

Discussion of: *Merging
Simulation and Projection
Approaches to Solve High-
Dimension Problems* by
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Overall Approach

- Problem: Solve for $y(s, z), s'(s, z), (z', y')$:

$$E_t[G(s, z, y, s', z', y')] = 0, z' = Z(Z, \epsilon).$$
- Method:
 - Start with a guess
 - Simulate the system
 - Construct an ϵ -distinguishable set (EDS)
 - Solve over the EDS using a projection method
 - Iterate until the grid converges

Innovation and Significance

- EDS method
- General ideas:
 - Restrict the grid region to high-probability events by deleting points below threshold on kernel-based density approximation
 - Obtain *uniform* collection of points by eliminating ϵ -neighbors iteratively (and using PCA)
- Results: Effective and efficient solutions even in high dimensions

Potential Problems in Paper

- Lack of convergence proof
- Unstable simulation results (use restricted rules with growth limits)
- Failed fixed-point iterations (use Newton or global search)
- Equilibrium not attained (check on second order conditions)

Other Issues and Extensions

- Alternative grids/bounds, e.g.: sparse grid (Smolyak)

For Sobolev spaces of functions f on R^d with smoothness r , error:

$$O\left(\frac{(\log n)^{(d-1)(r+1)}}{n^r}\right).$$

- Issue: may not have smoothness (although bounds maybe still be possible)
- Can adapt to different regions
- For smooth functions, irregular points can

help.
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Simulation Setup and Convexity

- Use of low-discrepancy sequences for simulation
 - Pseudo-random draws may help for building confidence intervals but here?
 - Low-discrepancy sequences would appear to offer better initial coverage
- Use of structural properties (e.g., convexity) in the approximations
 - Possible in neoclassical growth example
 - Can use global information

Monte Carlo Methods

- MCMC and particle-type methods with re-sampling for reducing variance and importance sampling

$$\int G(s'(s, z'), y'(s, z'), s, y, z') \phi(z'|z) dz' = \int G(s', y', s, y, z') \frac{\phi(z'|z)}{g(z')} g(z') dz'$$

- Latent variables for extended Gaussian mixtures

$$\int G(s', z', y', s, z, y) \phi(z'|z) dz' = \int \int G(s', z', y', s, z, y) g(w, s', z', y') \phi(z'|z) dw dz'$$

Analytical Results

- Convergence
 - Characterize conditions?
 - Guidance on implementation?
- Error bounds
 - Asymptotic
 - Explicit
 - Estimated