```
# Documentation
This version of the document is dated 2023-04-28.
Help on module randomgen:
NAME
    randomgen
DESCRIPTION
    Sample code for the article "Randomization and Sampling Methods"
    [https://www.codeproject.com/Articles/1190459/Random-Number-
Generation-Methods]
(https://www.codeproject.com/Articles/1190459/Random-Number-
Generation-Methods)
    Written by Peter 0.
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(https://creativecommons.org/publicdomain/zero/1.0/)
CLASSES
    builtins.object
        AlmostRandom
        BinaryExpansion
        BringmannLarsen
        ConvexPolygonSampler
        DensityInversionSampler
        DensityTiling
        {\sf FastLoadedDiceRoller}
        KVectorSampler
        OptimalSampler
        PascalTriangle
        PrefixDistributionSampler
        RandomGen
        RatioOfUniformsTiling
        SortedAliasMethod
        VoseAlias
```

class AlmostRandom(builtins.object)

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AlmostRandom(randgen, list)
     | Methods defined here:
       init (self, randgen, list)
           Initialize self. See help(type(self)) for accurate
signature.
       choose(self)
     | Data descriptors defined here:
     | __dict_
           dictionary for instance variables (if defined)
       weakref
           list of weak references to the object (if defined)
   class BinaryExpansion(builtins.object)
       BinaryExpansion(arr, zerosAtEnd=False)
     | Methods defined here:
       init (self, arr, zerosAtEnd=False)
           Binary expansion of a real number in [0, 1], initialized
           from an array of zeros and ones expressing the binary
           expansion.
           The first binary digit is the half digit, the second
           is the quarter digit, the third is the one-eighth digit,
           and so on. Note that the number 1 can be
           expressed by passing an empty array and specifying
           zerosAtEnd = False, and the number 0 can be
           expressed by passing an empty array and specifying
           zerosAtEnd = True.
           arr - Array indicating the initial digits of the binary
           expansion.
           zerosAtEnd - Indicates whether the binary expansion
           is expressed as 0.xxx0000... or 0.yyy1111... (e.g.,
0.1010000...
           vs. 0.1001111.... Default is the latter case (False).
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| entropy(self)
    | eof(self)
           Returns True if the end of the binary expansion was
reached; False otherwise.
    | fromFloat(f)
           Creates a binary expansion object from a 64-bit
floating-point number in the
          interval [0, 1].
    | fromFraction(f)
           Creates a binary expansion object from a fraction in the
           interval [0, 1].
    | get(f)
           Creates a binary expansion object from a fraction,
'int'. or
           'float' in the interval [0, 1]; returns 'f' unchanged,
otherwise.
    | get0rReset(f)
           Creates a binary expansion object from a fraction,
'int', or
           'float' in the interval [0, 1]; resets 'f' (calls its
reset method) otherwise.
    | nextbit(self)
           Reads the next bit in the binary expansion.
    | reset(self)
           Resets this object to the first bit in the binary
expansion.
    | value(self)
                .....
    | Data descriptors defined here:
    | dict
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dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
   class BringmannLarsen(builtins.object)
    | BringmannLarsen(weights)
    | Implements Bringmann and Larsen's sampler, which chooses a
random number in [0, n)
    | where the probability that each number is chosen is
weighted. The 'weights' is the
    list of weights each 0 or greater; the higher the weight,
the greater
    | the probability. This sampler supports only integer
weights.
    | This is a succinct (space-saving) data structure for this
purpose.
    | Reference:
    | K. Bringmann and K. G. Larsen, "Succinct Sampling from
Discrete
    | Distributions", In: Proc. 45th Annual ACM Symposium on
Theory
    | of Computing (STOC'13), 2013.
    | Methods defined here:
      __init__(self, weights)
           Initialize self. See help(type(self)) for accurate
signature.
    | next(self, randgen)
                  .....
    | Data descriptors defined here:
    | dict
           dictionary for instance variables (if defined)
      weakref
```

```
list of weak references to the object (if defined)
   class ConvexPolygonSampler(builtins.object)
    | ConvexPolygonSampler(randgen, points)
    | A class for uniform random sampling of
     | points from a convex polygon. This
    | class only supports convex polygons because
    | the random sampling process involves
    | triangulating a polygon, which is trivial
    | for convex polygons only. "randgen" is a RandomGen
     | object, and "points" is a list of points
    | (two-item lists) that make up the polygon.
     | Methods defined here:
      __init__(self, randgen, points)
           Initialize self. See help(type(self)) for accurate
signature.
      sample(self)
           Choose a random point in the convex polygon
           uniformly at random.
           | Data descriptors defined here:
      __dict__
           dictionary for instance variables (if defined)
       weakref
           list of weak references to the object (if defined)
   class DensityInversionSampler(builtins.object)
    DensityInversionSampler(pdf, bl, br, ures=1e-08)
    | A sampler that generates random samples from
         a continuous distribution for which
         only the probability density function (PDF) is known,
         using the inversion method. This sampler
         allows quantiles for the distribution to be calculated
```

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- pdf: A function that specifies the PDF. It takes a single
          number and outputs a single number. The area under
          the PDF need not equal 1 (this sampler works even if the
          PDF is only known up to a normalizing constant).
        - bl, br - Specifies the sampling domain of the PDF. Both
           bl and br are numbers giving the domain,
          which in this case is [bl, br]. For best results, the
           probabilities outside the sampling domain should be
           negligible (the reference cited below uses cutoff points
           such that the probabilities for each tail integrate to
           about ures*0.05 or less).
        - ures - Maximum approximation error tolerable, or
          "u-resolution". Default is 10^-8. This error tolerance
          "does not work for continuous distributions [whose PDFs
          have] high and narrow peaks or poles". This sampler's
          approximation error will generally be less than this
tolerance,
          but this is not guaranteed, especially for PDFs of the
kind
          just mentioned.
          Reference:
          Gerhard Derflinger, Wolfgang Hörmann, and Josef Leydold,
          "Random variate generation by numerical inversion when
          only the density is known", ACM Transactions on Modeling
          and Computer Simulation 20(4) article 18, October 2010.
       Methods defined here:
       init (self, pdf, bl, br, ures=1e-08)
           Initialize self. See help(type(self)) for accurate
signature.
       codegen(self, name='dist')
            Generates standalone Python code that samples
                    (approximately) from the distribution estimated
                    in this class. Idea from Leydold, et al.,
                    "An Automatic Code Generator for
                    Nonuniform Random Variate Generation", 2001.
            - name: Distribution name. Generates Python methods
```

from pregenerated uniform random numbers in [0, 1].

```
called
               sample_X (samples one random number), and quantile_X
               (finds the quantile
               for a uniform random number in [0, 1]),
              where X is the name given here.
       quantile(self, v)
            Calculates quantiles from uniform random numbers
                 in the interval [0, 1].
            - v: A list of uniform random numbers.
            Returns a list of the quantiles corresponding to the
            uniform random numbers. The returned list will have
           the same number of entries as 'v'.
       sample(self, rg, n=1)
           Generates random numbers that (approximately) follow the
                 distribution modeled by this class.
            - n: The number of random numbers to generate.
            Returns a list of 'n' random numbers.
       Data descriptors defined here:
       dict
           dictionary for instance variables (if defined)
        weakref
           list of weak references to the object (if defined)
   class DensityTiling(builtins.object)
     | DensityTiling(pdf, bl, br, cycles=8)
     | Produces a tiling of a probability density function (PDF)
             for the purposes of random number generation. The PDF
is
             decomposed into tiles; these tiles will either cross
the PDF
            or go below the PDF. In each recursion cycle, each
tile is
             split into four tiles, and tiles that end up above the
PDF are
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```
discarded.
      - pdf: A function that specifies the PDF. It takes a single
          number and outputs a single number. The area under
          the PDF need not equal 1 (this class tolerates the PDF
even if
          it is only known up to a normalizing constant). For best
results,
          the PDF should be less than or equal to a finite number
(thus, it should be free of _poles_, or points
          that approach infinity). If the PDF does contain a pole,
this class
         may accommodate the pole by sampling from a modified
     1
version of the PDF,
          so that points extremely close to the pole may be sampled
         at a higher or lower probability than otherwise (but not
in a way
         that significantly affects the chance of sampling points
          outside the pole region).
     | - bl, br - Specifies the sampling domain of the PDF. Both
           bl and br are numbers giving the domain,
          which in this case is [bl, br].
       - cycles - Number of recursion cycles in which to split
tiles
          that follow the PDF. Default is 8.
        Additional improvements not yet implemented: Hörmann et
al.,
         "Inverse Transformed Density Rejection for Unbounded
Monotone Densities", 2007.
        Reference:
        Fulger, Daniel and Guido Germano. "Automatic generation of
         non-uniform random variates for arbitrary pointwise
computable
         probability densities by tiling",
         arXiv:0902.3088v1 [cs.MS], 2009.
     | Methods defined here:
        __init__(self, pdf, bl, br, cycles=8)
            Initialize self. See help(type(self)) for accurate
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```
signature.
       codegen(self, name, pdfcall=None)
            Generates Python code that samples
                    (approximately) from the distribution estimated
                    in this class. Idea from Leydold, et al.,
                    "An Automatic Code Generator for
                    Nonuniform Random Variate Generation", 2001.
            - name: Distribution name. Generates a Python method
called
               sample X where X is the name given here (samples one
               random number).
            - pdfcall: Name of the method representing pdf (for more
information,
              see the __init__ method of this class). Optional; if
not given
               the name is pdf_X where X is the name given in the
name parameter.
       maybeAppend(self, pdfevals, newtiles, xmn, xmx, ymn, ymx)
       sample(self, rg, n=1)
           Generates random numbers that (approximately) follow the
                  distribution modeled by this class.
            - n: The number of random numbers to generate.
            Returns a list of 'n' random numbers.
       Data descriptors defined here:
       dict
           dictionary for instance variables (if defined)
        weakref
            list of weak references to the object (if defined)
   class FastLoadedDiceRoller(builtins.object)
       FastLoadedDiceRoller(weights)
     | Implements the Fast Loaded Dice Roller, which chooses a
random number in [0, n)
```

```
| where the probability that each number is chosen is
weighted. The 'weights' is the
     | list of weights each 0 or greater; the higher the weight,
the greater
     | the probability. This sampler supports only integer
weights.
     Reference: Saad, F.A., Freer C.E., et al. "The Fast Loaded
Dice Roller: A
     | Near-Optimal Exact Sampler for Discrete Probability
Distributions", in
     _AISTATS 2020: Proceedings of the 23rd International
Conference on Artificial
     | Intelligence and Statistics, Proceedings of Machine Learning
Research_ 108,
     | Palermo, Sicily, Italy, 2020.
     | Methods defined here:
       __init__(self, weights)
           Initialize self. See help(type(self)) for accurate
signature.
       codegen(self, name='sample discrete')
           Generates standalone Python code that samples
                   from the distribution modeled by this class.
                   Idea from Leydold, et al.,
                    "An Automatic Code Generator for
                   Nonuniform Random Variate Generation", 2001.
            - name: Method name. Default: 'sample discrete'.
     | next(self, randgen)
       Data descriptors defined here:
     | dict
           dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
```

```
class KVectorSampler(builtins.object)
       KVectorSampler(cdf, xmin, xmax, pdf=None, nd=200)
     | A K-Vector-like sampler of a non-discrete distribution
     | with a known cumulative distribution function (CDF).
       Uses algorithms
     described in Arnas, D., Leake, C., Mortari, D., "Random
       Sampling using k-vector", Computing in Science &
       Engineering 21(1) pp. 94-107, 2019, and Mortari, D.,
       Neta, B., "k-Vector Range Searching Techniques".
     | Methods defined here:
       __init__(self, cdf, xmin, xmax, pdf=None, nd=200)
           Initializes the K-Vector-like sampler.
           Parameters:
            - cdf: Cumulative distribution function (CDF) of the
               distribution. The CDF must be
               strictly increasing everywhere in the
               interval [xmin, xmax] and must output values in [0,
11;
               for best results, the CDF should
               be increasing everywhere in [xmin, xmax].
            - xmin: Maximum x-value to generate.
            - xmax: Maximum x-value to generate. For best results,
               the range given by xmin and xmax should cover all or
               almost all of the distribution.
            - pdf: Optional. Distribution's probability density
               function (PDF), to improve accuracy in the root-
finding
               process.
            - nd: Optional. Size of tables used in the sampler.
               Default is 200.
       quantile(self, uniforms)
            Returns a list of 'n' numbers that correspond
            to the given uniform random numbers and follow
            the distribution represented by this sampler.
'uniforms'
            is a list of uniform random values in the interval
            [0, 1]. For best results, this sampler's range
```

