A Proposal for Community (Quasi-) Monte Carlo Software

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Follow-Up to Discussions, June 28, 2018

Motivation

■ Where can I find quality, free quasi-Monte Carlo (qMC) software?

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Landscape

- Where can I find quality, free quasi-Monte Carlo (gMC) software?
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My initial attempts in this direction over the past 5 years have produced GAIL¹

Motivation

Can We Have qMC Community Software that Grows Up to Be Like ...

Chebfun Computing with Chebyhsev polynomials, chebfun.org

Clawpack Solution of conservation laws, clawpack.org

deal.II Finite-elements, http://dealii.org

Mission: To provide well-documented tools to build finite element codes for a broad variety of PDEs, from laptops to supercomputers.

Vision: To create an open, inclusive, participatory community providing users and developers with a state-of-the-art, comprehensive software library that constitutes the go-to solution for all finite element problems.

FEniCS Finite-elements, fenicsproject.org

Gromacs Molecular dynamics, gromacs.org

Stan Markov Chain Monte Carlo, mc-stan.org

Trilinos Multiphysics computations, trilinos.org

- Developed and supported by multiple research groups
- Used beyond the research groups that develop it
- A recognized standard in its field

What Is Available Now

Motivation

John Burkhardt Variety of gMC Software in C++, Fortran, MATLAB, and Python, people.sc.fsu.edu/~iburkardt/

Mike Giles Multi-Level Monte Carlo Software in C++, MATLAB, Python, and R, people.maths.ox.ac.uk/gilesm/mlmc/

Fred Hickernell Guaranteed Automatic Integration Library (GAIL) in MATLAB, gailgithub.github.io/GAIL_Dev/

Stephen Joe & Frances Kuo Sobol' generators in C++. Generating vectors for lattices. web.maths.unsw.edu.au/~fkuo/

Pierre L'Ecuyer Random number generators, Stochastic Simulation, Lattice Builder in C/C++ and Java. simul.iro.umontreal.ca

Dirk Nuyens Magic Point Shop, QMC4PDE, etc. in MATLAB, Python, and C++, people.cs.kuleuven.be/~dirk.nuyens/

Art Owen Various code, statweb.stanford.edu/~owen/code/

MATLAB Sobol' and Halton sequences

Python Sobol' and Halton sequences

R randtoolbox Sobol', lattice, and Halton sequences

Key Elements

Motivation

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- Packages that display output in tables or plots

Languages and Architectures

Motivation

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- How will parallel computing be supported?

Good Development Practices

Motivation

■ Start small, with good skeleton

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- Marketing on websites and at conferences

Bigger than One Research Group

- We have experience over the past 5 years developing GAIL²
- Our group has learned some of the discipline required to develop good software
- Our group does not have the capacity to tackle this whole project, and neither does your group
- A good software library should attract developers
- Let's leave a legacy to our community that goes beyond theorems and algorithms

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6/28



Costs	Benefits
Less time to prove theorems	More impact for our theorems
Learning a new language	Wider access and better performance for our code
Compromise with other research groups	More capable code than can be produced by one research group
Time spent writing documentation and tests	Fewer bugs for those who use the code
	Attract more qMC developers
	Attract more qMC users
	Happier funding agencies

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 - An initial use case
- We welcome you to join us. Email me to join our Google group

Elements of the Community Software—Discrete Distributions

```
classdef (Abstract) discreteDistribution
%Specifies and generates the components of a_n \sum w_i \delta_{\mathbf{x}_i}(\cdot)
properties (Abstract)
   distribData %information required to generate the distribution
   state %state of the generator
   nSt reams
end
properties
   domain = [0 0; 1 1]; %domain of the discrete distribution, X
   domainType = 'box' %domain of the discrete distribution, X
   dimension = 2 %dimension of the domain, d
   trueDistribution = 'uniform' %name of the distribution that the discrete ...
       distribution attempts to emulate
end
```

Elements of the Community Software—Discrete Distributions

```
classdef (Abstract) discreteDistribution
methods (Abstract)
  genDistrib(obj, nStart, nEnd, n, coordIndex)
  % nStart = starting value of i
  % nEnd = ending value of i
  % n = value of n used to determine a<sub>n</sub>
  % coordIndex = which coordinates in sequence are needed
end
end
```

Elements of the Community Software—Discrete IID Uniform

```
classdef IIDDistribution < discreteDistribution
%Specifies and generates the components of \frac{1}{n}\sum^n \delta_{\mathbf{x}_i}(\cdot)
%where the \mathbf{x}_i are IID uniform on [0,1]^d or IID standard Gaussian
properties
   distribData %stream data
   state = [] %not used
   nStreams = 1
end
methods.
   function obj = initStreams(obj,nStreams)
       obj.nStreams = nStreams;
       obj.distribData.stream = ...
           RandStream.create('mrg32k3a','NumStreams',nStreams,'CellOutput',true);
   end
```

```
classdef IIDDistribution < discreteDistribution</pre>
   function [x, w, a] = genDistrib(obj, nStart, nEnd, n, coordIndex, ...
       streamIndex)
      if nargin < 6
         streamIndex = 1;
      end
      nPts = nEnd - nStart + 1; %how many points to be generated
      if strcmp(obj.trueDistribution, 'uniform') %generate uniform points
         x = \dots
             rand(obj.distribData.stream{streamIndex},nPts,numel(coordIndex)); ...
             %nodes
      else %standard normal points
         x = \dots
             randn(obj.distribData.stream{streamIndex}, nPts, numel(coordIndex)); ...
             %nodes
      end
      w = 1:
      a = 1/n:
   end
end
```

