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2 `Main_models_1_3.m`

Notes:

- Main Matlab file: "GMMDiffusion/Main_models_1_3" for model without λ (endorsement effects) and "GMMDiffusion/Main_models_2_4" for model with λ

3 `Main_models_1_3.m`

Part 0

- 26: 2 model types: `modelType == 1` for $q_N = q_P$ and `== 3` for $q_N \neq q_P$

Part 1

- 29: Picking 43 BSS villages

Part 2

- 33: Different versions of moments (In the paper, they say they use 6 moments, but no version says 6 moments, why?)

Part 3

- 49: **TMonths**: number of months in data of each village
- 57: **T**: number of trimesters calculated from **TMonths**. *ceil*(*X*) rounds each element of *X* to the nearest integer greater than or equal to that element, and *./* does element-wise right division.
- 59: This line checks whether the number of elements of **T** is equal to the number of graphs. has same *assert(cond)* throws an error if *cond* is false. *numel*(*A*) returns the number of elements, *n*, in array *A*.

Part 4

- 63-69: Parameter grids: although explained in detail in the Supplementary Materials, not shown how they did it. Also, grid values differ from the supplementary material.

Part 5

- 90: Load the network data, which contains adjacency matrices of 43 villages
- 91: What and why do they do this here? (Element-wise $X \cdot X$) (Sorry for a silly questions.)
- 95-: For each village,
 - 101: Load the leader data: **ID** and dummy of being a leader or not (only use the latter)
 - 106: Load the take-up data: Only dummy (maybe the same order as the one above?)
 - 107: **EmpRate** is the proportion of non-leaders who took up MF (\sim means what?)
 - 110: **inGiant**??? Maybe the biggest cluster in the village? (Yes, belong to largest component or not.)
 - 113-114: **d** = degree of network, **hermits** = households without any connections with other households (= isolates). *sum*(*A*, 2) is a column vector containing the sum of each row of matrix *A* (2 is the dimension).
 - 117: Load the covariates of the households (not sure which column corresponds to which variable, maybe in readme file or somewhere?) (6 covariates, in the order of number of rooms in a household, number of beds, whether the household has private/government/no electricity, whether the household has own/common/no latrine, number of rooms per capita and number of beds per capita according to Readme file)
 - 120: Select covariates to use (but apparently in the end they used all the covariates in the data)
 - 123-130: Restrict the sample to households belonging to the giant networks and households of leaders (number of rows will be reduced). *logical*(*A*) converts *A* into an array of logical values. In the subsequent simulations, the authors seem to focus only on the giant component.
 - * 123: Leaders
 - * 124: Take-up
 - * 125: Covariates
 - * 127: Take-up by leaders
 - * 128: Covariates of leaders
 - * 129: Outcome = Take-up by leaders
 - * 130: Covars = Covariates of leaders
 - **Outcome** and **Covars** will be used to estimate β
 - 133-137: Second neighbors (**sec**) = matrix indicating households whose distance between them is 2. Here, X^2 means $X * X$. (See [here](#))

- * 133: > 0 turns the elements of the matrix into binary number (0 or 1)
- * 134-136: Ruling out households themselves (with two steps, they can always come back to themselves)
- * 137: Ruling out households with distance 1 (think about a triangle)

Part 6

- 142-143: Logistic regression by using leaders to estimate β (vector of 6 coefficient estimates)

Part 7

- 147-160: Calculate the moments in the case where $q_N = q_P$
 - 153-160: Calculate moments at each grid of $q_N (= q_P)$
 - * 154: `theta` is the choice of parameters q_N and q_P
 - * 155: `tic` starts stopwatch timer (`toc` ends timer) to measure elapsed time
 - * 156: Call the function `divergence_model` which calculates moments
- 163-178: Calculate moments in the case where $q_N \neq q_P$
 - 169-174: Calculate moments at each grid of (q_N, q_P)
 - * 172: Call the function `divergence_model` which calculates moments
- Note: In the calculations here, a matrix D is obtained. This is a $\text{length}(q_N \text{ grids})$ -by- $\text{length}(q_P \text{ grids})$ matrix, and each element is a $(\# \text{ villages})$ -by- $(\# \text{ moments})$ matrix.
- 181-183: Save the computed data in the mat format.