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(xmin and xmax in the constructor)
            should cover all or almost all of the desired
distribution and
            the distribution's CDF should be strictly
            increasing everywhere (every number that can be taken on
           by the distribution has nonzero probability of
occurring), since
            among other things,
            this method maps each uniform value to the
            range of CDFs covered by this distribution (that is,
            [0, 1] is mapped to [minCDF, maxCDF]), and
            uniform values in "empty" regions (regions with
            constant CDF) are handled by replacing those
           values with the minimum CDF value covered.
       sample(self, rg, n)
            Returns a list of 'n' random numbers of
            the distribution represented by this sampler.
            - rg: A random generator (RandGen) object.
      Data descriptors defined here:
       dict
           dictionary for instance variables (if defined)
       weakref
            list of weak references to the object (if defined)
   class OptimalSampler(builtins.object)
     | OptimalSampler(m)
     Implements a sampler which chooses a random number in [0, n)
     | where the probability that each number is chosen is
weighted. The 'weights' is the
     | list of weights each 0 or greater; the higher the weight,
the greater
     | the probability. This sampler supports only integer
weights, but the sampler is
     | entropy-optimal as long as the sum of those weights is of
the form 2^k or 2^k-2^m.
```

```
Reference: Feras A. Saad, Cameron E. Freer, Martin C.
Rinard, and Vikash K. Mansinghka.
     | Optimal Approximate Sampling From Discrete Probability
Distributions. Proc.
     | ACM Program. Lang. 4, POPL, Article 36 (January 2020), 33
pages.
       Methods defined here:
       init (self, m)
           Initialize self. See help(type(self)) for accurate
signature.
       codegen(self, name='sample_discrete')
           Generates standalone Python code that samples
                    from the distribution modeled by this class.
                    Idea from Leydold, et al.,
                    "An Automatic Code Generator for
                    Nonuniform Random Variate Generation", 2001.
            - name: Method name. Default: 'sample discrete'.
       next(self, rg)
       nextFromMatrix(self, pm, rg)
     | Data descriptors defined here:
     | dict
           dictionary for instance variables (if defined)
        __weakref__
            list of weak references to the object (if defined)
   class PascalTriangle(builtins.object)
     | Generates the rows of Pascal's triangle, or the
     \mid weight table for a binomial(n,1/2) distribution.
       Methods defined here:
```

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| init (self)
            Initialize self. See help(type(self)) for accurate
signature.
      aliasinfo(self, desiredRow)
       getrow(self, desiredRow)
            Calculates an arbitrary row of Pascal's triangle.
       next(self)
            Generates the next row of Pascal's triangle, starting
with
            row 0. The return value is a list of row-number-choose-k
            values.
       nextto(self, desiredRow)
            Generates the row of Pascal's triangle with the given
row number,
            skipping all rows in between. The return value is a
list of
            row-number-choose-k values.
       row(self)
            Gets the row number of the row that will be generated
            the next time _next_ is called.
     | Data descriptors defined here:
     | dict
            dictionary for instance variables (if defined)
        __weakref__
            list of weak references to the object (if defined)
    class PrefixDistributionSampler(builtins.object)
     | PrefixDistributionSampler(pdf)
     | An arbitrary-precision sampler for probability distributions
     | supported on [0, 1] and less than or equal to a finite
number.
```

```
| Note that this sampler currently relies on floating-point
operations
    | and thus the evaluations of the PDF (the distribution's
probability
    | density function) could incur rounding errors.
    - pdf: PDF, which takes a value in [0, 1] and returns a
probability
         density at that value (which is 0 or greater). Currently,
    1
         the PDF must be strictly increasing or strictly
decreasing.
    | Reference: Oberhoff, Sebastian, "Exact Sampling and Prefix
    | Distributions", Theses and Dissertations, University of
    | Wisconsin Milwaukee, 2018.
    | Methods defined here:
    | __init__(self, pdf)
           Initialize self. See help(type(self)) for accurate
signature.
    fill(self, rg, prefixLength, prefix, precision=53)
    | next(self, rg, precision=53)
                           -----
    | Data descriptors defined here:
      __dict__
           dictionary for instance variables (if defined)
       weakref
           list of weak references to the object (if defined)
   class RandomGen(builtins.object)
    | RandomGen(rng=None)
    | A class that implements many methods for
     | random number generation and sampling. It takes
    | an underlying RNG as specified in the constructor.
      Methods defined here:
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init (self, rng=None)
            Initializes a new RandomGen instance.
            NOTES:
            1. Assumes that 'rng' implements
            a 'randint(a, b)' method that returns a random
            integer in the interval [a, b]. Currently, this
            class assumes 'a' is always \theta.
            2. 'rndint' (and functions that ultimately call it) may
be
            slower than desirable if many random numbers are
            needed at once. Ways to improve the performance
            of generating many random numbers at once include
            vectorization (which is often PRNG specific) and
multithreading
            (which is too complicated to show here).
       ball_point(self, dims, radius=1)
            Generates an independent and uniform random point inside
a 'dims'-dimensional
            ball (disc, solid sphere, etc.) centered at the origin.
       bernoulli(self, p)
            Returns 1 at probability p, 0 otherwise.
       beta(self, a, b, nc=0)
            Generates a beta-distributed random number.
            `a` and `b` are the two parameters of the beta
distribution.
            and `nc` is a parameter such that `nc` other than 0
            indicates a noncentral distribution.
       binomial(self, trials, p, n=None)
       binomial_int(self, trials, px, py)
       boundedGeometric(self, px, py, n)
            Generates a bounded geometric random number, defined
            here as the number of failures before the first success
(but no more than n),
            where the probability of success in
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each trial is px/py.
            Reference:
            Bringmann, K. and Friedrich, T., 2013, July. Exact and
efficient generation
            of geometric random variates and random graphs, in
            International Colloquium on Automata, Languages, and
            Programming (pp. 267-278).
       cauchy(self)
       choice(self, list)
       derangement(self, list)
            Returns a copy of list with each of its elements
            moved to a different position.
       derangement algorithm s(self, list)
            Returns a copy of 'list' with each of its elements
            moved to a different position (a derangement),
            but with the expected number of cycle lengths
            in probability, even though the list
            need not be a uniformly randomly
            chosen derangement. Uses importance sampling.
            Reference:
            J.R.G. Mendonça, "Efficient generation of
            random derangements with the expected
            distribution of cycle lengths", arXiv:1809.04571v4
            [stat.CO], 2020.
        derangement algorithm_t(self, list)
            Returns a copy of 'list' with each of its elements
            moved to a different position (a derangement),
            but with the expected number of cycle lengths
            in probability, even though the list
            need not be a uniformly randomly
            chosen derangement. Reference:
            J.R.G. Mendonça, "Efficient generation of
            random derangements with the expected
            distribution of cycle lengths", arXiv:1809.04571v4
            [stat.CO], 2020.
```

```
diceRoll(self, dice, sides=6, bonus=0)
        dirichlet(alphas)
       discretegen(self, probs)
            Generates a random integer in [0, n), where the
probability
            of drawing each integer is specified as a list
            of probabilities that sum to 1, where n is the
            number of probabilities. This method is optimal,
            or at least nearly so, in terms of the number of random
            bits required to generate the number
            on average. This method implements
            a solution to exercise 3.4.2 of chapter 15 of Luc
Devrove's
            Non-Uniform Random Variate Generation , 1986.
            - probs. List of probability objects, where for each
item
               in the probability list, the integer 'i' is chosen
               with probability 'probs[i]'.
               Each probability object provides access to a binary
               expansion of the probability, which must be a real
number in
               the interval [0, 1]. The binary expansion is a
sequence of zeros and ones
               expressed as follows: The first binary digit is the
half digit, the second
               is the quarter digit, the third is the one-eighth
digit,
               and so on. Note that any probability with a
terminating binary
               expansion (except 0) can be implemented by
"subtracting" 1
               from the expansion and then appending an infinite
sequence
               of ones at the end. The probability object must
implement the following
               three methods:
               - reset(): Resets the probability object to the first
digit in
                  the binary expansion.
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- nextbit(): Gets the next digit in the binary
expansion.
               - eof(): Gets whether the end of the binary expansion
was reached
                  (True or False), meaning the rest of the digits in
the expansion are
                  all zeros.
               The probability object will have to be mutable for
this method
               to work.
               The BinaryExpansion class is a convenient way to
express numbers
               as probability objects that meet these criteria.
     Т
Each probability object
               can also be a float, int, or Fraction in the interval
[0, 1].
       expoNumerator(self, denom)
            Generates the numerator of an exponential random
            number with a given denominator,
            using von Neumann's
            algorithm ("Various techniques used in connection with
            random digits", 1951).
        expoRatio(self, base, rx=1, ry=1)
            Generates an exponential random number
            (in the form of a ratio, or two-element list) given
            the rate `rx`/`ry` and the base `base`.
            The number will have the denominator `base*rx`.
        exponential(self, lamda=1.0)
        exprandfill(self, a, bits)
            Fills the unsampled bits of the given exponential random
number
            'a' as necessary to make a number whose fractional part
            has 'bits' many bits. If the number's fractional part
already has
            that many bits or more, the number is rounded using the
round-to-nearest,
            ties to even rounding rule. Returns the resulting
number as a
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multiple of 2^'bits'.
        exprandless(self, a, b)
            Determines whether one partially-sampled exponential
number
            is less than another; returns
            True if so and False otherwise. During
            the comparison, additional bits will be sampled in both
numbers
            if necessary for the comparison.
        exprandnew(self, lamdanum=1, lamdaden=1)
            Returns an object to serve as a partially-sampled
            exponential random number with the given
            rate 'lamdanum'/'lamdaden'. The object is a list of
five numbers:
            the first is a multiple of 1/(2^X), the second is X, the
third is the integer
            part (initially -1 to indicate the integer part wasn't
sampled yet),
            and the fourth and fifth are the lamda parameter's
            numerator and denominator, respectively. Default for
'lamdanum'
            and 'lamdaden' is 1.
            The number created by this method will be "empty"
            (no bits sampled yet).
       frechet(self, a, b, mu=0)
        fromDyadicDecompCode(self, code, precision=53)
            Generates a uniform random number contained in a box
described
                by the given universal dyadic decomposition code.
                - code: A list returned by the getDyadicDecompCode
                  or getDyadicDecompCodePdf method.
                - precision: Desired minimum precision in number of
binary digits
                  after the point. Default is 53.
            Reference: C.T. Li, A. El Gamal, "A Universal Coding
Scheme for
            Remote Generation of Continuous Random Variables",
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arXiv:1603.05238v1 [cs.IT], 2016.
     | gamma(self, mean, b=1.0, c=1.0, d=0.0)|
           Generates a random number following a gamma
distribution.
       gaussian copula(self, cov)
     | gbas(self, coin, k=385)
           Estimates the probability of heads of a coin. GBAS =
Gamma Bernoulli approximation scheme.
           The algorithm is simple to describe: "Flip a coin until
it shows heads
              k times. The estimated probability of heads is
then (k-1)/GammaDist(r, 1),
              where r is the total number of coin flips."
            The estimate is unbiased (multiple estimates average to
the true probability
           of heads) but has nonzero probability of being
           greater than 1 (that is, the estimate does not lie in
[0, 1] almost surely).
            Assumes the probability of heads is in the interval (0,
11.
            [[[NOTE: As can be seen in Feng et al., the following
are equivalent to the previous
           algorithm:
             Geometric: "Let G be 0. Do this _k_ times: 'Flip a
coin until it shows heads, let r be the number of flips (including
the last), and add GammaDist(r, 1) to G.' The estimated probability
              of heads is then (k-1)/G."
             Bernoulli: "Let G be 0. Do this until heads is shown
k times: 'Flip a coin and add Expo(1) to G.' The estimated
probability of heads is then (k-1)/G."
             Both algorithms use the fact that (k-1)/(X1+...+Xk) is
an unbiased estimator
             of p, namely 1 divided by the mean of an Expo(p)
random variable (X1, X2, ... Xk
             are i.i.d. Expo(p) random variates), with p>0. In the
same way, any algorithm to turn
             an endless sequence of random numbers with mean M into
k many i.i.d. Expo(M)
             random variates will work, as with the Poisson
```

```
distribution, for example.
              Note that GammaDist(r,1) is distributed as the sum of
_r_ many i.i.d. Expo(1) variates.]]]
           References: Huber, M., 2017. A Bernoulli mean estimate
with
               known relative error distribution. Random Structures
& Algorithms, 50(2),
               pp.173-182. (preprint in arXiv:1309.5413v2
[math.ST], 2015).
               Feng, J. et al. "Monte Carlo with User-Specified
Relative Error." (2016).
            coin: A function that returns 1 (or heads) with unknown
probability and 0 otherwise.
            k: Number of times the coin must return 1 (heads) before
the estimation
                stops.
                To ensure an estimate whose relative error's
absolute value exceeds
                epsilon with probability at most delta, calculate
     the smallest
                integer k such that:
                   gammainc(k,(k-1)/(1+epsilon)) +
                       (1 - gammainc(k,(k-1)/(1-epsilon))) \le delta
                (where gammainc is the regularized lower incomplete
gamma function,
                implemented, e.g., as scipy.special.gammainc), and
set this parameter
                to the calculated k value or higher.
                 The default is 385, which allows the relative
error to exceed 0.1 (epsilon) with
                  probability at most 0.05 (delta).
                  A simpler suggestion is
k>=ceiling(-6*ln(2/delta)/((epsilon**2)*(4*epsilon-3))).
                  For both suggestions, epsilon is in the interval
(0, 3/4) and delta is in (0, 1).
                  Note: "14/3" in the paper should probably read
"4/3".
     | gbas01(self, coin, k=385)
            Estimates the mean of a random variable lying in [0, 1].
            This is done using gbas and a "coin" that returns 1 if a
random uniform [0, 1]
```

```
otherwise.
            The estimate is unbiased but has nonzero probability of
beina
            greater than 1 (that is, the estimate does not lie in
[0, 1] almost surely).
            coin: A function that returns a number in [0, 1].
            k: See gbas.
        geoellipsoid point(self, a=6378.137, invf=298.2572236)
            Generates an independent and uniform random
            point on the surface of a geoellipsoid. The
            geoellipsoid uses the following parameters:
            a - semimajor axis (distance from the center of
               the geoellipsoid to the equator). The default
               is the WGS 84 ellipsoid's semimajor axis
               in kilometers.
            invf - inverse flattening. The default is the
              WGS 84 ellipsoid's inverse flattening.
       geometric(self, p)
        getDyadicDecompCode(self, point, f=None, fbox=None)
            Finds a code describing the position and size of a box
that covers the given
            point in the universal dyadic decomposition for random
number generation.
            - point: A list of coordinates of a point in space.
This method assumes
             the point was a randomly generated member of a
geometric set (such as a
              sphere, ellipse, polygon, or any other volume). Let N
be the number
             of coordinates of this parameter (the number of
dimensions).
            - f: A function that determines whether a point belongs
in the geometric set.
              Returns True if so, and False otherwise. This method
takes as input a list
              containing N coordinates describing a point in space.
If this parameter is
              given, this method assumes the geometric set is convex
```

