

Step 1: Setup the working environment

- Rstudio uses a default directory (folder) to access the data it needs. In this initial step you will tell Rstudio where it can find the data to run the model.
- This can be done using the `setwd()` function.
- This function takes as an input the direction where the data is stored in your computer and it changes the working directory to that direction.

```
#A. Setup the working environment
setwd("Copy here the Path to the <Demo> folder on your computer")
rm(list = ls())
```

Step 2: Install (or load) the IGC.CSM package

- You need to install the IGC.CSM package if it the first time you use it. This can be done using the function `install.packages()` function.
- Once you have installed it, you need to load it every time you want to use it. This can be done using the function `library()`.

```
#B. Install the package from the CRAN repository  
install.packages('IGC.CSM')  
#library('IGC.CSM')
```

Step 3: Read the data

- The next step consists in reading the data. This can be done using the function `read.csv()`. Notice that you must read both data files.
- Once the data is read, you must save every characteristic as a variable. In this way, you can call them independently.

```
#C. Upload the required data
data_chars = read.csv("1. Data for model/1. Chars.csv")
data_times = read.csv("1. Data for model/2. TravelTimes.csv")

L_j = as.data.frame(data_chars$L_j)           # Number of workers in each location
L_i = as.data.frame(data_chars$L_i)           # Number of residents in each location
L_i = L_i*sum(L_j)/sum(L_i)                   # Normalize the number of residents in each location

K = as.data.frame(data_chars$K)               # Area in sq. km from each location
Q = as.data.frame(data_chars$Q)               # Average floorspace price in each location

N = dim(L_i)[1]                               # Define the number of locations in the model
t_ij = as.matrix(data_times[,2:(N+1)], dim=c(N,N)) # Use the travel times matrix without the index column
```

Step 4: Invert the model

- Next, you will use the function `inversionModel()`.
- This function takes as input the data you read in the previous step and it returns the unobserved characteristics of the city (amenities, productivities, wages and density of development).

```
#D. Invert the model
inversion_m_b1 = inversionModel(N=N,
                                L_i=L_i,
                                L_j=L_j,
                                Q=Q,
                                K=K,
                                t_ij=t_ij)
```

Step 5: Solve the model

- Once you know both the observed and unobserved characteristics of the city, you can solve the model using the function `solveModel()`.
- This function takes as input the observed and unobserved characteristics and it returns features of the city such as the share of residents who commute i to j or the expected utility of living in the city.

```
#E. Solve model
results_m_b1 = solveModel(N=N,
                           L_i=L_i,
                           L_j=L_j,
                           varphi=inversion_m_b1$varphi,
                           t_ij=t_ij,
                           K=K,
                           a=inversion_m_b1$a,
                           b=inversion_m_b1$b,
                           w_eq=inversion_m_b1$w,
                           u_eq=inversion_m_b1$u,
                           Q_eq=inversion_m_b1$Q_norm,
                           ttheta_eq=inversion_m_b1$ttheta)
```

Step 6: Run counterfactuals

- The function `solveModel()` can also be used to run counterfactuals.
- The idea is to load new values for either the observed or unobserved variables to assess how the remaining variables change.
- In the code template, we input a new commuting time matrix as a counterfactual scenario.

```
#F. Solve the model after changes in travel times

data_times_policy = read.csv("1. Data for model/3. TravelTimes_Policy.csv")

# Use the travel times matrix without the index column
t_ij_policy = as.matrix(data_times_policy[,2:(N+1)], dim=c(N,N))

results_m_trans = solveModel(N=N,
                             L_i=L_i,
                             L_j=L_j,
                             varphi=inversion_m_b1$varphi,
                             t_ij=t_ij_policy,
                             K=K,
                             a=inversion_m_b1$a,
                             b=inversion_m_b1$b,
                             w_eq=inversion_m_b1$w,
                             u_eq=inversion_m_b1$u,
                             Q_eq=inversion_m_b1$Q_norm,
                             ttheta_eq=inversion_m_b1$ttheta)
```