

Week 17 Exercises: SDA (April 26th 2022)

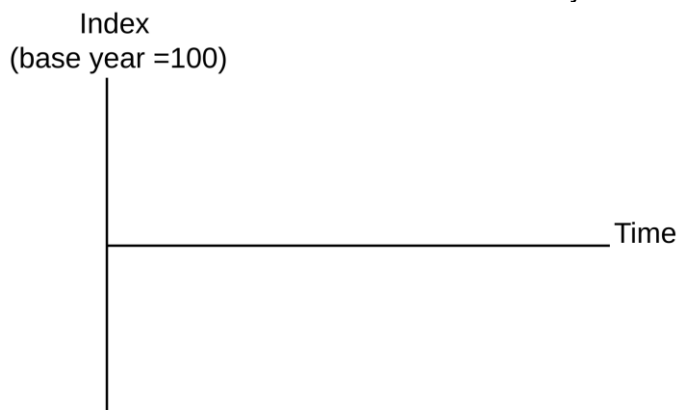
Objectives

- Interpret SDA results
- Construct SDA in Python
- Implement SDA in MRIO

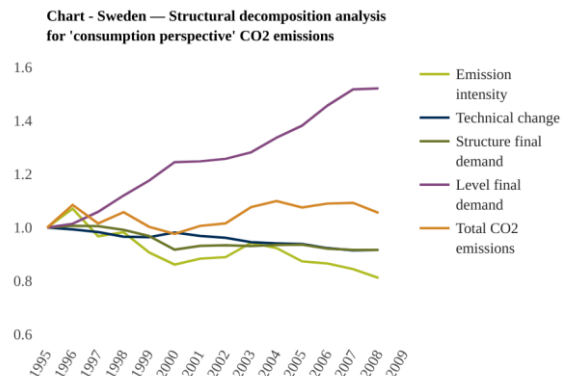
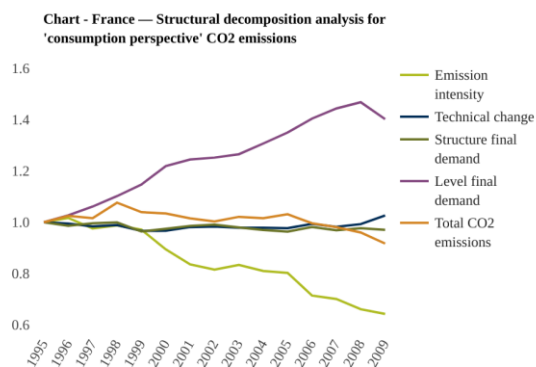
Conceptual exercises

Part 1: Understanding absolute and relative decoupling

Draw a graph depicting i) relative decoupling and ii) absolute decoupling of economic activity and resource use. The y-axis of the graph can represent an index of economic activity and resource use where 100 indicates the base year. The x-axis can represent time.



Part 2: Interpreting SDA results



- Between 2002 and 2003 what was the relative change in consumption-based CO2 emissions in France and Sweden
- Which factor(s) were responsible for these observed trends
- To what extent is relative and absolute decoupling of emissions and economic growth achieved in both countries?
- Identify three other factors not listed above that may influence the emissions footprint of economic activity and could be incorporated into SDA
- How might rebound effects offset the emissions savings of energy efficiency gains?

Python exercises

Part 1 (no python code)

Write a structural decomposition analysis Python program that recreates the values in Table 13.1 of Miller & Blair (page 593-602). The table and the data needed are shown below.

$$\mathbf{Z}^0 = \begin{bmatrix} 10 & 20 & 25 \\ 15 & 5 & 30 \\ 30 & 40 & 5 \end{bmatrix}, \quad \mathbf{f}^0 = \begin{bmatrix} 45 \\ 30 \\ 25 \end{bmatrix}, \quad \mathbf{Z}^1 = \begin{bmatrix} 12 & 15 & 35 \\ 24 & 11 & 30 \\ 36 & 50 & 8 \end{bmatrix}, \quad \mathbf{f}^1 = \begin{bmatrix} 50 \\ 35 \\ 26 \end{bmatrix}$$

Table 13.1 Alternative Structural Decompositions

	Technology Change Contribution	Final-Demand Change Contribution	Interaction Term
Equation (13.3)	$\begin{bmatrix} 0.90 \\ 8.62 \\ 9.01 \end{bmatrix}$	$\begin{bmatrix} 11.10 \\ 11.38 \\ 10.99 \end{bmatrix}$	$\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$
Equation (13.4)	$\begin{bmatrix} 0.78 \\ 9.66 \\ 9.96 \end{bmatrix}$	$\begin{bmatrix} 11.22 \\ 10.34 \\ 10.04 \end{bmatrix}$	$\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$
Equation (13.5)	$\begin{bmatrix} 0.90 \\ 8.62 \\ 9.01 \end{bmatrix}$	$\begin{bmatrix} 11.22 \\ 10.34 \\ 10.04 \end{bmatrix}$	$+\begin{bmatrix} -0.12 \\ 1.04 \\ 0.95 \end{bmatrix}$
Equation (13.6)	$\begin{bmatrix} 0.78 \\ 9.66 \\ 9.96 \end{bmatrix}$	$\begin{bmatrix} 11.10 \\ 11.38 \\ 10.99 \end{bmatrix}$	$-\begin{bmatrix} -0.12 \\ 1.04 \\ 0.95 \end{bmatrix}$
Equation (13.7)	$\begin{bmatrix} 0.84 \\ 9.14 \\ 9.49 \end{bmatrix}$	$\begin{bmatrix} 11.16 \\ 10.86 \\ 10.51 \end{bmatrix}$	$\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$