number is less the result of the given function or 0

```
(and this method
             may return incorrect results for concave sets),
because the method checks
             only the corners of each box to determine whether the
box is entirely included
              in the geometric set.
            - fbox: A function that determines whether a box is
included
             in the geometric set. This method takes
             as input a list containing N items, where each item is
a list containing the
              lowest and highest value of the box for the
corresponding dimension. Returns 0 if the
             box is entirely outside the set, 1 if the box is
partially inside the set (or if the
             method is not certain whether the box is inside or
outside the set), and 2
             if the box is entirely inside the set.
            Returns a list containing two items. The first describes
the size of the box
            (as a negative power of 2). The second is a list of
coordinates describing the
            position. Let v be 2^{**}-ret[0]. The box is then
calculated as (ret[1][0]*v,
            ret[1]*v+v), ..., (ret[1][n-1]*v, ret[1][n-1]*v+v).
            Raises an error if the point was determined not to
belong in the geometric set.
            Either f or fset must be passed to this method, but not
both.
            Reference: C.T. Li, A. El Gamal, "A Universal Coding
Scheme for
            Remote Generation of Continuous Random Variables",
            arXiv:1603.05238v1 [cs.IT], 2016.
     getDyadicDecompCodePdf(self, point, pdf=None,
pdfbounds=None, precision=53)
            Finds a code describing the position and size of a box
that covers the given
            point in the universal dyadic decomposition for random
number generation,
```