Part 8

- 188-194: Select if just obtaining point estimates or standard errors through bootstrap
 - `bootstrap == 0`: point estimates
 - `bootstrap == 1`: standard errors through bootstrap (1000 times)
- 197-207: Select if two step optimal weights are used or not
 - `twoStepOptimal == 1`:
 - * 202: With some guess of $q_N (= 0.09)$ and $q_P (= 0.45)$, obtain moments by the function `divergence_model`
 - * 203: Calculate the outer product of the moments, divided by the number of villages
 - * 204: Take the inverse of the above expression, which will be the weight in the second step
 - * **Note:** I guess usually people (i) solve the GMM problem with an identity matrix as a weight to get the consistent estimate of parameters, and (ii) get the weight in the way described above. Here, somehow authors set q_N and q_P and calculate the weight based on these parameters. If these are not derived by consistent estimates, then the weight is inconsistent and thus the estimates in the second step will be biased as well. (q_N and q_P seems to set by the first-stage estimate `theta` via the same algorithm with identity matrix weighting according to the Supplementary Materials)
 - `twoStepOptimal == 0`: use an identity matrix as the weight in the “second” step (`eye` function)

If these are not derived by consistent estimates, then the weight is inconsistent and thus the estimates in the second step will be biased as well.

- 211-216: Obtain a new 4-dimension (length(q_N grids)-by-length(q_P grids)-by-(# villages)-by-(# moments)) D_{new} from a nested matrix D
- 223-259: Bootstrap (note: here in the bootstrap, randomly generated village weights are used, not that villages are randomly chosen with replacement)
 - 226-231: Generate bootstrap weights
 - * bootstrap == 0 (ie. for point estimate): same weights for all villages
 - * bootstrap == 1 (ie. for bootstrap standard errors): randomly generated weights follow exponential distribution (normalized so that the sum of weights is 1)
 - 236-244: Calculate the criterion function for each (q_N, q_P) grid
 - * 240: Calculate weighted moments
 - * 242: Calculate the criterion function
 - 246-254: Derive the parameter values that minimize the objective function and store the results

4 Main_models_2_4.m

The same structure as `Main_models_1_3.m`, but with λ (endorsement effect).

- 66: The grid for λ is determined.
- 68-73: The grid for q_P, q_N are set here. What's the logic behind changing these grid based on whether to have λ ?

5 divergence_model.m

"This computes the deviation of the empirical moments from the simulated ones"

Arguments

- X: adjacency matrix
- Z: covariate
- Betas: estimates of β 's
- leaders: vector indicating leaders
- TakeUp: vector indicating who took up microfinance
- Sec: matrix indicating second neighbors (= households with distance two)
- theta: parameter values (q_N, q_P)
- m: number of moments
- S: number of simulations
- T: number of trimesters (or months/quarters)
- EmpRate: proportion of non-leaders who took up MF
- version: specification of which moments to use

Main part

- 14: Calculate the number of villages (G)
- 23-39: Calculate empirical and simulated moments
 - 25: Calculate empirical moments, using a function `moments`
 - 28-32: Calculate simulated moments
 - * 30: Simulate the take-up of microfinance, using a function `diffusion_model`
 - * 31: Calculate simulated moments based on the simulated take-ups, using a function `moments`
 - 35: Calculate the mean of simulated moments
 - 36: Take the difference of empirical and simulated moments

Outputs

- D: Calculated deviation (difference of empirical and simulated moments)
- TimeSim: Just `zeros(G,S)`? Not used later?

6 `moments.m`

Arguments

- X: adjacency matrix
- leaders: vector indicating leaders
- infected: vector indicating who took up microfinance
- Sec: matrix indicating second neighbors (= households with distance two)
- j: village index
- version: specification of which moments to be used

Main part

- 3: Declare a variable “netstats” which contains information of networks in each village (for the function `persistent`, see [here](#))
- 5: `size(X,1)` returns the length of 1st dimension, which is 43 (villages).
- 7-49: If a variable “netstats” is already defined, then almost skip the parts, but if not yet, then calculate each statistics of the networks (7-43)
 - 8: Use a function `breadthdistRAL` to compute a reachability matrix (R) and distance matrix (D). D is the distance between each leader and household which is calculated with a function, `breadth.m`: By inputting `leaders`, the D matrix include distances from leaders.
 - * I think this function calculates the distance between everyone in the data set. Not too sure.
 - 10-11: Calculate the minimum and average distances from leaders
 - 13-22: Calculate the minimum distance from infected and non-infected leaders
 - * `.*` does element-wise multiplication.

