

EGMⁿ: The Sequential Endogenous Grid Method

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Heterogeneous agent models with multiple decision choices are often solved using inefficient grid search methods that require a large number of points and are time intensive. This paper provides a novel method for solving such models using an extension of the endogenous grid method (EGM) that uses Gaussian Process Regression (GPR) to interpolate functions on unstructured grids. First, separating models into smaller, sequential problems allows the problems to be more tractable and easily analyzed. Second, using an exogenous grid of post-decision states and solving for an endogenous grid of pre-decision states that obey a first order condition greatly speeds up the solution process. Third, since the resulting endogenous grid can often be curvilinear at best and unstructured at worst, GPR provides an efficient and accurate method for interpolating the value, marginal value, and policy functions. Applied sequentially to each decision within the overarching problem, the method is able to solve heterogeneous agent models with multiple decision choices in a fraction of the time and with less computational resources than are required by standard grid search methods currently used. This paper also illustrates how this method can be applied to a number of increasingly complex models. Software is provided in the form of a `python` module under the **Econ-ARK/HARK** package.

PDF: [SequentialEGM.pdf](#) (Current Version)
html: <https://alanlujan91.github.io/SequentialEGM/>
GitHub: <https://github.com/alanlujan91/SequentialEGM>

I would like to thank Christopher D. Carroll and Simon Scheidegger for their helpful comments and suggestions. Remaining errors are my own. All figures and other numerical results were produced using the **Econ-ARK/HARK** toolkit (Carroll, Kaufman, Kazil, Palmer, and White (2018)). Additional libraries used in the production of this paper include but are not limited to: **scipy** (Virtanen, Gommers, Oliphant, Haberland, Reddy, Cournapeau, Burovski, Peterson, Weckesser, Bright, van der Walt, Brett, Wilson, Millman, Mayorov, Nelson, Jones, Kern, Larson, Carey, Polat, Feng, Moore, VanderPlas, Laxalde, Perktold, Cimrman, Henriksen, Quintero, Harris, Archibald, Ribeiro, Pedregosa, van Mulbregt, and SciPy 1.0 Contributors (2020)), **numpy** (Harris, Millman, van der Walt, Gommers, Virtanen, Cournapeau, Wieser, Taylor, Berg, Smith, Kern, Picus, Hoyer, van Kerkwijk, Brett, Haldane, Del Río, Wiebe, Peterson, Gérard-Marchant, Sheppard, Reddy, Weckesser, Abbasi, Gohlke, and Oliphant (2020)), **numba** (Lam, Pitrou, and Seibert (2015)), **cupy** (Okuta, Unno, Nishino, Hido, and Loomis (2017)), **scikit-learn** (Pedregosa, Varoquaux, Gramfort, Michel, Thirion, Grisel, Blondel, Prettenhofer, Weiss, Dubourg, Vanderplas, Passos, Cournapeau, Brucher, Perrot, and Duchesnay (2011)), **pytorch** (Paszke, Gross, Massa, Lerer, Bradbury, Chanan, Killeen, Lin, Gimelshein, Antiga, and Others (2019)), and **gpytorch** (Gardner, Pleiss, Bindel, Weinberger, and Wilson (2018)).

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