based on a non-uniform probability density function. It

generates a
random number for this purpose, so the return value may
differ from call to
call.
 point: A list of coordinates of a point in space.
This method assumes
the point was random generated and within the support
of a continuous
distribution with a PDF. Let N be the number of
coordinates of this parameter
(the number of dimensions).
<pre>- pdf: The probability density function (PDF) of the</pre>
continuous distribution.
This method takes as input a list
<pre>containing N coordinates describing a point in space,</pre>
and returns the probability
density of that point as a single number. If this
parameter is given, however:
 This method assumes the PDF is unimodal and strictly
decreasing in every direction away from the PDF's mode, and may
return incorrect results if that is not the case.
 If the given PDF outputs floating-point numbers, the
resulting
dyadic decomposition code may be inaccurate due to
rounding errors.
 pdfbounds: A function that returns the lower and upper
bounds of the PDF's value
at a box. This method takes as input a list containing
N items, where each item
is a list containing the lowest and highest value of
the box for the
corresponding dimension. Returns a list
containing two items: the lower bound and the upper
bound, respectively, of the
PDF anywhere in the given box. If this parameter is
given, this method assumes the PDF is continuous
almost everywhere and bounded
<pre>from above; the dyadic decomposition will generally</pre>
work only if that is the case.
 - precision: Precision of random numbers generated by
this method, in binary digits
after the point. Default is 53.

```
Returns a list containing two items. The first describes
the size of the box
            (as a negative power of 2). The second is a list of
coordinates describing the
            position. Let v be 2^{**}-ret[0]. The box is then
calculated as (ret[1][0]*v,
            ret[1]*v+v), ..., (ret[1][n-1]*v, ret[1][n-1]*v+v).
            Raises an error if the point is determined to be outside
the support of the PDF.
            Either pdf or pdfbounds must be passed to this method,
but not both.
            Reference: C.T. Li, A. El Gamal, "A Universal Coding
Scheme for
            Remote Generation of Continuous Random Variables",
            arXiv:1603.05238v1 [cs.IT], 2016.
     | gumbel(self, a, b)
       hypercube_point(self, dims, sizeFromCenter=1)
            Generates an independent and uniform random point on the
surface of a 'dims'-dimensional
            hypercube (square, cube, etc.)
            centered at the origin.
       hypergeometric(self, trials, ones, count)
       hypersphere point(self, dims, radius=1)
            Generates an independent and uniform random point on the
surface of a 'dims'-dimensional
            hypersphere (circle, sphere, etc.)
            centered at the origin.
     integersWithSum(self, n, total)
            Returns a list of 'n' integers 0 or greater that sum to
'total'.
            The combination is chosen uniformly at random among all
            possible combinations.
       integers_from_pdf(self, pdf, mn, mx, n=1)
            Generates one or more random integers from a discrete
probability
```

```
distribution expressed as a probability density
            function (PDF), which is also called the probability
mass
            function for discrete distributions. The random
integers
            will be in the interval [mn, mx]. `n` random integers
     1
will be
            generated. `pdf` is the PDF; it takes one parameter and
     1
returns,
            for that parameter, a weight indicating the relative
probability
            that a random integer will equal that parameter.
            The area under the "curve" of the PDF need not be 1.
            By default, `n` is 1.
       integers from u01(self, u01, pmf)
            Transforms one or more random numbers into numbers
            (called quantiles) that
            follow a discrete distribution, assuming the
distribution
                  produces only integers 0 or greater.
                  - `u01` is a list of uniform random numbers, in
[0, 1].
                  - `pmf` is the probability mass function (PMF)
                  of the discrete distribution; it takes one
parameter and returns,
                  for that parameter, the probability that a random
number is
                  equal to that parameter (each probability is in
the interval [0, 1]).
                 The area under the PMF must be 1; it
                  is not enough for the PMF to be correct up to a
constant.
     intsInRangeSortedWithSum(self, numSamples, numPerSample, mn,
mx, sum)
            Generates one or more combinations of
             'numPerSample' numbers each, where each
             combination's numbers sum to 'sum' and are listed
             in sorted order, and each
             number is in the interval '[mn, mx]'.
             The combinations are chosen uniformly at random.
```

```
'mn', 'mx', and
             'sum' may not be negative. Returns an empty
             list if 'numSamples' is zero.
              The algorithm is thanks to a Stack Overflow
            answer (`questions/61393463`) by John McClane.
            Raises an error if there is no solution for the given
            parameters.
        intsInRangeWithSum(self, numSamples, numPerSample, mn, mx,
sum)
            Generates one or more combinations of
             'numPerSample' numbers each, where each
             combination's numbers sum to 'sum' and are listed
             in any order, and each
             number is in the interval '[mn, mx]'.
             The combinations are chosen uniformly at random.
                 'mn', 'mx', and
             'sum' may not be negative. Returns an empty
             list if 'numSamples' is zero.
              The algorithm is thanks to a _Stack Overflow_
            answer (`questions/61393463`) by John McClane.
            Raises an error if there is no solution for the given
            parameters.
        intsInRangesWithSum(self, numSamples, ranges, total)
            Generates one or more combinations of
             'len(ranges)' numbers each, where each
             combination's numbers sum to 'total', and each number
             has its own valid range. 'ranges' is a list of valid
ranges
             for each number; the first item in each range is the
minimum
             value and the second is the maximum value. For
example,
             'ranges' can be [[1,4],[3,5],[2,6]], which says that
the first
             number must be in the interval [1, 4], the second in
     [3, 5],
             and the third in [2, 6].
             The combinations are chosen uniformly at random.
                 Neither the integers in the 'ranges' list nor
             'total' may be negative. Returns an empty
```

```
list if 'numSamples' is zero.
      This is a modification I made to an algorithm that
        was contributed in a _Stack Overflow_
    answer (`questions/61393463`) by John McClane.
    Raises an error if there is no solution for the given
    parameters.
kth smallest of n u01(self, k, n)
    Generates the kth smallest number among n random numbers
    in the interval [0, 1].
kthsmallest(self, n, k, b)
    Generates the 'k'th smallest 'b'-bit uniform random
    number out of 'n' of them.
kthsmallest psrn(self, n, k)
    Generates the 'k'th smallest 'b'-bit uniform random
    number out of 'n' of them; returns the result in
    the form of a uniform partially-sampled random number.
latlon(self)
    Generates an independent and uniform random latitude and
    longitude, in radians. West and south coordinates
    are negative.
lognormal(self, mu=0.0, sigma=0.0)
lower bound copula(self)
mcmc(self, pdf, n)
    Generates 'n' random numbers that follow
    the probability density given in 'pdf' using
    a Markov-chain Monte Carlo algorithm, currently
    Metropolis--Hastings. The resulting random numbers
    are not independent, but are often close to
    being independent. 'pdf' takes one number as
    a parameter and returns a number 0 or greater.
    The area under the curve (integral) of 'pdf'
    need not be equal to 1.
mcmc2(self, pdf, n)
    Generates 'n' pairs of random numbers that follow
```

```
the probability density given in 'pdf' using
            a Markov-chain Monte Carlo algorithm, currently
            Metropolis--Hastings. The resulting random pairs
            are not independent, but are often close to
            being independent. 'pdf' takes one parameter,
            namely, a list of two numbers giving a sampled
            point and returns a number 0 or greater.
            The volume under the surface (integral) of 'pdf'
            need not be equal to 1.
       monte carlo integrate(self, func, bounds, samples=1000)
            Estimates the integral (volume) of a function within the
            given bounds using Monte Carlo integration, which
generates
            an estimate using the help of randomization.
            func - Function to integrate. Takes the same number
               of parameters as the length of bounds.
            bounds - Bounds of integration at each dimension.
               An N-length array of arrays. Each array in turn
               contains two items: the lower bound and upper bound
               for that dimension.
            samples - Number of times to sample the bounds of
               integration randomly. The default is 1000 samples.
            Returns an array containing two items: the estimated
            integral and the standard error.
       moyal(self, mu=0, sigma=1)
            Sample from a Moyal distribution, using the
            method given in C. Walck, "Handbook on
            Statistical Distributions for Experimentalists",
            pp. 93-94.
        multinomial(self, trials, weights)
       multinormal(self, mu, cov)
       multinormal_n(self, mu, cov, n=1)
        multipoisson(self, firstmean, othermeans)
            Multivariate Poisson distribution (as found in
Mathematica).
```

```
multivariate t(self, mu, cov, df)
            Multivariate t-distribution, mu is the mean (can be
None),
            cov is the covariance matrix, and df is the degrees of
freedom.
        negativeMultinomial(self, succ, failures)
            Negative multinomial distribution.
            Models the number of failures of one or more
            kinds before a given number of successes happens.
            succ: Number of successes.
            failures: Contains probabilities for each kind of
failure.
            The sum of probabilities must be less than 1.
            Returns: A list containing a random number
            of failures of each kind of failure.
       negativebinomial(self, successes, p)
       negativebinomialint(self, successes, px, py)
            Generates a negative binomial random number, defined
            here as the number of failures before 'successes' many
            successful trials, where the probability of success in
            each trial is px/py.
       nonzeroIntegersWithSum(self, n, total)
            Returns a list of 'n' integers greater than 0 that sum
to 'total'.
            The combination is chosen uniformly at random among all
            possible combinations.
       normal(self, mu=0.0, sigma=1.0)
            Generates a normally-distributed random number.
       numbersWithSum(self, count, sum=1.0)
        numbers from cdf(self, cdf, mn, mx, n=1)
            Generates one or more random numbers from a non-discrete
probability
            distribution by numerically inverting its cumulative
            distribution function (CDF).
```

```
- cdf: The CDF; it takes one parameter and returns,
            for that parameter, the probability that a random number
will
            be less than or equal to that parameter.
            - mn, mx: Sampling domain. The random number
            will be in the interval [mn, mx].
            - n: How many random numbers to generate. Default is 1.
        numbers from dist(self, pdf, mn=0, mx=1, n=1, bitplaces=53)
            Generates 'n' random numbers that follow a continuous
            distribution in an interval [mn, mx]. The distribution
must have a
            PDF (probability density function) and the PDF must be
less than or equal to a finite number and be continuous almost
everywhere
            in the interval. Implements section 4 of Devroye and
Gravel,
            "The expected bit complexity of the von Neumann
rejection
            algorithm", arXiv:1511.02273v2 [cs.IT], 2016.
     Ι
            - 'n' is the number of random numbers to generate.
Default is 1.
            - 'pdf' is a procedure that takes three arguments: xmin,
xmax, bitplaces,
               and returns an array of two items: the greatest lower
bound of f(x) anywhere
               in the interval [xmin, xmax] (where f(x) is the PDF),
and the least upper
               bound of f(x) anywhere there. Both bounds are
multiples of 2^-bitplaces.
            - 'bitplaces' is an accuracy expressed as a number of
bits after the
               binary point. The random number will be a multiple of
2^-bitplaces,
               or have a smaller granularity. Default is 53.
            - 'mn' and 'mx' express the interval. Both are optional
and
               are set to 0 and 1, respectively, by default.
       numbers from dist inversion(self, icdf, n=1, digitplaces=53,
base=2)
```

```
Generates 'n' random numbers that follow a discrete or
non-discrete
            probability distribution, using the inversion method.
            Implements section 5 of Devroye and Gravel,
            "Sampling with arbitrary precision", arXiv:1502.02539v5
[cs.IT], 2015.
            - 'n' is the number of random numbers to generate.
Default is 1.
            - 'icdf' is a procedure that takes three arguments: u,
ubits, digitplaces,
               and returns a number within base^-digitplaces of the
True inverse
               CDF (inverse cumulative distribution function, or
quantile function)
               of u/base^ubits. For a given value of `digitplaces`,
icdf(x)<=icdf(y)</pre>
               whenever 0 <= x < y <= 1.
            - 'digitplaces' is an accuracy expressed as a number of
digits after the
               point. Each random number will be a multiple of
base^-digitplaces,
               or have a smaller granularity. Default is 53.
            - base is the digit base in which the accuracy is
expressed. Default is 2
               (binary). (Note that 10 means decimal.)
     numbers from pdf(self, pdf, mn, mx, n=1, steps=100)
            Generates one or more random numbers from a continuous
probability
            distribution expressed as a probability density
            function (PDF). The random number
            will be in the interval [mn, mx]. `n` random numbers
will be
            generated. `pdf` is the PDF; it takes one parameter and
     1
returns,
            for that parameter, a weight indicating the relative
probability
             that a random number will be close to that parameter.
`steps`
            is the number of subintervals between sample points of
     1
the PDF.
            The area under the curve of the PDF need not be 1.
```

```
By default, `n` is 1 and `steps` is 100.
       numbers from u01(self, u01, pdf, cdf, mn, mx, ures=None)
            Transforms one or more random numbers into numbers
            (called quantiles) that follow a non-discrete
probability distribution, based on its PDF
            (probability density function) and/or its CDF
(cumulative distribution
            function).
            - u01: List of uniform random numbers in [0, 1] that
will be
            transformed into numbers that follow the distribution.
            - pdf: The PDF; it takes one parameter and returns,
            for that parameter, the relative probability that a
            random number close to that number is chosen. The area
under
            the PDF need not be 1 (this method works even if the PDF
            is only known up to a normalizing constant). Optional if
a CDF is given.
            - cdf: The CDF; it takes one parameter and returns,
            for that parameter, the probability that a random number
will
            be less than or equal to that parameter. Optional if a
PDF is given.
            For best results, the CDF should be
            strictly increasing everywhere in the
            interval [xmin, xmax] and must output values in [0, 1];
            for best results, the CDF should
            be increasing everywhere in [xmin, xmax].
            - mn, mx: Sampling domain. The random number
            will be in the interval [mn, mx]. For best results,
            the range given by mn and mx should cover all or
            almost all of the distribution.
            - ures - Maximum approximation error tolerable, or
            "u-resolution". Default is 10^-8. The underlying
sampler's approximation
            error will generally be less than this tolerance, but
this is not guaranteed.
            Currently used only if a
            PDF is given.
```

```
pareto(self, minimum, alpha)
       partialshuffle(self, list, k)
           Does a partial shuffle of
            a list's items (stops when 'k' items
           are shuffled); the shuffled items
            will appear at the end of the list.
           Returns 'list'.
       piecewise_linear(self, values, weights)
       piecewise_linear_n(self, values, weights, n=1)
       poisson(self, mean)
           Generates a random number following a Poisson
distribution.
     poissonint(self, mx, my)
           Generates a random number following a Poisson
distribution with mean mx/my.
     polya int(self, sx, sy, px, py)
           Generates a negative binomial (Polya) random number,
defined
           here as the number of failures before 'successes' many
            successful trials (sx/sy), where the probability of
success in
            each trial is px/py.
     powerlognormal(self, p, sigma=1.0)
            Power lognormal distribution, as described in
NIST/SEMATECH
           e-Handbook of Statistical Methods,
[http://www.itl.nist.gov/div898/handbook/,]
(http://www.itl.nist.gov/div898/handbook/,)
            accessed Jun. 9, 2018, sec. 1.3.6.6.14.
     powernormal(self, p)
            Power normal distribution, as described in NIST/SEMATECH
            e-Handbook of Statistical Methods,
[http://www.itl.nist.gov/div898/handbook/,]
(http://www.itl.nist.gov/div898/handbook/,)
```

```
accessed Jun. 9, 2018, sec. 1.3.6.6.13.
       product_copula(self, n=2)
       randbit(self)
       randbits(self, n)
            Generates an n-bit random integer.
       randomwalk_posneg1(self, n)
            Random walk of uniform positive and negative steps.
       randomwalk u01(self, n)
            Random walk of uniform 0-1 random numbers.
       rayleigh(self, a)
            Generates a random number following a Rayleigh
distribution.
       rndint(self, maxInclusive)
       rndint fastdiceroller(self, maxInclusive)
       rndintexc(self, maxExclusive)
        rndintexcrange(self, minInclusive, maxExclusive)
        rndintrange(self, minInclusive, maxInclusive)
        rndrange(self, minInclusive, maxInclusive)
        rndrangemaxexc(self, minInclusive, maxExclusive)
       rndrangeminexc(self, mn, mx)
        rndrangeminmaxexc(self, mn, mx)
       rndu01(self)
       rndu01oneexc(self)
        rndu01zeroexc(self)
```

```
rndu01zerooneexc(self)
       sample(self, list, k)
       sattolo(self, list)
            Puts the elements of 'list' in random order, choosing
            from among all cyclic permutations (Sattolo's
algorithm).
            Returns 'list'.
       shell_point(self, dims, outerRadius=1, innerRadius=0.5)
            Generates an independent and uniform random point inside
a 'dims'-dimensional
            spherical shell (donut, hollow sphere, etc.)
            centered at the origin.
       shuffle(self, list)
            Puts the elements of 'list' in random order (does an
            in-place shuffle). Returns 'list'.
        simplex point(self, points)
            Generates an independent and uniform random point on the
surface of an N-dimensional
            simplex (line segment, triangle, tetrahedron, etc.)
            with the given coordinates.
       slicesample(self, pdf, n, xstart=0.1)
            Slice sampling of R. M. Neal.
            Generates 'n' random numbers that follow
            the probability density given in 'pdf' using
            slice sampling. The resulting random numbers
            are not independent, but are often close to
              being independent. 'pdf' takes one number as
              a parameter and returns a number 0 or greater.
              The area under the curve (integral) of 'pdf'
              need not be equal to 1. 'xstart' should be
            chosen such that `pdf(xstart)>0`.
       spsa_minimize(self, func, guess, iterations=200,
constrain=None, a=None, c=None, acap=None)
            Tries to find a choice of parameters that minimizes the
```

```
value
            of a scoring function, also called the objective
function or loss
            function, starting from an initial guess. This method
uses an
            algorithm called "simultaneous perturbation
            stochastic approximation", which is a randomized
            search for the minimum value of the objective function.
            func - Objective function, a function that calculates a
score for the
             given array of parameters and returns that score. The
score is a
             single number; the lower the score, the better.
             The score can be negative. (Note that the problem of
maximizing
             the score is the same as minimizing it except
             that the score's sign is reversed at the end.)
            guess - Initial guess for the best choice of parameters.
This is an
             array of parameters, each of which is a number. This
array has
             as many items as the array passed to 'func'.
            iterations - Maximum number of iterations in which to
run the
             optimization process. Default is 200.
            constrain - Optional. A function that takes the given
array of
             parameters and constrains them to fit the bounds of a
valid
             array of parameters. This function modifies the array
in place.
            a - Optional. A setting used in the optimization
process; greater than 0.
            c - Optional. A setting used in the optimization
     1
process; greater than 0. As a guideline,
              'c' is about equal to the "standard deviation of the
measurement noise"
              for several measurements at the initial guess, and is
a "small positive
              number" if measurements are noise-free (Spall 1998).
Default
             is 0.001.
```

```
acap - Optional. A setting used in the optimization
process; an
              integer greater than 0.
     | stable(self, alpha, beta)
            Generates a random number following a stable
distribution.
     stableO(self, alpha, beta, mu=0, sigma=1)
            Generates a random number following a 'type 0' stable
distribution.
     | surface point(self, f, bounds, ngrad, gmax)
            Generates a uniform random point on
               a parametric surface, using a rejection
               approach developed by Williamson, J.F.,
               "Random selection of points distributed on
                curved surfaces", Physics in Medicine & Biology
32(10), 1987.
            - f: Takes two parameters (u and v) and returns
              a 3-element array expressing
              a 3-dimensional position at the given point.
            - bounds: Two 2-element arrays expressing bounds
              for u and v. Of the form [[umin, umax], [vmin,
              vmax]].
            - ngrad: Takes two parameters (u and v) and returns
              the norm of the gradient (stretch factor)
              at the given point. Can be None, in which
              the norm-of-gradient is calculated numerically.
            - gmax: Maximum norm-of-gradient
              for entire surface.
        t copula(self, cov, df)
            Multivariate t-copula. 'cov' is the covariance matrix
            and 'df' is the degrees of freedom.
       triangular(self, startpt, midpt, endpt)
        truncnormal(randgen, a, b)
            Samples from a truncated normal distribution in [a, b];
this method is
            designed to sample from either tail of that
```

```
distribution.
            Reference:
            Botev, Z. and L'Ecuyer, P., 2019. Simulation from the
Tail of the
           Univariate and Multivariate Normal Distribution. In
Systems
           Modeling: Methodologies and Tools (pp. 115-132).
     Ι
Springer, Cham.
       upper bound copula(self, n=2)
     vonmises(self, mean, kappa)
     | weibull(self, a, b)
            Generates a Weibull-distributed random number.
      weighted choice(self, weights)
     weighted_choice_inclusion(self, weights, n)
            Chooses a random sample of `n` indices from a list of
items (whose weights are given as `weights`), such that the chance
that index `k` is in the sample is given as
\weights[k]*n/Sum(weights). It implements the splitting method
referenced below.
            Deville, J.-C. and Tillé, Y. Unequal probability
sampling without replacement through a splitting method. Biometrika
85 (1998).
       weighted choice n(self, weights, n=1)
     wiener(self, st, en, step=1.0, mu=0.0, sigma=1.0)
            Generates random numbers following a Wiener
            process (Brownian motion). Each element of the return
            value contains a timestamp and a random number in that
order.
       zero or one(self, px, py)
            Returns 1 at probability px/py, 0 otherwise.
       zero or one exp minus(self, x, y)
```

```
Generates 1 with probability exp(-px/py); 0 otherwise.
           Reference:
           Canonne, C., Kamath, G., Steinke, T., "The Discrete
Gaussian
         for Differential Privacy", arXiv:2004.00010 [cs.DS],
   2020.
    | zero_or_one_power(self, px, py, n)
           Generates 1 with probability (px/py)^n (where n can be
positive, negative, or zero); 0 otherwise.
    zero_or_one_power_ratio(self, px, py, nx, ny)
          Generates 1 with probability (px/py)^(nx/ny) (where
nx/ny can be positive, negative, or zero); 0 otherwise.
    _____
    | Data descriptors defined here:
    | __dict__
           dictionary for instance variables (if defined)
    | __weakref__
          list of weak references to the object (if defined)
    | Data and other attributes defined here:
    | FPPRECISION = 53
    | FPRADIX = 2
      MINEXPONENT = -1074
   class RatioOfUniformsTiling(builtins.object)
    RatioOfUniformsTiling(pdf, mode=0, y0=-10, y1=10, cycles=8)
    | Produces a tiling for the purposes
           of fast sampling from a probability distribution via
the
          ratio of uniforms method.
```

```
- pdf: The probability density function (PDF); it takes one
parameter and returns,
           for that parameter, the relative probability that a
           random number close to that number is chosen. The area
under
           the PDF need not be 1; this method works even if the PDF
           is only known up to a normalizing constant, and even if
           the distribution has infinitely extending tails to the
left and/or right.
          However, for the ratio of uniforms method to work, both
pdf(x) and
          x*x*pdf(x) must be less than or equal to a finite number
(thus, if the distribution has
          tails, they must drop off at a faster than quadratic
rate).
     - mode: X-coordinate of the PDF's highest peak or one of
them,
          or a location close to it. Optional; default is 0.
       - y0, y1: Bounding coordinates for the ratio-of-uniforms
tilina.
           For this class to work, y0 \le min(x*sqrt(pdf(x))) and
          y1 >= max( x*sqrt(pdf(x)) ) for every x. Optional; the
default is y0=-10, y1=10.
     - cycles - Number of recursion cycles in which to split
tiles
           for the ratio-of-uniforms tiling. Default is 8.
       Additional improvements not yet implemented:
        Generalized ratio-of-uniforms in Hörmann et al., "Automatic
        Nonuniform Random Variate Generation", 2004.
        References:
         Section IV.7 of Devroye, L., "Non-Uniform Random Variate
Generation", 1986.
        Section 4.5 of Fulger, D., "From phenomenological modelling
of anomalous
        diffusion through continuous-time random walks and
fractional
        calculus to correlation analysis of complex systems",
dissertation,
        Philipps-Universität Marburg, 2009.
```

```
Methods defined here:
       init (self, pdf, mode=0, y0=-10, y1=10, cycles=8)
            Initialize self. See help(type(self)) for accurate
signature.
       codegen(self, name, pdfcall=None)
            Generates Python code that samples
                    (approximately) from the distribution estimated
                   in this class. Idea from Leydold, et al.,
                    "An Automatic Code Generator for
                   Nonuniform Random Variate Generation", 2001.
            - name: Distribution name. Generates a Python method
called
              sample X where X is the name given here (samples one
               random number).
            - pdfcall: Name of the method representing pdf (for more
information,
              see the __init__ method of this class). Optional; if
not given
              the name is pdf X where X is the name given in the
name parameter.
       maybeAppend(self, newtiles, xmn, xmx, ymn, ymx, depth=0)
       sample(self, rg, n=1)
           Generates random numbers that (approximately) follow the
                 distribution modeled by this class.
            - n: The number of random numbers to generate.
            Returns a list of 'n' random numbers.
       svg(self)
     | Data descriptors defined here:
       dict
           dictionary for instance variables (if defined)
       weakref
```

```
list of weak references to the object (if defined)
   class SortedAliasMethod(builtins.object)
      SortedAliasMethod(p)
    | Implements a weighted sampling table
     | where each weight must be in sorted
    | order (ascending or descending).
    | When many entries are in the table,
    | the initialization is faster than with
     | FastLoadedDiceRoller or VoseAlias. Reference:
    | K. Bringmann and K. Panagiotou, "Efficient Sampling
    | Methods for Discrete Distributions." In: Proc. 39th
     | International Colloquium on Automata, Languages,
      and Programming (ICALP'12), 2012.
       - p: List of weights, in sorted order (ascending or
           descending).
      Methods defined here:
       __init__(self, p)
           Initialize self. See help(type(self)) for accurate
signature.
      next(self, rg)
      _____
    | Data descriptors defined here:
    | dict
           dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
   class VoseAlias(builtins.object)
    | VoseAlias(weights)
    | Implements Vose's version of the alias sampler, which
chooses a random number in [0, n)
    | where the probability that each number is chosen is
```

```
weighted. The 'weights' is the
     | list of weights each 0 or greater; the higher the weight,
the greater
     | the probability. This sampler supports integer or non-
integer weights.
     | Reference:
     | Vose, Michael D. "A linear algorithm for generating random
numbers with a given
     | distribution." IEEE Transactions on software engineering 17,
no. 9 (1991): 972-975.
     | Methods defined here:
     | __init__(self, weights)
           Initialize self. See help(type(self)) for accurate
signature.
     | next(self, randgen)
              -----
     | Data descriptors defined here:
     | __dict__
           dictionary for instance variables (if defined)
     | __weakref__
         list of weak references to the object (if defined)
FUNCTIONS
   numericalTable(func, x, y, n=100)
DATA
   ArcTanHTable = [0, 294906490, 137123709, 67461703, 33598225,
16782680,...
   CRUDELOG = [0, -726816, -681390, -654818, -635964, -621340,
-609392, -...
   CRUDELOG ARCTANBITDIFF = 13
   CRUDELOG ARCTANFRAC = 29
   CRUDELOG BITS = 16
   CRUDELOG LOG2BITS = 45426
```

```
CRUDELOG LOGMIN = 9830
    LNPOLY2 = [(-28986367995118693...8591117027361355259,
10000000000000000...
    LNPOLY3 = [(-13476514299119388...8263005361644498323,
50000000000000000...
    REALHALFPI = RealPi(1/2)
    REALPI = RealPi(1)
    REAL 858 1000 = RealFraction(429/500)
FILE
/home/peter/Documents/SharpDevelopProjects/peteroupc.github.io/randor
4
Help on module fixed:
NAME
    fixed
CLASSES
    builtins.object
        Fixed
    class Fixed(builtins.object)
     | Fixed(i)
     | Fixed-point numbers, represented using integers that store
multiples
     | of 2^-BITS. They are not necessarily faster than floating-
point numbers, nor
     do they necessarily have the same precision or resolution of
floating-point
     \ensuremath{\mid} numbers. The main benefit of fixed-point numbers is that
they improve
     | determinism for applications that rely on non-integer real
numbers (notably
     | simulations and machine learning applications), in the sense
that the operations
     | given here deliver the same answer for the same input across
computers,
     whereas floating-point numbers have a host of problems that
make repeatable
```

```
| results difficult, including differences in their
implementation, rounding
     | behavior, and order of operations, as well as
nonassociativity of
     | floating-point numbers.
     The operations given here are not guaranteed to be
"constant-time"
     | (non-data-dependent and branchless) for every relevant
input.
     Any copyright to this file is released to the Public Domain.
In case this is not
     | possible, this file is also licensed under Creative Commons
Zero version 1.0.
     | Methods defined here:
     | __abs__(self)
     | __add__(a, b)
     | __cmp__(self, other)
     | __div__(a, b)
      __eq__(self, other)
           Return self==value.
      float (a)
     __floordiv__(a, b)
      __ge__(self, other)
           Return self>=value.
       __gt__(self, other)
           Return self>value.
       __init__(self, i)
           Initialize self. See help(type(self)) for accurate
signature.
```

```
__int__(a)
        __le__(self, other)
           Return self<=value.
       __lt__(self, other)
           Return self<value.
       __mod__(a, b)
       __mul__(a, b)
       ne (self, other)
           Return self!=value.
      __neg__(self)
     __pos__(self)
       __rdiv__(a, b)
       __repr__(self)
           Return repr(self).
       __rtruediv__(a, b)
       __str__(self)
          Return str(self).
     | __sub__(a, b)
       __truediv__(a, b)
     | acos(a)
           Calculates an approximation of the inverse cosine of the
given number.
     | asin(a)
           Calculates an approximation of the inverse sine of the
given number.
```

```
| atan2(y, x)
           Calculates the approximate measure, in radians, of the
angle formed by the
           X axis and a line determined by the origin and the given
coordinates of a 2D
            point. This is also known as the inverse tangent.
     | cos(a)
           Calculates the approximate cosine of the given angle;
the angle is in radians.
           For the fraction size used by this class, this method is
accurate to within
            1 unit in the last place of the correctly rounded result
     Т
for every input
           in the range [-pi*2, pi*2].
           This method's accuracy decreases beyond that range.
     | exp(a)
           Calculates an approximation of e (base of natural
logarithms) raised
           to the power of this number. May raise an error if this
     number
           is extremely high.
       floor(a)
     | log(a)
           Calculates an approximation of the natural logarithm of
this number.
     | pow(a, b)
           Calculates an approximation of this number raised to the
power of another number.
       round(a)
     | sin(a)
           Calculates the approximate sine of the given angle; the
angle is in radians.
            For the fraction size used by this class, this method is
accurate to within
            1 unit in the last place of the correctly rounded result
```