- * *infected*. * *leaders* is a vector with elements that take the value one for infected leaders (leaders who took up microfinance).
- * $(1 - \textit{infected})$. * *leaders* is a vector with elements that take the value one for non-infected leaders.
- * 19: Detect the size of each row in the connection/adjacency matrix (CIJ) which is really **X**.
- 22-24: I'm not sure if I understand this color stuff.
- 24-27: Store calculated distances (matrices) in **netstats**
- 29-31: Display the size of stored matrices with village index.
- 32-34: Display matrices with binary elements which takes the value one when minimum distances is equal to 1 with village index.
 - * 34: **source** is the index of the leader dummy being 1.
- 36-37: Calculate whether each household is neighbor to infected and non-infected leaders
- 39: Calculate the degree
- 40- : Measure the distance
 - * 40: Calculate total number of edges
 - * 41: ??
 - * 42: Calculate total number of leaders
 - * 43: Calculate the total number of leaders' edges (*leaders' * X* is a vector of degrees of leaders, and **leaders* adds up those degrees)
- 52-: Moments under different versions
 - 53-88: Case1 (moments (2)-(6) in the paper):
 - * 54-62: “Fraction of nodes that have no taking neighbors but are takers themselves” (moment (2) in the paper)
 - 56: Number of infected neighbors:
`ones(N,1)*infected'`: a matrix with indicators of infected households in each row (same row vector of infected households repeated n times)
`ones(N,1)*infected'`: a matrix with indicators of infected households in each row
 $\Rightarrow (\text{ones}(N,1)*\text{infected}') .* X$: a matrix with indicators of infected neighbors for each household
 $\Rightarrow \text{sum}((\text{ones}(N,1)*\text{infected}') .* X, 2)$: number of infected neighbors for each household
 - 58-59: If there is any household who are linked with other households (`netstats(j).degree > 0`) but has no infected neighbors (`infected Neighbors == 0`), calculate the fraction $\frac{\# \text{ HH who are linked with other households, don't have any infected neighbors, but are infected themselves}}{\# \text{ HH who are linked with other households and don't have any infected neighbors}}$
 - 60-61: If there is no such household in the village, then just let the moment be 0 (since the denominator of the above fraction will be 0 in this case)
 - * 64-69: “Fraction of individuals that are infected in the neighborhood of infected leaders `stats(1) = 0`” (moment (3) in the paper)
 - If `sum(netstats(j).neighborOfInfected) > 0` (there is at least one household who is neighbor to infected leaders), calculate the fraction $\frac{\# \text{ HH who are infected in the neighborhood of infected leaders}}{\# \text{ HH who are in the neighborhood of infected leaders}}$
 - Otherwise, just let the moment be 0 (since the denominator of the above fraction will be 0 in this case)
 - * 71-76: “Fraction of individuals that are infected in the neighborhood of non-infected leaders” (moment (4) in the paper)

- If `sum(netstats(j).neighborOfNonInfected) > 0` (there is at least one household who is neighbor to non-infected leaders), calculate the fraction
$$\frac{\# \text{ HH who are infected in the neighborhood of non-infected leaders}}{\# \text{ HH who are in the neighborhood of non-infected leaders}}$$
- Otherwise, just let the moment be 0 (since the denominator of the above fraction will be 0 in this case)
- * 78-82: “Covariance of individuals taking with share of neighbors taking” (moment (5) in the paper)
 - 79: `NonHermits`: Indicator of non-isolated households
 - 80: `ShareofTakingNeighbors`: For each non-isolated household,
$$\frac{\# \text{ infected neighbors}}{\# \text{ neighbors}}$$
 - 81: `NonHermitTakers`: Indicator of non-isolated households who took up
 - 82: `moment` =
$$\frac{\sum_{i:\text{non-isolated}} (\text{Take-up})_i \times (\text{Share of taking neighbors})_i}{\# \text{ Non-isolated households}}$$
 - Covariance without subtracting means?
- * 82-88: “Covariance of individuals taking with share of second neighbors taking” (moment (6) in the paper)
 - 86: `infectedSecond`: number of infected second neighbors
 - 87: `ShareofSecond`: Share of infected second neighbors for non-isolated households,
$$\frac{\# \text{ infected second neighbors}}{\# \text{ first neighbors}}$$
 (** Why is the denominator about first neighbors? **)
 - 88: `moment` =
$$\frac{\sum_{i:\text{non-isolated}} (\text{Take-up})_i \times (\text{Share of taking second neighbors})_i}{\# \text{ Non-isolated households}}$$
- 91-112: Case2 (moments (2), (5), and (6) in the paper):
- 92-100: “Fraction of nodes that have no taking neighbors but are takers themselves” (moment (2) in the paper)
- 102-106: “Covariance of individuals taking with share of neighbors taking” (moment (5) in the paper)
- 110-112: “Covariance of individuals taking with share of second neighbors taking” (moment (6) in the paper)
- 115-141: Case3 (moments (2), (5), and (6) in the paper, same as case 2, but purged of leader injection points):
- 117: a variable that denotes whether a node is a leader
- 143-169: Case4 (moments (2), (5), and (6) in the paper, same as case 2, but purged of all leader points):
- 117: a variable that denotes whether a node is a leader
- 166-167: Moment (6), but second neighbors who are leaders are not taken into account (see line 167)

