# Empirically Calibrated Simple SFC Models

Antoine Godin (Kingston University)

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# Calibration/Estimation

We will now briefly discuss the different ways to calibrate/estimate a model and then try the various solutions to a (extremely) simple model: PCEX.

#### Steady/stationary state calibration

Assuming you want to start with a model at the steady/stationary state, one way of calibrating these model is by doing "forward deduction":

- 1. Find in the literature (SFC or other) the values for the parameters you have in your model
- 2. Run the model, hoping it will converge towards the steady state
- 3. Fiddle as needed with some parameters such that the steady state meets your requirements

## Steady/stationary state calibration

Assuming you want to start with a model at the steady/stationary state, one way of calibrating these model is by doing "backward induction":

- 1. First you need to change your model equation so that it represent a steady state model
  - a. All lags are removed
  - b. All expectation are equal to the corresponding steady state values
- c. Some equation have to be changed completely: i.e. assuming a change in wealth function (V = $V_{-1} + YD - C$ , it becomes (C = YD) but then V is undefined. This implies that the consumption function  $(C = \alpha_1 Y D^e + \alpha_2 V_{-1})$  has to change as well so as to show the relationship between wealth and income  $(V = \frac{1-\alpha_1}{\alpha_2}YD)$ 2. Endogenise the parameters and exogenise some stock-flow norms or accepted values for specific indicators
- - a. Endogenising the parameters imply that instead of having the wealth to income function (V = $\frac{1-\alpha_1}{\alpha_2}YD$ ), we now specify  $\alpha_2$  as being determined by this equation:  $\alpha_2 = (1-\alpha_1)\frac{YD}{V}$ .
  - b. Exogenising the stock-flow norms or accepted indicators implies fixing debt-to-gdp ratios or unemployment rates to realistic values
- 3. Once you have enough equations for the number of parameters, and enough constraints (i.e. fixed values for stock-flow norms), the model should be determined and it is possible to solve the system and obtain both the parameters and initial values for the specific steady state. Note that you can use the PK-SFC package for that.

## Out-of-Steady State calibration/estimation: massaged data

one of the problem with SFC models is that they explicitly constrain the data into a specific data structure. I.e. the balance sheets and other transaction flow matrices are sparse while it is not the case in reality. This means that you will have to massage the data so that it meets your requirement. There is a lot of subjectivity happening there. A pseudo-algorithm for this part would look like this.

1. Individuate the time series you want to be exactly like the real world (e.g. GDP, unemployment, Household wealth)

- 2. Individuate the time series you are ok to play the buffer: these will have to be constructed from the accounting structure of the model. E.g. if you have households having two type of assets: money and bonds and you want households wealth and bond holding to be based on data, then money will have to be constructed as wealth-bonds.
- 3. Once you have a stock-flow consistent data set: run regressions to compute the parameters
  - a. I personally use OLS or dynamic OLS equation per equation
  - b. You could use 2-stage LS or minimum likelihood to estimate the system as a whole
  - c. Gennaro Zezza and others use co-integration methods
- 4. Initial conditions are determined by the data you have.
- 5. You need to say what you will do with residuals (coming out of your regressions): set them a 0, have them fixed at a non-zero value, have them decreasing through time?

#### Pure estimation

If you want to do a pure empirical model, then you have to start from the data and construct the model around it. You will need to look at how disaggregated the data is and what you can afford to do. A pseudo algorithm would look like this:

- 1. Aggregate/Disaggregate existing data sets such as it meets with the model structure you think is relevant or modify the model structure you want to have such that it fits with the data you have
- 2. Once you are happy with the data and the model structure, run regressions to compute the parameters
  - a. I personally use OLS or dynamic OLS equation per equation
  - b. You could use 2-stage LS or minimum likelihood to estimate the system as a whole
  - c. Gennaro Zezza and others use co-integration methods
- 3. Initial conditions are determined by the data you have.
- 4. You need to say what you will do with residuals (coming out of your regressions): set them a 0, have them fixed at a non-zero value, have them decreasing through time?