for every	
	in the range [-pi*2, pi*2].
	This method's accuracy decreases beyond that range.
sq	rt(a)
	Calculates an approximation of the square root of the
given numb	er.
1	
ta	n(a)
	Calculates the approximate tangent of the given angle;
the angle	is in radians.
1	For the fraction size used by this class, this method is
accurate t	
	2 units in the last place of the correctly rounded
result for	every input
I	in the range [-pi*2, pi*2].
i	This method's accuracy decreases beyond that range.
i I	
' 	
'	
l St	atic methods defined here:
1	dete methods derined herer
	i)
V(.	Converts a string, integer, Decimal, or other number
typo into	converts a string, integer, becimat, or other number
type into	a fixed-point number. If the parameter is a Fixed,
returns it	
returns it	
 	If the given number is a non-integer, returns the
closest va	
<u> </u>	a Fixed after rounding using the round-to-nearest-ties-
to-even	
	rounding mode. The parameter is recommended to be a
string	
	or integer, and is not recommended to be a `float`.
Da	ta descriptors defined here:
I	dict
	dictionary for instance variables (if defined)
İ	

```
| weakref
           list of weak references to the object (if defined)
     | Data and other attributes defined here:
     | ArcTanBitDiff = 9
     | ArcTanFrac = 29
     | ArcTanHTable = [0, 294906490, 137123709, 67461703, 33598225,
16782680,...
     | ArcTanTable = [421657428, 248918914, 131521918, 66762579,
33510843, 16...
     \mid BITS = 20
     | ExpK = 648270061
     | HALF = 524288
     | HalfPiArcTanBits = 843314856
       HalfPiBits = 1647099
       HalfPiHighRes = 130496653328243011213339889301986179
       HighResFrac = 116
       Ln2ArcTanBits = 372130559
      Log2Bits = 726817
       LogMin = 157286
     | MASK = 1048575
       PiAndHalfHighRes = 391489959984729033640019667905958538
       PiArcTanBits = 1686629713
```

```
PiBits = 3294199
       PiHighRes = 260993306656486022426679778603972359
       QuarterPiArcTanBits = 421657428
       SinCosK = 326016435
       TwoTimesPiArcTanBits = 3373259426
       TwoTimesPiBits = 6588397
       TwoTimesPiHighRes = 521986613312972044853359557207944718
       hash = None
FILE
/home/peter/Documents/SharpDevelopProjects/peteroupc.github.io/fixed
Help on module bernoulli:
NAME
    bernoulli
CLASSES
    builtins.object
        Bernoulli
        DiceEnterprise
    class Bernoulli(builtins.object)
     | This class contains methods that generate Bernoulli random
numbers,
     1
           (either 1 or heads with a given probability, or 0 or
tails otherwise).
          This class also includes implementations of so-called
"Bernoulli factories", algorithms
     | that sample a new probability given a coin that shows heads
with an unknown probability.
     | Written by Peter 0.
```

```
| References:
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machines and numbers",
     | arXiv:0906.5560v2 [math.PR], 2010.
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local correctness",
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mean problems",
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Roberts, G.O., "Simulating
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Bernstein
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Wendland. "From the Bernoulli factory to a dice enterprise via
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Rad Niazadeh. "Bernoulli factories and black-box reductions in
mechanism design." Journal of the ACM (JACM) 68, no. 2 (2021): 1-30.
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(2017). Exact Monte
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processes.
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G. O. Efficient
     | Bernoulli factory MCMC for intractable posteriors,
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for certain

```
| functions that can be expressed as power series." Stochastic
Processes and their
     | Applications 129, no. 11 (2019): 4366-4384.
     - Canonne, C., Kamath, G., Steinke, T., "The Discrete
Gaussian
     for Differential Privacy", arXiv:2004.00010 [cs.DS], 2020.
     - Lee, A., Doucet, A. and Łatuszyński, K., 2014. Perfect
simulation using
     | atomic regeneration with application to Sequential Monte
Carlo,
     | arXiv:1407.5770v1 [stat.CO]
     | Methods defined here:
     | __init__(self)
           Creates a new instance of the Bernoulli class.
     | a bag div b bag(selfnumerator, numbag, intpart, bag)
            Simulates (numerator+numbag)/(intpart+bag).
     | a div b bag(self, numerator, intpart, bag)
           Simulates numerator/(intpart+bag).
     | add(self, f1, f2, eps=Fraction(1, 20))
           Addition Bernoulli factory: B(p), B(q) \Rightarrow B(p+q) (Dughmi
et al. 2021)
           - f1, f2: Functions that return 1 if heads and 0 if
tails.
           - eps: A Fraction in (0, 1). eps must be chosen so that
p+q \ll 1 - eps,
             where p and q are the probability of heads for f1 and
f2, respectively.
     | alt_series(self, f, series)
           Alternating-series Bernoulli factory: B(p) -> B(s[0] -
s[1]*p + s[2]*p^2 - ...
           (Łatuszyński et al. 2011).
            - f: Function that returns 1 if heads and 0 if tails.
            - series: Object that generates each coefficient of the
series starting with the first.
              Each coefficient must be less than or equal to the
previous and all of them must
```