Outputs

- `stats`: a vector of m calculated moments for the village j

7 breadthdistRAL.m

The source of this function is the Brain Connectivity Toolbox (brain-connectivity-toolbox.net) which is a MATLAB toolbox for complex-network analysis of structural and functional brain-connectivity data sets. (See [here](#)). It seems like the authors modified `breadthdist.m` to add `dummies` in the arguments. `breadthdist.m` is slower but less memory intensive than “`reachdist.m`”.

Arguments

- CIJ: connection/adjacency matrix
- dummies: Dummy vector of length N for which you wish to compute the distance and reachability matrices.

Outputs

- R: reachability matrix, which describes reachability between all pairs of nodes. An entry (u,v)=1 means that there exists a path from node u to node v; alternatively (u,v)=0.
- D: distance matrix, which contains lengths of shortest paths between all pairs of nodes. An entry (u,v) represents the length of shortest path from node u to node v. The average shortest path length is the characteristic path length of the network.

8 diffusion_model.m

Arguments

- parms: parameters (q_N, q_P)
- Z: covariate
- Betas: estimates of β 's
- X: adjacency matrix
- leaders: vector indicating leaders
- j: village index
- T: number of months
- EmpRate: proportion of non-leaders who took up MF

Main part

- 7-11: Prepare arrays to be updated in the simulation below
 - 7: **infected**: people who are already infected and newly infected (initial value = false for everyone)
 - 8: **infectedbefore**: people who are already infected (initial value = false for everyone)
 - 9: **contagiousbefore**: people who are already informed (initial value = false for everyone)
 - 10: **contagious**: people who are newly informed and already informed (?) (initial value = true only for leaders)
 - 11: **dynamicInfection**: vector that tracks the infection rate for the number of periods it takes place
- 16-39: Loop to simulate the diffusion for T periods
 - 20-27: Step 1: Take-up decision based on newly informed
 - * 21: Probability of take-up (calculated for everyone, both non-infected and already-infected)
 - * 22: Updated infected households:
The condition ($\sim\text{contagiousbefore} \ \& \ \text{contagious} \ \& \ \mathbf{x(:,t)} < \text{LOGITprob}$) is satisfied if:

- not contagious before, and
 - newly informed, and
 - the random number generated is smaller than the LOGITprob calculated in line 21.
 - * 25: Update **infectedbefore**
 - * 26: Update **contagiousbefore**
 - * 27: **C**: number of informed households
 - 29-31: Step 2: Information flows
 - * 30: **transmitPROB**: Vector of probability of transmission (individual specific, depending on informed and on take-up)
 - If informed & take-up, probability = q_P
 - If informed & not take-up, probability = q_N
 - * 31: **contagionlikelihood**: The probability to information flowing from informed households to neighbors (no matter whether those neighbors are already informed or not)
- eg) Suppose that $X = \begin{pmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \end{pmatrix}$, $\text{transmitPROB} = \begin{pmatrix} q_P \\ 0 \\ 0 \\ q_N \end{pmatrix}$, and $\text{contagious} = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix}$.
- Then,

$$\begin{aligned}
 & X(\text{contagious}, :) . * (\text{transmitPROB}(\text{contagious}) * \text{ones}(1, N)) \\
 &= \begin{pmatrix} 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 \end{pmatrix} . * \begin{pmatrix} q_P & q_P & q_P & q_P \\ q_N & q_N & q_N & q_N \end{pmatrix} \\
 &= \begin{pmatrix} 0 & q_P & q_P & q_P \\ q_N & q_N & 0 & 0 \end{pmatrix}
 \end{aligned}$$

- 33-36: Step 3: Simulate newly informed households
 - * 34: Update **contagious**:
 - $(\text{contagionlikelihood} > \text{rand}(C, N))$: simulation of whether households are informed from already informed neighbors
 - $((\text{contagionlikelihood} > \text{rand}(C, N))' * \text{ones}(C, 1) > 0)$: informed at least from one neighbor
 - * 36: Update **dynamicInfection** to keep track of the fraction of take-up households