$$\Delta \mathbf{x} = \mathbf{L}^1 (\mathbf{f}^0 + \Delta \mathbf{f}) - (\mathbf{L}^1 - \Delta \mathbf{L}) \mathbf{f}^0 = (\Delta \mathbf{L}) \mathbf{f}^0 + \mathbf{L}^1 (\Delta \mathbf{f}) \quad (13.3)$$

$$\Delta \mathbf{x} = (\mathbf{L}^0 + \Delta \mathbf{L}) \mathbf{f}^1 - \mathbf{L}^0 (\mathbf{f}^1 - \Delta \mathbf{f}) = (\Delta \mathbf{L}) \mathbf{f}^1 + \mathbf{L}^0 (\Delta \mathbf{f}) \quad (13.4)$$

$$\Delta \mathbf{x} = (\mathbf{L}^0 + \Delta \mathbf{L}) (\mathbf{f}^0 + \Delta \mathbf{f}) - \mathbf{L}^0 \mathbf{f}^0 = (\Delta \mathbf{L}) \mathbf{f}^0 + \mathbf{L}^0 (\Delta \mathbf{f}) + (\Delta \mathbf{L}) (\Delta \mathbf{f}) \quad (13.5)$$

$$\Delta \mathbf{x} = \mathbf{L}^1 \mathbf{f}^1 - (\mathbf{L}^1 - \Delta \mathbf{L}) (\mathbf{f}^1 - \Delta \mathbf{f}) = (\Delta \mathbf{L}) \mathbf{f}^1 + \mathbf{L}^1 (\Delta \mathbf{f}) - (\Delta \mathbf{L}) (\Delta \mathbf{f}) \quad (13.6)$$

$$\Delta \mathbf{x} = (1/2)(\Delta \mathbf{L})(\mathbf{f}^0 + \mathbf{f}^1) + (1/2)(\mathbf{L}^0 + \mathbf{L}^1)(\Delta \mathbf{f}) \quad (13.7)$$

Python exercises

Part 1 (with python code)

Write a structural decomposition analysis Python program that recreates the values in Table 13.1 of Miller & Blair (page 593-602). The table and the data needed are shown above.

```
###
# Input data
# t=0 transaction matrix and final demand vector
Z_0 = np.array([
    [10, 20, 25],
    [15, 5, 30],
    [30, 40, 5]])

f_0 = np.array([
    [45],
    [30],
    [25]])

# t=1 transaction matrix and final demand vector
Z_1 = np.array([
    [12, 15, 35],
    [24, 11, 30],
    [36, 50, 8]])

f_1 = np.array([
    [50],
    [35],
    [26]])

###
# Calculate total requirements matrix
x_0 = Z_0.sum(axis=1, keepdims=True) + f_0
A_0 = Z_0 @ invdiag(x_0[:, 0])
L_0 = np.linalg.inv(np.eye(3) - A_0)

x_1 = Z_1.sum(axis=1, keepdims=True) + f_1
A_1 = Z_1 @ invdiag(x_1[:, 0])
L_1 = np.linalg.inv(np.eye(3) - A_1)

# All input data for our SDA is now complete.
# Start decomposing the effect of technology change and final demand change on the
# total output. A two factor decomposition.
```

```

%%
# change in the total requirements matrix
L_delta = L_1 - L_0
# change in the final demand
f_delta = f_1 - f_0
# change in total output - used as check
x_delta = x_1 - x_0

%%
# decomposition (eq 13.3)
tech_change = L_delta.dot(f_0)
fd_change = L_1.dot(f_delta)

# check if our result is complete
tot_change = tech_change + fd_change
check = x_delta - tot_change
check_decomposition(check)

result = {"13.3": [tech_change, fd_change]}

print(f"\n13.3:\ntech_change: {tech_change}\nfd_change: {fd_change}")

%%
# decomposition (eq 13.4)
tech_change = np.dot(L_delta, f_1)
fd_change = np.dot(L_0, f_delta)

# check if our result is complete
tot_change = tech_change + fd_change
check = x_delta - tot_change
check_decomposition(check)

result["13.4"] = [tech_change, fd_change]

print(f"\n13.4:\ntech_change: {tech_change}\nfd_change: {fd_change}")

%%
# decomposition (eq 13.5)
tech_change = np.dot(L_delta, f_0)
fd_change = np.dot(L_0, f_delta)
interaction = np.dot(L_delta, f_delta)

# check if our result is complete
tot_change = tech_change + fd_change + interaction
check = x_delta - tot_change
check_decomposition(check)

result["13.5"] = [tech_change, fd_change, interaction]

print(

```

```

(f"\n13.5:\ntech_change: {tech_change}\nfd_change: {fd_change}"
 f"\ninteraction: {interaction}")

###
# decomposition (eq 13.6)
tech_change = np.dot(L_delta, f_1)
fd_change = np.dot(L_1, f_delta)
interaction = np.dot(L_delta, f_delta)

# check if our result is complete
tot_change = tech_change + fd_change - interaction
check = x_delta - tot_change
check_decomposition(check)

result["13.6"] = [tech_change, fd_change, interaction]

print((f"\n13.6:\ntech_change: {tech_change}\nfd_change: {fd_change}"
 f"\ninteraction: {interaction}")

###
# decomposition (eq 13.7)
tech_change = 0.5 * np.dot(L_delta, f_0 + f_1)
fd_change = 0.5 * np.dot(L_0 + L_1, f_delta)

# check if our result is complete
tot_change = tech_change + fd_change
check = x_delta - tot_change
check_decomposition(check)

result["13.7"] = [tech_change, fd_change]

print(f"\n13.7:\ntech_change: {tech_change}\nfd_change: {fd_change}")

```