```
be 1 or less.
              Implements the following two methods: reset() resets
the object to the first
              coefficient; and next() generates the next
coefficient.
     | arctan n div n(self, f)
            Arctan div N: B(p) \rightarrow B(arctan(p)/p). Uses a uniformly-
fast special case of
            the two-coin Bernoulli factory, rather than the even-
parity construction in
            Flajolet's paper, which does not have bounded expected
running time for all heads probabilities.
            Reference: Flajolet et al. 2010.
             - f: Function that returns 1 if heads and 0 if tails.
     | bernoulli_x(self, f, x)
            Bernoulli factory with a given probability: B(p) \Rightarrow B(x)
(Mendo 2019).
                Mendo calls Bernoulli factories "non-randomized" if
their randomness
                is based entirely on the underlying coin.
            - f: Function that returns 1 if heads and 0 if tails.
            - x: Desired probability, in [0, 1].
       bernstein(self, f, alpha)
            Polynomial Bernoulli factory: B(p) =>
B(Bernstein(alpha))
                 (Goyal and Sigman 2012).
            - f: Function that returns 1 if heads and 0 if tails.
            - alpha: List of Bernstein coefficients for the
polynomial (when written
               in Bernstein form),
               whose degree is this list's length minus 1.
               For this to work, each coefficient must be in [0, 1].
       coin(self, c)
            Convenience method to generate a function that returns
            1 (heads) with the given probability c (which must be in
[0, 1])
            and 0 (tails) otherwise.
```

```
Complement (NOT): B(p) \Rightarrow B(1-p) (Flajolet et al. 2010)
            - f: Function that returns 1 if heads and 0 if tails.
       conditional(self, f1, f2, f3)
            Conditional: B(p), B(q), B(r) \Rightarrow B((1-r)*q+r*p)
(Flajolet et al. 2010)
            - f1, f2, f3: Functions that return 1 if heads and 0 if
tails.
       cos(self, f)
            Cosine Bernoulli factory: B(p) \Rightarrow B(cos(p)). Special
            case of Algorithm3 of reverse-time martingale paper.
       disjunction(self, f1, f2)
            Disjunction (OR): B(p), B(q) \Rightarrow B(p+q-p*q) (Flajolet et
al. 2010)
            - f1, f2: Functions that return 1 if heads and 0 if
tails.
       divoneplus(self, f)
            Divided by one plus p: B(p) \Rightarrow B(1/(1+p)), implemented
                     as a special case of the two-coin construction.
Prefer over even-parity
                     for having bounded expected running time for all
heads probabilities.
            - f: Function that returns 1 if heads and 0 if tails.
            Note that this function is slow as the probability of
heads approaches 1.
       eps div(self, f, eps)
            Bernoulli factory as follows: B(p) \rightarrow B(eps/p) (Lee et
al. 2014).
            - f: Function that returns 1 if heads and 0 if tails.
            - eps: Fraction in (0, 1), must be chosen so that eps <
p, where p is
              the probability of heads.
       evenparity(self, f)
            Even parity: B(p) \Rightarrow B(1/(1+p)) (Flajolet et al. 2010)
            - f: Function that returns 1 if heads and 0 if tails.
            Note that this function is slow as the probability of
```

complement(self, f)

```
| exp minus(self, f)
            Exp-minus Bernoulli factory: B(p) -> B(exp(-p))
(Łatuszyński et al. 2011).
            - f: Function that returns 1 if heads and 0 if tails.
     | exp minus ext(self, f, c=0)
            Extension to the exp-minus Bernoulli factory of
(Łatuszyński et al. 2011):
           B(p) \rightarrow B(exp(-p - c))
            To the best of my knowledge, I am not aware
                   of any article or paper that presents this
particular
                   Bernoulli factory (before my articles presenting
                   accurate beta and exponential generators).
            - f: Function that returns 1 if heads and 0 if tails.
            - c: Integer part of exp-minus. Default is 0.
     fill_geometric_bag(self, bag, precision=53)
     | geometric bag(self, u)
            Bernoulli factory for a uniformly-distributed random
number in (0, 1)
            (Flajolet et al. 2010).
            - u: List that holds the binary expansion, from left to
right, of the uniformly-
              distributed random number. Each element of the list
is 0, 1, or None (meaning
              the digit is not yet known). The list may be expanded
as necessary to put
              a new digit in the appropriate place in the binary
expansion.
     linear(self, f, cx, cy=1, eps=Fraction(1, 20))
            Linear Bernoulli factory: B(p) \Rightarrow B((cx/cy)*p) (Huber
2016).
            - f: Function that returns 1 if heads and 0 if tails.
            - cx, cy: numerator and denominator of c; the
probability of heads (p) is multiplied
              by c. c must be 0 or greater. If c > 1, c must be
chosen so that c*p \le 1 - eps.
```

heads approaches 1.

```
- eps: A Fraction in (0, 1). If c > 1, eps must be
chosen so that c*p \le 1 - eps.
     linear lowprob(self, f, cx, cy=1, m=Fraction(249, 500))
            Linear Bernoulli factory which is faster if the
probability of heads is known
               to be less than half: B(p) \Rightarrow B((cx/cy)*p) (Huber
2016).
            - f: Function that returns 1 if heads and 0 if tails.
            - cx, cy: numerator and denominator of c; the
probability of heads (p) is multiplied
              by c. c must be 0 or greater. If c > 1, c must be
chosen so that c*p \ll m \ll 1/2.
            - m: A Fraction in (0, 1/2). If c > 1, m must be chosen
so that c*p <= m < 1/2.
     linear_power(self, f, cx, cy=1, i=1, eps=Fraction(1, 20))
            Linear-and-power Bernoulli factory: B(p) =>
B((p*cx/cy)^i) (Huber 2019).
           - f: Function that returns 1 if heads and 0 if tails.
            - cx, cy: numerator and denominator of c; the
probability of heads (p) is multiplied
             by c. c must be 0 or greater. If c > 1, c must be
chosen so that c*p \le 1 - eps.
            - i: The exponent. Must be an integer and 0 or greater.
            - eps: A Fraction in (0, 1). If c > 1, eps must be
chosen so that c*p \le 1 - eps.
     | logistic(self, f, cx=1, cy=1)
            Logistic Bernoulli factory: B(p) \rightarrow B(cx*p/(cy+cx*p)) or
                B(p) \rightarrow B((cx/cy)*p/(1+(cx/cy)*p)) (Morina et al.
2019)
           - f: Function that returns 1 if heads and 0 if tails.
Note that this function can
              be slow as the probability of heads approaches 0.
            - cx, cy: numerator and denominator of c; the
probability of heads (p) is multiplied
     by c. c must be in (0, 1).
     | martingale(self, coin, coeff)
            General martingale algorithm for alternating power
            series.
```

```
that takes an index 'i' and calculates the coefficient
            for index 'i'. Indices start at 0.
     | mean(self, f1, f2)
            Mean: B(p), B(q) \Rightarrow B((p+q)/2) (Flajolet et al. 2010)
            - f1, f2: Functions that return 1 if heads and 0 if
tails.
       old linear(self, f, cx, cy=1, eps=Fraction(1, 20))
            Linear Bernoulli factory: B(p) \Rightarrow B((cx/cy)*p). Older
algorithm given in (Huber 2014).
            - f: Function that returns 1 if heads and 0 if tails.
            - cx, cy: numerator and denominator of c; the
probability of heads (p) is multiplied
              by c. c must be 0 or greater. If c > 1, c must be
chosen so that c*p < 1 - eps.
            - eps: A Fraction in (0, 1). If c > 1, eps must be
chosen so that c*p < 1 - eps.
     | one div pi(self)
            Generates 1 with probability 1/pi.
            Reference: Flajolet et al. 2010.
     power(self, f, ax, ay=1)
            Power Bernoulli factory: B(p) \Rightarrow B(p^(ax/ay)). (case of
(0, 1) provided by
             Mendo 2019).
            - f: Function that returns 1 if heads and 0 if tails.
            - ax, ay: numerator and denominator of the desired power
to raise the probability
             of heads to. This power must be 0 or greater.
     powerseries(self, f)
            Power series Bernoulli factory: B(p) \Rightarrow B(1 - c(0)*(1-p)
+ c(1)*(1-p)^2 +
              c(2)*(1-p)^3 + ...), where c(i) = c[i]/sum(c)
(Mendo 2019).
            - f: Function that returns 1 if heads and 0 if tails.
            - c: List of coefficients in the power series, all of
which must be
           non-negative integers.
```

'coin' is the coin to be flipped; 'coeff' is a function

```
probgenfunc(self, f, rng)
            Probability generating function Bernoulli factory: B(p)
\Rightarrow B(E[p^x]), where x is rng()
             (Dughmi et al. 2021). E[p^x] is the expected value of
p^x and is also known
             as the probability generating function.
            - f: Function that returns 1 if heads and 0 if tails.
            - rng: Function that returns a non-negative integer at
random.
              Example (Dughmi et al. 2021): if 'rng' is
Poisson(lamda) we have
              an "exponentiation" Bernoulli factory as follows:
              B(p) \Rightarrow B(exp(p*lamda-lamda))
     product(self, f1, f2)
            Product (conjunction; AND): B(p), B(q) \Rightarrow B(p*q)
(Flajolet et al. 2010)
            - f1, f2: Functions that return 1 if heads and 0 if
tails.
     | randbit(self)
            Generates a random bit that is 1 or 0 with equal
probability.
     | rndint(self, maxInclusive)
     | rndintexc(self, maxexc)
            Returns a random integer in [0, maxexc).
      simulate(self, coin, fbelow, fabove, fbound,
nextdegree=None)
            Simulates a general factory function defined by two
            sequences of polynomials that converge from above and
below.
            - coin(): Function that returns 1 or 0 with a fixed
probability.
            - fbelow(n, k): Calculates the kth Bernstein coefficient
(not the value),
              or a lower bound thereof, for the degree-n lower
polynomial (k starts at 0).
            - fabove(n, k): Calculates the kth Bernstein coefficient
```

```
or an upper bound thereof, for the degree-n upper
polynomial.
            - fbound(n): Returns a tuple or list specifying a lower
and upper bound
               among the values of fbelow and fabove, respectively,
for the given n.
             - nextdegree(n): Returns a lambda returning the next
degree after the
               given degree n for which a polynomial is available;
the lambda
               must return an integer greater than n.
               Optional. If not given, the first degree is 1 and
the next degree is n*2
               (so that for each power of 2 as well as 1, a
polynomial of that degree
               must be specified).
       sin(self, f)
            Sine Bernoulli factory: B(p) \Rightarrow B(\sin(p)). Special
            case of Algorithm3 of reverse-time martingale paper.
       square(self, f1, f2)
            Square: B(p) \Rightarrow B(1-p). (Flajolet et al. 2010)
            - f1, f2: Functions that return 1 if heads and 0 if
tails.
       twocoin(self, f1, f2, c1=1, c2=1, beta=1)
            Two-coin Bernoulli factory: B(p), B(q) =>
                      B(c1*p*beta / (beta * (c1*p+c2*q) - (beta -
1)*(c1+c2)))
                (Gonçalves et al. 2017, Vats et al. 2020; in Vats et
al..
                C1,p1 corresponds to cy and C2,p2 corresponds to
cx).
                Logistic Bernoulli factory is a special case with
q=1, c2=1, beta=1.
            - f1, f2: Functions that return 1 if heads and 0 if
tails.
            - c1, c2: Factors to multiply the probabilities of heads
for f1 and f2, respectively.
            - beta: Early rejection parameter ("portkey" two-coin
```

(not the value),

```
When beta = 1, the formula simplifies to
B(c1*p/(c1*p+c2*q)).
       twofacpower(self, fbase, fexponent)
            Bernoulli factory B(p, q) \Rightarrow B(p^q).
            Based on algorithm from (Mendo 2019),
            but changed to accept a Bernoulli factory
            rather than a fixed value for the exponent.
            To the best of my knowledge, I am not aware
            of any article or paper that presents this particular
            Bernoulli factory (before my articles presenting
            accurate beta and exponential generators).
            - fbase, fexponent: Functions that return 1 if heads and
0 if tails.
              The first is the base, the second is the exponent.
       zero or one(self, px, py)
            Returns 1 at probability px/py, 0 otherwise.
       zero or one arctan n div n(self, x, y=1)
            Generates 1 with probability arctan(x/y)*y/x; 0
otherwise.
               x/y must be in [0, 1]. Uses a uniformly-fast special
case of
            the two-coin Bernoulli factory, rather than the even-
parity construction in
            Flajolet's paper, which does not have bounded expected
running time for all heads probabilities.
            Reference: Flajolet et al. 2010.
     zero or one exp minus(self, x, y)
            Generates 1 with probability exp(-x/y); 0 otherwise.
            Reference: Canonne et al. 2020.
       zero or one log1p(self, x, y=1)
            Generates 1 with probability log(1+x/y); 0 otherwise.
            Reference: Flajolet et al. 2010. Uses a uniformly-fast
special case of
            the two-coin Bernoulli factory, rather than the even-
parity construction in
            Flajolet's paper, which does not have bounded expected
```

factory).

