth
312   geom_density(
313     kernel = "gaussian",
314     bw = "ucv",
315     alpha = 0.5,
316     #Here for single group: color and fill without aes()
317     color = "dodgerblue", fill = "dodgerblue",
318     #shut the legend
319     show.legend = FALSE
320   ) +
321   scale_y_continuous(expand = c(0.01, 0)) +
322   xlab("PL81") +
323   ylab("Density") +
324
325   RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
326     "sans")
327 print(.pl81)
328 rm(.df, .nbins)
329
330 require("ggplot2")
331 .df <- na.omit(data.frame(x = inc90))
332 .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
333 .c90<-ggplot(data = .df, aes(x = x, y = ..density..)) +
334   # Epanechnikov kernel and CV bandwidth
335   geom_density(
336     kernel = "gaussian",
337     bw = "ucv",
338     alpha = 0.5,
339     #Here for single group: color and fill without aes()
340     color = "gold", fill = "gold",
341     #shut the legend
342     show.legend = FALSE
343   ) +
344   scale_y_continuous(expand = c(0.01, 0)) +
345   xlab("C90") +
346   ylab("Density") +
347
348   RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
349     "sans")
350 print(.c90)
351 rm(.df, .nbins)
352
353 require("ggplot2")
354 .df <- na.omit(data.frame(x = inbc92))
355 .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
356 .bc92<-ggplot(data = .df, aes(x = x, y = ..density..)) +
357   # Epanechnikov kernel and CV bandwidth
358   geom_density(
359     kernel = "gaussian",
360     bw = "ucv",
361     alpha = 0.5,
362     #Here for single group: color and fill without aes()
363     color = "blueviolet", fill = "blueviolet",
364     #shut the legend

```

12 *Stochastic Frontier Analysis for Healthcare, with Illustrations in R*

```
363     show.legend = FALSE
364   ) +
365   scale_y_continuous(expand = c(0.01, 0)) +
366   xlab("BC92") +
367   ylab("Density") +
368
369   RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
370   "sans")
370 print(.bc92)
371 rm(.df, .nbins)
372
373 require("ggplot2")
374 .df <- na.omit(data.frame(x = ing05))
375 .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
376 .g05<-ggplot(data = .df, aes(x = x, y = ..density..)) +
377   # Epanechnikov kernel and CV bandwidth
378   geom_density(
379     kernel = "gaussian",
380     bw = "ucv",
381     alpha = 0.5,
382     #Here for single group: color and fill without aes()
383     color = "coral1", fill = "coral1",
384     #shut the legend
385     show.legend = FALSE
386   ) +
387   scale_y_continuous(expand = c(0.01, 0)) +
388   xlab("G05") +
389   ylab("Density") +
390
391   RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
392   "sans")
392 print(.g05)
393 rm(.df, .nbins)
394
395 require("ggplot2")
396 # Disable scientific notation
397 options(scipen=1000)
398
399 .df <- na.omit(data.frame(x = inklh14t))
400 .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
401 .klh14t<-ggplot(data = .df, aes(x = x, y = ..density..)) +
402   # Epanechnikov kernel and CV bandwidth
403   geom_density(
404     kernel = "gaussian",
405     bw = "ucv",
406     alpha = 0.5,
407     #Here for single group: color and fill without aes()
408     color = "darkolivegreen1", fill = "darkolivegreen1",
409     #shut the legend
410     show.legend = FALSE
411   ) +
412   scale_y_continuous(expand = c(0.01, 0)) +
413   xlab("KLH14 - Transitory") +
414   ylab("Density") +
415
416   RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
417   "sans")
417 print(.klh14t)
```

```

418  rm(.df, .nbins)
419
420  require("ggplot2")
421  .df <- na.omit(data.frame(x = inklh14p))
422  .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
423  .klh14p<-ggplot(data = .df, aes(x = x, y = ..density..)) +
424    # Epanechnikov kernel and CV bandwidth
425    geom_density(
426      kernel = "gaussian",
427      bw = "ucv",
428      alpha = 0.5,
429      #Here for single group: color and fill without aes()
430      color = "royalblue", fill = "royalblue",
431      #shut the legend
432      show.legend = FALSE
433    ) +
434    scale_y_continuous(expand = c(0.01, 0)) +
435    xlab("KLH14-Persistent") +
436    ylab("Density") +
437
438  RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
439    "sans")
440  print(.klh14p)
441  rm(.df, .nbins)
442
443  require("ggplot2")
444  .df <- na.omit(data.frame(x = inbc95))
445  .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
446  .bc95<-ggplot(data = .df, aes(x = x, y = ..density..)) +
447    # Epanechnikov kernel and CV bandwidth
448    geom_density(
449      kernel = "gaussian",
450      bw = "ucv",
451      alpha = 0.5,
452      #Here for single group: color and fill without aes()
453      color = "khaki3", fill = "khaki3",
454      #shut the legend
455      show.legend = FALSE
456    ) +
457    scale_y_continuous(expand = c(0.01, 0)) +
458    xlab("BC95") +
459    ylab("Density") +
460
461  RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
462    "sans")
463  print(.bc95)
464  rm(.df, .nbins)
465
466  require("ggplot2")
467  .df <- na.omit(data.frame(x = insvkz))
468  .nbins <- pretty(range(.df$x), n = nclass.FD(.df$x), min.n = 1)
469  .svkz<-ggplot(data = .df, aes(x = x, y = ..density..)) +
470    # Epanechnikov kernel and CV bandwidth
471    geom_density(
472      kernel = "gaussian",
473      bw = "ucv",
474      alpha = 0.5,
475      #Here for single group: color and fill without aes()

```

```
474   color = "mediumorchid1", fill = "mediumorchid1",
475   #shut the legend
476   show.legend = FALSE
477 ) +
478 scale_y_continuous(expand = c(0.01, 0)) +
479 xlab("SVKZ") +
480 ylab("Density") +
481
482 RcmdrPlugin.KMggplot2::theme_simple(base_size = 14, base_family =
483   "sans")
484 print(.svkz)
485 rm(.df, .nbins)
486
487 library("ggpubr")
488 Inefficiencies <- ggarrange(.als, .ss84, .pl81, .c90, .bc92, .g05,
489   .klh14t, .klh14p, .bc95, .svkz,
490   ncol = 2, nrow = 5)
491 Inefficiencies
492 #-----END-----
```