# Practical example

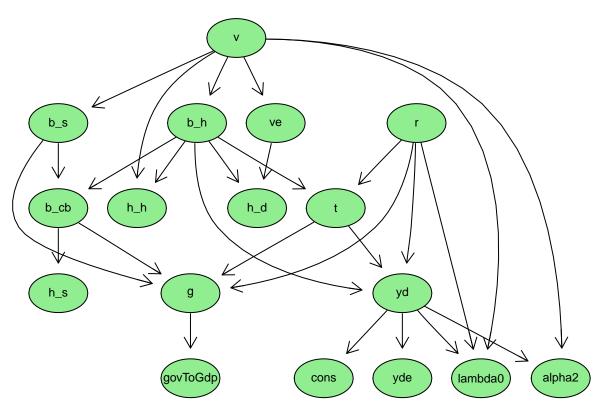
We will try to apply the middle two techniques (Backward induction and massaged data) to model PCEX to the UK economy.

#### **Backward** induction

First we need to create a steady state version of the model, this is done in file PCEXSS.txt.

```
# Loading the Steady State version of the model: you need to write down that file.
# Bear in mind that you dont need to have more than two periods in the steady state
modelSS<-sfc.model("Models/PCEXSS.txt")

plot.dag(modelSS)</pre>
```



The Steady State model is now ready to be simulated in order to obtain the calibration value.

```
# Solving the steady state
dataSS<-simulate(modelSS)
#Getting the value for all parameters and steady state values
calib<-as.data.frame(t(dataSS$baseline[2,]))
kable(calib)</pre>
```

У	yde	yd	t	cons	ve	alpha2	$\mathbf{v}$	h_h	h_d	lambda0	b_h	b_s	g	${\rm h}_{-}$
100	81.21	81.21	20.3025	81.21	110	0.0738273	110	34.375	34.375	0.7593273	75.625	110	18.79	34.3

What follows is a bit tricky but not too complicated to understand. What we will do is to create a temporary file called calibration.txt based on the values obtained via the steady state model. This file will then be merged into the PCEX\_forCalibration.txt file so as to dynamically create a complete model file (i.e. equations + calibration) which can then be simulated.

```
cat(paste("b_cb=",calib$b_cb,"\n"))
cat(paste("b_h=",calib$b_h,"\n"))
cat(paste("b_s=",calib$b_s,"\n"))
cat(paste("h_h=",calib$h_h,"\n"))
cat(paste("h_s=",calib$h_s,"\n"))
cat(paste("r=",calib$r,"\n"))
cat(paste("v=",calib$v,"\n"))
cat(paste("yd=",calib$yd,"\n"))
sink()
```

Now that the calibration text is created, we can load it and generate the automatic model.

We are now ready to look at the results of a baseline simulation.

```
#And load the model. The model is now ready to be simulated
calibratedModel<-sfc.model("Models/PCEX_Auto.sfc")

dataSim<-as.data.frame(simulate(calibratedModel)$baseline)

#Checking that we have a steady state model
matplot(calibratedModel$time,dataSim[,c("y","yd","cons")],type='l',lty=1,lwd=2,yab="",xlab="")

## Warning in plot.window(...): "yab" is not a graphical parameter

## Warning in plot.xy(xy, type, ...): "yab" is not a graphical parameter

## Warning in axis(side = side, at = at, labels = labels, ...): "yab" is not a

## graphical parameter

## Warning in axis(side = side, at = at, labels = labels, ...): "yab" is not a

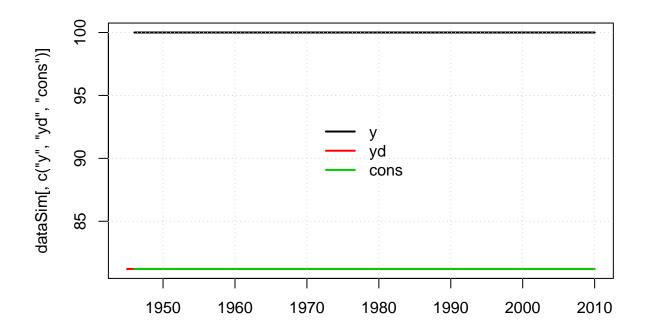
## graphical parameter

## Warning in box(...): "yab" is not a graphical parameter

## Warning in title(...): "yab" is not a graphical parameter

## Warning in title(...): "yab" is not a graphical parameter

## Warning in title(...): "yab" is not a graphical parameter</pre>
```