```
| zero or one pi div 4(self)
           Generates 1 with probability pi/4.
           Reference: Flajolet et al. 2010.
      zero or one power(self, px, py, n)
           Generates 1 with probability (px/py)^n (where n can be
           positive, negative, or zero); 0 otherwise.
    zero or one power ratio(self, px, py, nx, ny)
           Generates 1 with probability (px/py)^{(nx/ny)} (where
nx/ny can be
           positive, negative, or zero); 0 otherwise.
    | -----
    | Data descriptors defined here:
    | __dict__
           dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
   class DiceEnterprise(builtins.object)
    | Implements the Dice Enterprise algorithm for
    | turning loaded dice with unknown probability of heads into
loaded dice
    | with a different probability of heads. Specifically, it
supports specifying
    | the probability that the output die will land on a given
    | number, as a polynomial function of the input die's
probability of heads.
    | The case of coins to coins is also called
    the Bernoulli factory problem; this class allows the output
    | coin's probability of heads to be specified as a polynomial
function of the
    | input coin's probability of heads.
    Reference: Morina, G., Łatuszyński, K., et al., "From the
    | Bernoulli Factory to a Dice Enterprise via Perfect
```

running time for all heads probabilities.

```
Sampling of Markov Chains", arXiv:1912.09229v1 [math.PR],
2019.
       Example:
       >>> from bernoulli import DiceEnterprise
       >>> import math
      >>> import random
     | >>> ent=DiceEnterprise()
      >>> # Example 3 from the paper
      >>> ent.append_poly(1,[[math.sqrt(2),3]])
     >>> ent.append_poly(0,[[-5,3],[11,2],[-9,1],[3,0]])
      >>> coin=lambda: 1 if random.random() < 0.60 else 0
       >>> print([ent.next(coin) for i in range(100)])
       Methods defined here:
        init (self)
            Initialize self. See help(type(self)) for accurate
signature.
       append_poly(self, result, poly)
            Appends a probability that the output die will land on
            a given number, in the form of a polynomial.
            result - A number indicating the result (die roll or
coin
             flip) that will be returned by the output coin or
_output_
             die with the probability represented by this
polynomial.
             Must be an integer 0 or greater. In the case of dice-
to-coins
             or coins-to-coins, must be either 0 or 1, where 1
means
             heads and 0 means tails.
            poly - Polynomial expressed as a list of terms as
follows:
             Each term is a list of two or more items that each
express one of
             the polynomial's terms; the first item is the
coefficient, and
```

```
the remaining items are the powers of the input die's
              probabilities. The number of remaining items in each
term
              is the number of faces the input die has.
Specifically, the
              term has the following form:
              In the case of coins-to-dice or coins-to-coins (so the
probabilities are 1-p and p,
             where the [unknown] probability that the input coin
returns 0
              is 1 - p, or returns 1 is p):
                       term[0] * p**term[1] * (1-p)**term[2].
              In the case of dice-to-dice or dice-to-coins (so the
probabilities are p1, p2, etc.,
              where the [unknown] probability that the input die
returns
              0 is p1, returns 1 is p2, etc.):
                       term[0] * p1**term[1] * p2**term[2] * ... *
pn**term[n].
              For example, [3, 4, 5] becomes:
                       3 * p**4 * (1-p)**5
              As a special case, the term can contain two items and
a zero is
              squeezed between the first and second item.
              For example, [3, 4] is the same as [3, 0, 4], which in
turn becomes:
                       3 * p**4 * (1-p)**0 = 3 * p **4
              For best results, the coefficient should be a rational
number
              (such as int or Python's Fraction).
              Each term in the polynomial must have the same number
of items (except
              for the special case given above). For example, the
following is not a valid
             way to express this parameter:
                       [[1, 1, 0], [1, 3, 4, 5], [1, 1, 2], [2, 3,
4]]
             Here, the second term has four items, not three like
```

```
the rest.
           Returns this object.
       augment(self, count=1)
           Augments the degree of the function represented
           by this object, which can improve performance in some
cases
           (for details, see the paper).
           - count: Number of times to augment the ladder.
           Returns this object.
      next(self, coin)
           Returns the next result of the flip from a coin or die
           that is transformed from the given input coin or die by
the function
           represented by this Dice Enterprise object.
           coin - In the case of coins-to-dice or coins-to-coins
(see the "append poly" method),
              this specifies the input coin , which must be a
function that
              returns either 1 (heads) or 0 (tails). In the case
    of dice-to-dice or dice-to-coins,
             this specifies an _input die_ with _m_ faces, which
must be a
            function that returns an integer in the interval [0,
    m), which
             specifies which face the input die lands on.
    | ------
    | Data descriptors defined here:
       dict
           dictionary for instance variables (if defined)
        weakref
           list of weak references to the object (if defined)
FILE
/home/peter/Documents/SharpDevelopProjects/peteroupc.github.io/bernou
```

```
Help on module interval:
NAME
   interval
DESCRIPTION
   # Implements interval numbers and interval arithmetic, backed
   # by Fractions.
   # Written by Peter O. Any copyright to this file is released to
the Public Domain.
   # In case this is not possible, this file is also licensed
under Creative Commons Zero
   # (https://creativecommons.org/publicdomain/zero/1.0/).
CLASSES
    builtins.object
        FInterval
    class FInterval(builtins.object)
     | FInterval(v, sup=None, prec=None)
     An interval of two Fractions. x.sup holds the upper bound,
and x.inf holds
     | the lower bound.
     | Methods defined here:
     | __abs__(self)
     __add__(self, v)
     | __max__(a, b)
     | __min__(a, b)
     __mul__(self, v)
       __neg__(self)
```

```
| __radd__(self, v)
  __repr__(self)
     Return repr(self).
| __rmul__(self, v)
| __rsub__(self, v)
 __rtruediv__(self, v)
| __sub__(self, v)
  truediv (self, v)
| abs(self)
| atan(self, precision)
  atan2(self, x, precision)
  ceil(self)
  clamp(self, a, b)
  clampleft(self, a)
  containedIn(self, y)
  cos(self, precision)
  exp(self, precision)
  floor(self)
  greaterEqualScalar(self, a)
  greaterThanScalar(self, a)
  intersect(self, y)
  isAccurateTo(self, v)
```

```
lessEqualScalar(self, a)
      lessThanScalar(self, a)
      log(self, precision)
      magnitude(self)
       mignitude(self)
       negate(self)
      pi(precision)
      pow(self, v, precision)
     rem(self, v)
      sin(self, precision)
      sqrt(self, n)
    | tan(self, precision)
    | union(v)
    | width(self)
    | Static methods defined here:
    | __new__(cl, v, sup=None, prec=None)
          Create and return a new object. See help(type) for
accurate signature.
               -----
    | Data descriptors defined here:
    | __dict__
```

```
dictionary for instance variables (if defined)
        weakref__
            list of weak references to the object (if defined)
DATA
    ArcTanHTable = [0, 294906490, 137123709, 67461703, 33598225,
16782680,...
    CRUDELOG = [0, -726816, -681390, -654818, -635964, -621340,
-609392, -...
    CRUDELOG ARCTANBITDIFF = 13
    CRUDELOG_ARCTANFRAC = 29
    CRUDELOG BITS = 16
    CRUDELOG LOG2BITS = 45426
    CRUDELOG LOGMIN = 9830
    LNPOLY2 = [(-28986367995118693...8591117027361355259,
10000000000000000...
    LNPOLY3 = [(-13476514299119388...8263005361644498323,
50000000000000000...
    REALHALFPI = RealPi(1/2)
    REALPI = RealPi(1)
    REAL 858 1000 = RealFraction(429/500)
FILE
/home/peter/Documents/SharpDevelopProjects/peteroupc.github.io/interv
Help on module moore:
NAME
    moore
DESCRIPTION
    # Implements the Moore Rejection Sampler.
    # Written by Peter O. Any copyright to this file is released to
the Public Domain.
    # In case this is not possible, this file is also licensed
under Creative Commons Zero
    # (https://creativecommons.org/publicdomain/zero/1.0/).
```

```
CLASSES
   builtins.object
       MooreSampler
   class MooreSampler(builtins.object)
       MooreSampler(pdf, mn, mx, numLabels=1, bitAccuracy=53)
     | Moore rejection sampler, for generating independent samples
     | from a distribution in a way that minimizes error,
     | if the distribution has a PDF (probability density function)
     and the PDF uses "well-defined" arithmetic expressions.
     | It can sample from one-dimensional or multidimensional
     | distributions. It can also sample from so-called
"transdimensional
     | distributions" if the distribution is the union of several
component
     | distributions that may have different dimensions and are
associated
     | with one of several _labels_.
     | Parameters:
     | - pdf: A function that specifies the PDF. It takes a single
parameter that
            differs as follows, depending on the case:
            - One-dimensional case: A single FInterval. (An
FInterval is a mathematical
             object that specifies upper and lower bounds of a
number.)
           - Multidimensional case: A list of FIntervals, one for
each dimension.
           - Transdimensional case (numLabels > 1): A list of two
items: the FInterval
               or FIntervals, followed by a label number (an integer
     1
in [0, numLabels)).
           This function returns an FInterval. For best results,
           the function should use interval arithmetic throughout.
The area under
           the PDF need not equal 1 (this sampler works even if the
PDF is only known
           up to a normalizing constant).
```

- mn, mx: Specifies the sampling domain of the PDF. There are three cases: - One-dimensional case: Both mn and mx are numbers giving the domain, 1 which in this case is [mn, mx]. - Multidimensional case: Both mn and mx are lists giving the minimum 1 and maximum bounds for each dimension in the sampling domain. In this case, both lists must have the same size. - Transdimensional case: Currently, this class assumes the component distributions share the same sampling domain, which is given depending on the preceding two cases. For this sampler to work, the PDF must be "locally Lipschitz" in the sampling domain, meaning that the PDF is continuous and there is a constant L such that PDF(x) and PDF(y) are in the sampling domain and no more than _L_ times _ε_ apart whenever _x_ and _y_ are no more than _ε_ apart. - numlabels: The number of labels associated with the distribution, if it's a transdimensional distribution. Optional; the default is 1. - bitAccuracy: Bit accuracy of the sampler; the sampler will sample from a distribution (truncated to the sampling domain) that is 1 close to the ideal distribution by 2^-bitAccuracy. The default is 53. | Reference: Sainudiin, Raazesh, and Thomas L. York. "An Auto-Validating, Trans-Dimensional, | Universal Rejection Sampler for Locally Lipschitz Arithmetical Expressions." | Reliable Computing 18 (2013): 15-54. The following reference describes an optimization, not yet implemented here: | Sainudiin, R., 2014. An Auto-validating Rejection Sampler for Differentiable | Arithmetical Expressions: Posterior Sampling of Phylogenetic

```
Quartets. In
     | Constraint Programming and Decision Making (pp. 143-152).
Springer, Cham.
     | Methods defined here:
     init (self, pdf, mn, mx, numLabels=1, bitAccuracy=53)
           Initialize self. See help(type(self)) for accurate
signature.
     | acceptRate(self)
     | sample(self)
            Samples a number or vector (depending on the number of
dimensions)
           from the distribution and returns that sample.
           If the sampler is transdimensional (the number of labels
is greater than 1),
           instead returns a list containing the sample and a
random label in the
           interval [0, numLabels), in that order.
     | Data descriptors defined here:
     | __dict__
            dictionary for instance variables (if defined)
       weakref
           list of weak references to the object (if defined)
FILE
/home/peter/Documents/SharpDevelopProjects/peteroupc.github.io/moore
Help on module betadist:
NAME
   betadist
```

```
DESCRIPTION
    # Written by Peter O. Any copyright to this file is released to
the Public Domain.
    # In case this is not possible, this file is also licensed
under Creative Commons Zero
    # (https://creativecommons.org/publicdomain/zero/1.0/).
CLASSES
    builtins.object
        BernsteinPoly
        FPInterval
        PhaseType
        PiecewiseBernstein
        Real
            RandPSRN
            RandUniform
            RandUniformIntFrac
            RandUniformNegIntFrac
            RealAdd
            RealArcTan
            RealArcTan2
            RealCos
            RealDivide
            RealErf
            RealExp
            RealFraction
            RealLn
            RealLogGammaInt
            RealMultiply
            RealNegate
            RealPi
            RealPow
            RealSin
            RealSqrt
            RealSubtract
            RealTan
        ShapeSampler
        ShapeSampler2
        SinFunction
    class BernsteinPoly(builtins.object)
```

