Matlab files for Pareto Extrapolation

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1 Introduction

This note explains the functionalities in the MATLAB package PE, which implements the Pareto Extrapolation algorithm. The user should use these files at their own responsibility. Whenever you use these codes for your research, please cite Gouin-Bonenfant and Toda (2018).

2 Package content

There are three main functionalities for Pareto extrapolation:

- getZeta.m
- getQ.m
- getTopShares.m

In addition, example.m contains a simple example.

2.1 Computing Pareto exponent

getZeta.m computes the Pareto exponent using the Beare and Toda (2017) formula. The usage is

where

• PS is the $S \times S$ transition probability matrix of exogenous states indexed by $s = 1, \dots, S$,

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- PJ is the $S^2 \times J$ matrix of conditional probabilities of transitory states indexed by $j = 1, \ldots, J$,
- p is the birth/death probability $p \in [0, 1)$,
- G is the $S^2 \times J$ matrix of gross growth rates, and
- zetaBound is a vector $(\underline{\zeta}, \overline{\zeta})$ that specifies the lower and upper bounds to search for the Pareto exponent (optional).

The S^2 rows in PJ and G should be ordered such that

$$(s, s') = (1, 1), \dots, (1, S); \dots; (s, 1), \dots, (s, S); \dots; (S, 1), \dots, (S, S).$$

If $PS = P = (p_{ss'})$, $PJ = (\pi_{ss'j})$, and $G = (G_{ss'j})$, then the Pareto exponent $z = \zeta$ is the solution to

$$(1-p)\rho(P\odot M(z))=1,$$

where ρ is the spectral radius and $M(z) = (M_{ss'}(z))$,

$$M_{ss'}(z) = \sum_{j=1}^{J} \pi_{ss'j} G_{ss'j}^{z},$$

and \odot is the Hadamard (entry-wise) product.

PJ can be either $1 \times J$, $S \times J$, or $S^2 \times J$. If it is $1 \times J$, it assumes $\pi_{ss'j} = \pi_j$ depends only on j. If it is $S \times J$, it assumes $\pi_{ss'j} = \pi_{sj}$ depends only on (s, j).

G can be either $S \times J$ or $S^2 \times J$. If it is $S \times J$, it assumes $G_{ss'j} = G_{sj}$ depends only on (s, j).

2.2 Computing joint transition probability matrix

getQ.m computes the $SN \times SN$ joint transition probability matrix $Q = (q_{sn,s'n'})$ and the stationary distribution $\pi = (\pi_{sn})$ for the exogenous state s and wealth. The usage is

where

- PS, PJ, p are as above,
- x0 is the initial wealth of newborn agents,
- xGrid is the $1 \times N$ grid of wealth (size variable) w_n ,
- gstjn is the $S^2 \times JN$ matrix of law of motion for wealth $g_{ss'i}(w_n)$,

- Gstj is the $S^2 \times J$ matrix of asymptotic slopes of law of motion $G_{ss'j}$ (optional),
- zeta is the Pareto exponent (optional), and
- h is the grid spacing for hypothetical grid points (optional).

The JN columns of \mathtt{gstjn} must be ordered such that the first N columns correspond to j=1, the next N columns correspond to j=2, and so on. \mathtt{Gstj} is the same as \mathtt{G} in $\mathtt{getZeta.m}$. If unspecified, it uses the slope of the law of motion between the two largest grid points. If \mathtt{zeta} is unspecified, it calls $\mathtt{getZeta.m}$ to compute. If \mathtt{h} is unspecified, it uses the distance between the two largest grid points.

gstjn can be either $S \times JN$ or $S^2 \times JN$. If it is $S \times JN$, it assumes $g_{ss'j}(w_n) = g_{sj}(w_n)$ depends only on (s, j, n).

2.3 Computing top wealth shares

getTopShares.m computes the top wealth shares. The usage is

topShare = getTopShares(topProb,wGrid,wDist,zeta)

where

- topProb is the vector of top probabilities to evaluate top shares,
- wGrid is the $1 \times N$ vector of wealth grid,
- wDist is the $1 \times N$ vector of wealth distribution, and
- zeta is the Pareto exponent (optional).

Given the stationary distribution π computed using getQ.m, one can compute the wealth distribution as $\pi_n = \sum_{s=1}^S \pi_{sn}$. If zeta is unspecified, getTopShares.m uses spline interpolation to compute top wealth shares.

References

Brendan K. Beare and Alexis Akira Toda. Geometrically stopped Markovian random growth processes and Pareto tails. 2017. URL https://arxiv.org/abs/1712.01431.

Émilien Gouin-Bonenfant and Alexis Akira Toda. Pareto extrapolation: Bridging theoretical and quantitative models of wealth inequality. 2018. URL https://ssrn.com/abstract=3260899.