#### Massaged data

```
library(PKSFC)
#1. fetch all the relevant data sets for your case, using pdfetch
#qetting the data in a raw format
NF_TR_raw = pdfetch_EUROSTAT("nasa_10_nf_tr", UNIT="CP_MNAC", NA_ITEM=c("..."),
                                             SECTOR=c("..."), DIRECT = "...")
#Transforming the data into a data.frame (easy to access and to view)
NF TR<-as.data.frame(NF TR raw)</pre>
#2. Construct the various massaged time series
NF_TR$MASSAGED<-NF_TR[,"..."]+NF_TR[,"..."]-NF_TR[,"..."]
#3. Run the regression
model.cons < -lm(CONS[-1] \sim YD[-19] + V[-19] - 1, data = NF TR)
calib$alpha1<-coef(model.cons)[1]</pre>
calib$alpha2<-coef(model.cons)[2]</pre>
. . .
#What follows is a bit tricky but not too complicated to understand. What we will do is
# to create a temporary file called calibration.txt based on the values obtained via the
# steady state model. This file will then be merged into the PCEX_forCalibration.txt file
# so as to dynamically create a complete model file (i.e. equations + calibration) which
# can then be simulated
#Creation of the calibration.txt file
sink("calibration.txt")
###### PARAMETERS
cat(paste("alpha1=",calib$alpha1,"\n"))
cat(paste("alpha2=",calib$alpha2,"\n"))
cat(paste("lambda0=",calib$lambda0,"\n"))
```

```
cat(paste("lambda1=",calib$lambda1,"\n"))
cat(paste("lambda2=",calib$lambda2,"\n"))
cat(paste("theta=",calib$theta,"\n"))
###### EXOGENOUS
cat(paste("g=",calib$g,"\n"))
cat(paste("r_bar=",calib$r_bar,"\n"))
###### Starting values for stocks
cat(paste("b cb=",calib$b cb,"\n"))
cat(paste("b_h=",calib$b_h,"\n"))
cat(paste("b_s=",calib$b_s,"\n"))
cat(paste("h_h=",calib$h_h,"\n"))
cat(paste("h_s=",calib$h_s,"\n"))
cat(paste("r=",calib$r,"\n"))
cat(paste("v=",calib$v,"\n"))
cat(paste("yd=",calib$yd,"\n"))
sink()
# Now we will read the calibration file
calibrationLines<-readLines("Models/calibration.txt")</pre>
#Next step is to read the model without calibration
modelLines<-readLines("Models/PCEX_forCalibration")</pre>
#We need to determine where to add the calibration lines, this is done where
# "{CALIBRATION}" has been inserted in the PCEX forCalibration.txt file
indexCalib<-grep("CALIBRATION", modelLines)</pre>
#Now we can merge the two files into a complete model
totModel<-c(modelLines[1:(indexCalib-1)],calibrationLines,</pre>
                    modelLines[(indexCalib+1):length(modelLines)])
#Write down the model into a txt file
writeLines(totModel, "Models/PCEX_Auto.sfc")
#And load the model. The model is now ready to be simulated
calibratedModel<-sfc.model("Models/PCEX_Auto.sfc")</pre>
```