Elements of the Community Software—Functions

```
classdef (Abstract) fun
% Specify and generate values f(x) for x \in X
properties
    domain = [0\ 0;\ 1\ 1] %domain of the function, X
    domainType = 'box' %e.g., 'box', 'ball'
   dimension = 2 %dimension of the domain, d
   distribType = 'uniform' %e.g., 'uniform', 'Gaussian'
   nominal Value = 0 % a nominal number, c, such that (c, \ldots, c) \in \mathcal{X}
end
methods (Abstract)
     v = f(obi, xu, coordIndex)
    % xu = nodes, \mathbf{x}_{u,i} = i^{\text{th}} row of an n \times |\mathbf{u}| matrix
    % coordIndex = set of those coordinates in sequence needed, \mathfrak u
    % y = n \times p matrix with values f(\mathbf{x}_{\mathfrak{u},i},\mathbf{c}) where if \mathbf{x}_i' = (x_{i,\mathfrak{u}},\mathbf{c})_i, then x_{ii}' = x_{ii} for
        j \in \mathfrak{u}, and x'_{ii} = c otherwise
end
```

Elements of the Community Software—Keister's Function

```
classdef KeisterFun < fun
% Specify and generate values f(\mathbf{x}) for \mathbf{x} \in \mathcal{X}
methods.
   function y = f(obj, x, coordIndex)
      %if the nominal Value = 0, this is efficient
      normx2 = sum(x.*x.2):
      if (numel(coordIndex) ≠ obj.dimension) && (obj.nominalValue ≠ 0)
          normx2 = normx2 + (obj.nominalValue.^2) * (obj.dimension - ...
              numel(coordIndex));
      end
      y = \exp(-normx2) \cdot * \cos(sgrt(normx2));
   end
end
end
```

Elements of the Community Software—StoppingCriteria

```
classdef (Abstract) stoppingCriterion
% Decide when to stop a
properties
   absTol = 1e-2 %absolute tolerance, d
   relTol = 0 %absolute tolerance, d
  nInit = 1024 %initial sample size
  nMax = 1e8 %maximum number of samples allowed
end
properties (Abstract)
   discDistAllowed %which discrete distributions are supported
   decompTypeAllowed %which decomposition types are supported
end
methods (Abstract)
    stopYet(obj, distribObj)
   % distribObj = data or summary of data computed already
end
end
```

Elements of the Community Software—CLTStopping

```
classdef CLTStopping < stoppingCriterion</pre>
% Stopping criterion based on the Central Limit Theorem
properties
   discDistAllowed = "IIDDistribution" %which discrete distributions are ...
       supported
   decompTypeAllowed = ["single"; "multi"] %which decomposition types are ...
       supported
   inflate = 1.2 %inflation factor
   alpha = 0.01;
end
properties (Dependent)
   quantile
end
```

Elements of the Community Software—CLTStopping

```
classdef CLTStopping < stoppingCriterion
methods
  function [obj, dataObj, distribObj] = ...
     stopYet(obj, dataObj, funObj, distribObj)
  if ¬numel(dataObj)
     dataObj = meanVarData;
  end
  switch dataObj.stage</pre>
```

Elements of the Community Software—CLTStopping

```
classdef CLTStopping < stoppingCriterion</pre>
      switch dataObj.stage
         case 'begin' %initialize
            dataObi.timeStart = tic;
            if ¬anv(strcmp(obj.discDistAllowed,class(distribObj)))
               error('Stoppoing criterion not compatible with sampling ...
                   distribution')
            end
            nf = numel(funObj); %number of functions whose integrals add up ...
                to the solution
            distribObj = initStreams(distribObj,nf); %need an IID stream ...
                for each function
            dataObj.prevN = zeros(1,nf); %initialize data object
            dataObj.nextN = repmat(obj.nInit,1,nf);
            dataObj.muhat = Inf(1,nf);
            dataObj.sighat = Inf(1,nf);
            dataObj.nSigma = obj.nInit; %use initial samples to estimate ...
                standard deviation
            dataObj.costF = zeros(1,nf);
            dataObj.stage = 'sigma'; %compute standard deviation next
```

Elements of the Community Software—CLTStopping

```
classdef CLTStopping < stoppingCriterion</pre>
         case 'sigma'
            dataObj.prevN = dataObj.nextN; %update place in the sequence
            tempA = sgrt(dataObi.costF); %use cost of function values to ...
               decide how to allocate
            tempB = sum(tempA .* dataObj.sighat); %samples for computation ...
               of the mean
            nM = ceil((tempB*(obj.quantile*obj.inflate ...
               /max(obj.absTol,dataObj.solution*obj.relTol))^2) ...
               * (dataObj.sighat./sqrt(dataObj.costF)));
            dataObj.nMu = min(max(dataObj.nextN,nM),obj.nMax - dataObj.prevN);
            dataObj.nextN = dataObj.nMu + dataObj.prevN;
            dataObj.stage = 'mu'; %compute sample mean next
```

Elements of the Community Software—CLTStopping

```
classdef CLTStopping < stoppingCriterion</pre>
         case 'mii'
            dataObj.solution = sum(dataObj.muhat);
            dataObj.nSamplesUsed = dataObj.nextN;
            errBar = (obj.quantile * obj.inflate) * ...
               sgrt(sum(dataObj.sighat.^2/dataObj.nMu));
            dataObj.errorBound = dataObj.solution + errBar*[-1 \ 1];
            dataObi.stage = 'done'; %finished with computation
      end
      dataObj.timeUsed = toc(dataObj.timeStart);
  end
   function value = get.quantile(obj)
      value = -\text{norminv}(obi.alpha/2);
   end
end
end
```

Elements of the Community Software—Integration

```
function [solution, dataObj] = integrate(funObj, distribObj, stopCritObj)
Specify and generate values f(x) for x \in X
% funObj = an object from class fun
% distribObj = an object from class discrete distribution
% stopcritObj = an object from class stopping criterion
%Initialize the accumData object and other crucial objects
[stopCritObj, dataObj, distribObj] = stopYet(stopCritObj, [], funObj, ...
   distribObj);
while -strcmp(dataObj.stage, 'done') %the dataObj.stage property tells us ...
   where we are in the process
  dataObj = updateData(dataObj, distribObj, funObj); %compute additional data
   [stopCritObj, dataObj] = stopYet(stopCritObj, dataObj, funObj); %update ...
      the status of the computation
end
solution = dataObj.solution; %assign outputs
dataObj.timeUsed = toc(dataObj.timeStart);
```

Elements of the Community Software—Integration Example

```
%% How to integrate a function using our community QMC framework
% An example with Keister's function integrated with respect to the uniform
% distribution over the unit cube
stopObj = CLTStopping %stopping criterion for IID sampling using the ...
   Central Limit Theorem
distribObj = IIDDistribution; %IID sampling with uniform distribution
[sol, out] = integrate(KeisterFun, distribObj, stopObj)
```

```
>> IntegrationExample
sol = 0.4310
out =
timeUsed: 0.0014
```

Motivation

nSamplesUsed: 7540 errorBound: [0.4210 0.4410]

Elements of the Community Software—Integration Example

```
stopObj.absTol = 1e-3 %decrease tolerance
[sol, out] = integrate(KeisterFun, distribObj, stopObj)
```

```
sol = 0.4253
out =
```

Motivation

timeUsed: 0.0333

nSamplesUsed: 652546

errorBound: [0.4243 0.4263]

Elements of the Community Software—Integration Example

```
stopObj.absTol = 0; %impossible tolerance
stopObj.nMax = 1e6; %calculation limited by sample budget
[sol, out] = integrate(KeisterFun, distribObj, stopObj)
```

```
sol = 0.4252
out =
```

Motivation

timeUsed: 0.0392

nSamplesUsed: 1000000

errorBound: [0.4244 0.4260]

Elements of the Community Software—Asian Call, Low D

```
%A multilevel example of Asian option pricing distribObj.trueDistribution = 'normal'; %Change to normal distribution stopObj.absTol = 0.01; %increase tolerance stopObj.nMax = 1e8; %pushing the sample budget back up OptionObj = AsianCallFun(4) %4 time steps [sol, out] = integrate(OptionObj, distribObj, stopObj)
```

```
OptionObj = dimension: 4 sol = 6.1740 out =
```

Motivation

+imalica.

timeUsed: 1.0517

nSamplesUsed: 5680546

errorBound: [6.1640 6.1840]

Elements of the Community Software—Asian Call, High D

```
OptionObj = AsianCallFun(64) %single level, 64 time steps
[sol, out] = integrate(OptionObj, distribObj, stopObj)
```

```
OptionObj =
AsianCallFun with properties:
dimension: 64
sol =6.2036
out =
meanVarData with properties:
timeUsed: 25.1910
nSamplesUsed: 5610402
errorBound: [6.1936 6.2136]
```

Elements of the Community Software—Asian Call, Multi-Level

```
OptionObj = AsianCallFun([4 4 4]) %multilevel, 64 time steps, faster
[sol, out] = integrate(OptionObj, distribObj, stopObj)
```

```
OptionObj =
1×3 AsianCallFun array with properties:
sol = 6.2052
out =
timeUsed: 2.2171
nSamplesUsed: [8080862 446720 85907]
errorBound: [6.1968 6.2135]
```

Thank you