```
BernsteinPoly(coeffs)
     | Methods defined here:
     | __init__(self, coeffs)
           Initialize self. See help(type(self)) for accurate
signature.
     | deriv(self, d=1)
     | diff(self, pt, d=1)
     | fromFunc(func, n)
     | lipschitz(self)
     | value(self, pt)
     | Data descriptors defined here:
     | __dict__
         dictionary for instance variables (if defined)
     | __weakref__
         list of weak references to the object (if defined)
   class FPInterval(builtins.object)
     | FPInterval(n, d, prec)
     | Methods defined here:
     | __init__(self, n, d, prec)
          Initialize self. See help(type(self)) for accurate
signature.
     | __repr__(self)
         Return repr(self).
     | addintv(self, intv)
```

```
addnumden(self, n, d)
    | copy(self)
    | mulnumden(self, n, d)
    | setprec(self, prec)
    | subintv(self, intv)
    | Data descriptors defined here:
    | __dict__
         dictionary for instance variables (if defined)
    | __weakref__
          list of weak references to the object (if defined)
   class PhaseType(builtins.object)
    | PhaseType(alpha, s)
    | Methods defined here:
    init (self, alpha, s)
         Initialize self. See help(type(self)) for accurate
signature.
    | sample(self)
    _____
    | Data descriptors defined here:
    | __dict__
          dictionary for instance variables (if defined)
    | weakref
          list of weak references to the object (if defined)
   class PiecewiseBernstein(builtins.object)
```

```
Methods defined here:
       __init__(self)
            Initialize self. See help(type(self)) for accurate
signature.
       deriv(self, d=1)
       diff(self, x, d=1)
      fromcoeffs(coeffs)
            Creates a PiecewiseBernstein given a
            polynomial's Bernstein coefficients.
       get_coeffs(self)
       lipschitz(self)
     | piece(self, coeffs, mn, mx)
     | value(self, x)
       Data descriptors defined here:
     | __dict__
            dictionary for instance variables (if defined)
        weakref
            list of weak references to the object (if defined)
    class RandPSRN(Real)
        RandPSRN(a)
       Method resolution order:
            RandPSRN
            Real
            builtins.object
       Methods defined here:
```

```
| __init__(self, a)
         Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
          Return repr(self).
     | ev(self, n)
     | isNegative()
     | Methods inherited from Real:
     | __abs__(a)
     __add__(a, b)
      __ge__(a, b)
          Return self>=value.
       __gt__(a, b)
         Return self>value.
      __le__(a, b)
         Return self<=value.
      __lt__(a, b)
         Return self<value.
     | __mul__(a, b)
     | __neg__(a)
     | __pow__(a, b)
     | __radd__(a, b)
     | __rmul__(a, b)
      __rpow__(b, a)
```

```
| __rsub__(a, b)
       __rtruediv__(a, b)
     | __sub__(a, b)
     | __truediv__(a, b)
     | disp(a)
     | evstable(a, prec)
     | Data descriptors inherited from Real:
     | __dict__
           dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
    class RandUniform(Real)
      Method resolution order:
           RandUniform
           Real
           builtins.object
      Methods defined here:
       __init__(self)
           Initialize self. See help(type(self)) for accurate
signature.
        __repr__(self)
          Return repr(self).
     | ev(self, n)
       isNegative(self)
```

```
| Methods inherited from Real:
| __abs__(a)
| __add__(a, b)
  __ge__(a, b)
    Return self>=value.
  __gt__(a, b)
     Return self>value.
  __le__(a, b)
     Return self<=value.
 __lt__(a, b)
     Return self<value.
| __mul__(a, b)
| __neg__(a)
| __pow__(a, b)
  __radd__(a, b)
 __rmul__(a, b)
| __rpow__(b, a)
  __rsub__(a, b)
| __rtruediv__(a, b)
 __sub__(a, b)
  __truediv__(a, b)
  disp(a)
```

```
| evstable(a, prec)
    | Data descriptors inherited from Real:
    | __dict__
        dictionary for instance variables (if defined)
    __weakref__
        list of weak references to the object (if defined)
   class RandUniformIntFrac(Real)
    | RandUniformIntFrac(i, f)
    | Method resolution order:
          RandUniformIntFrac
          Real
          builtins.object
    | Methods defined here:
    | __init__(self, i, f)
         Initialize self. See help(type(self)) for accurate
signature.
    | __repr__(self)
      Return repr(self).
    | ev(self, n)
    _____
    | Methods inherited from Real:
    | __abs__(a)
    | __add__(a, b)
    | __ge__(a, b)
          Return self>=value.
```

```
Return self>value.
  __le__(a, b)
      Return self<=value.
 __lt__(a, b)
    Return self<value.
| __mul__(a, b)
| __neg__(a)
__pow__(a, b)
__radd__(a, b)
| __rmul__(a, b)
 __rpow__(b, a)
| __rsub__(a, b)
| __rtruediv__(a, b)
| __sub__(a, b)
  __truediv__(a, b)
  disp(a)
| evstable(a, prec)
| isNegative(self)
 Data descriptors inherited from Real:
  __dict__
      dictionary for instance variables (if defined)
```

__gt__(a, b)

```
| __weakref__
           list of weak references to the object (if defined)
    class RandUniformNegIntFrac(Real)
        RandUniformNegIntFrac(i, f)
       Method resolution order:
           RandUniformNegIntFrac
           Real
           builtins.object
     | Methods defined here:
     | init (self, i, f)
           Initialize self. See help(type(self)) for accurate
signature.
        repr (self)
           Return repr(self).
     | ev(self, n)
     | Methods inherited from Real:
     | __abs__(a)
     | __add__(a, b)
       __ge__(a, b)
           Return self>=value.
       __gt__(a, b)
           Return self>value.
       __le__(a, b)
           Return self<=value.
       __lt__(a, b)
           Return self<value.
```

```
__mul__(a, b)
   __neg__(a)
 | __pow__(a, b)
   __radd__(a, b)
   __rmul__(a, b)
   __rpow__(b, a)
   __rsub__(a, b)
 | __rtruediv__(a, b)
   __sub__(a, b)
   __truediv__(a, b)
 | disp(a)
   evstable(a, prec)
 | isNegative(self)
 | Data descriptors inherited from Real:
 | __dict__
       dictionary for instance variables (if defined)
   __weakref__
       list of weak references to the object (if defined)
class Real(builtins.object)
 | Methods defined here:
 | __abs__(a)
   __add__(a, b)
```

```
__ge__(a, b)
     Return self>=value.
  __gt__(a, b)
      Return self>value.
  __le__(a, b)
     Return self<=value.
  __lt__(a, b)
      Return self<value.
  __mul__(a, b)
| __neg__(a)
| __pow__(a, b)
  __radd__(a, b)
  __repr__(a)
     Return repr(self).
  __rmul__(a, b)
  __rpow__(b, a)
 __rsub__(a, b)
| __rtruediv__(a, b)
  __sub__(a, b)
 __truediv__(a, b)
| disp(a)
 ev(self, n)
  evstable(a, prec)
```

```
| isNegative(self)
    | Data descriptors defined here:
    | __dict__
    dictionary for instance variables (if defined)
    __weakref__
        list of weak references to the object (if defined)
   class RealAdd(Real)
    | RealAdd(a, b)
    | Method resolution order:
          RealAdd
          Real
          builtins.object
    | Methods defined here:
    | __init__(self, a, b)
        Initialize self. See help(type(self)) for accurate
signature.
    | __repr__(self)
      Return repr(self).
    | ev(self, n)
    _____
    | Methods inherited from Real:
    | __abs__(a)
    | __add__(a, b)
    | __ge__(a, b)
          Return self>=value.
```

```
__gt__(a, b)
    Return self>value.
  __le__(a, b)
      Return self<=value.
 __lt__(a, b)
    Return self<value.
| __mul__(a, b)
| __neg__(a)
__pow__(a, b)
__radd__(a, b)
| __rmul__(a, b)
 __rpow__(b, a)
| __rsub__(a, b)
| __rtruediv__(a, b)
| __sub__(a, b)
  __truediv__(a, b)
  disp(a)
| evstable(a, prec)
| isNegative(self)
 Data descriptors inherited from Real:
  __dict__
      dictionary for instance variables (if defined)
```

```
| __weakref__
          list of weak references to the object (if defined)
   class RealArcTan(Real)
     | RealArcTan(a)
     | Method resolution order:
           RealArcTan
           Real
           builtins.object
     | Methods defined here:
     | init (self, a)
           Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
          Return repr(self).
     | ev(self, n)
     | Methods inherited from Real:
     | __abs__(a)
     | __add__(a, b)
       __ge__(a, b)
          Return self>=value.
     | __gt__(a, b)
           Return self>value.
       __le__(a, b)
           Return self<=value.
       __lt__(a, b)
           Return self<value.
```

```
__mul__(a, b)
   __neg__(a)
 | __pow__(a, b)
   __radd__(a, b)
   __rmul__(a, b)
   __rpow__(b, a)
   __rsub__(a, b)
 | __rtruediv__(a, b)
   __sub__(a, b)
 | __truediv__(a, b)
 | disp(a)
  evstable(a, prec)
 | isNegative(self)
 | Data descriptors inherited from Real:
 | __dict__
       dictionary for instance variables (if defined)
   __weakref__
       list of weak references to the object (if defined)
class RealArcTan2(Real)
 | RealArcTan2(y, x)
 | Method resolution order:
       RealArcTan2
       Real
```

```
builtins.object
     | Methods defined here:
      __init__(self, y, x)
           Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
         Return repr(self).
     | ev(self, n)
     | Methods inherited from Real:
    | __abs__(a)
    | __add__(a, b)
      __ge__(a, b)
         Return self>=value.
       __gt__(a, b)
         Return self>value.
       __le__(a, b)
         Return self<=value.
      __lt__(a, b)
         Return self<value.
     | __mul__(a, b)
     | __neg__(a)
     | __pow__(a, b)
      __radd__(a, b)
      __rmul__(a, b)
```

```
__rpow__(b, a)
       __rsub__(a, b)
      __rtruediv__(a, b)
     __sub__(a, b)
       __truediv__(a, b)
       disp(a)
     | evstable(a, prec)
     | isNegative(self)
     | Data descriptors inherited from Real:
     | __dict__
          dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
   class RealCos(Real)
     | RealCos(a)
     | Method resolution order:
           RealCos
           Real
           builtins.object
     | Methods defined here:
       __init__(self, a)
           Initialize self. See help(type(self)) for accurate
signature.
     | __repr__(self)
```

```
Return repr(self).
 ev(self, n)
| Methods inherited from Real:
| __abs__(a)
| __add__(a, b)
  __ge__(a, b)
     Return self>=value.
  __gt__(a, b)
     Return self>value.
  __le__(a, b)
    Return self<=value.
  __lt__(a, b)
    Return self<value.
 __mul__(a, b)
 ___neg___(a)
 __pow__(a, b)
 __radd__(a, b)
  __rmul__(a, b)
 __rpow__(b, a)
  __rsub__(a, b)
| __rtruediv__(a, b)
  __sub__(a, b)
```

```
__truediv__(a, b)
    | disp(a)
    | evstable(a, prec)
    | isNegative(self)
    | Data descriptors inherited from Real:
    | __dict__
         dictionary for instance variables (if defined)
      weakref
        list of weak references to the object (if defined)
   class RealDivide(Real)
    | RealDivide(a, b)
    | Method resolution order:
          RealDivide
          Real
          builtins.object
    | Methods defined here:
      __init__(self, a, b)
         Initialize self. See help(type(self)) for accurate
signature.
     __repr__(self)
        Return repr(self).
    | ev(self, n)
    | ------
    | Methods inherited from Real:
    | __abs__(a)
```

```
__add__(a, b)
  __ge__(a, b)
      Return self>=value.
  __gt__(a, b)
      Return self>value.
  __le__(a, b)
    Return self<=value.
  __lt__(a, b)
    Return self<value.
| __mul__(a, b)
| __neg__(a)
 __pow__(a, b)
 __radd__(a, b)
 __rmul__(a, b)
| __rpow__(b, a)
| __rsub__(a, b)
  __rtruediv__(a, b)
| __sub__(a, b)
  __truediv__(a, b)
  disp(a)
 evstable(a, prec)
  isNegative(self)
```

```
| Data descriptors inherited from Real:
    | __dict__
          dictionary for instance variables (if defined)
    | __weakref__
          list of weak references to the object (if defined)
   class RealErf(Real)
    | RealErf(a)
    | Method resolution order:
          RealErf
          Real
          builtins.object
    | Methods defined here:
    | __init__(self, a)
         Initialize self. See help(type(self)) for accurate
signature.
    | __repr__(self)
      Return repr(self).
    | ev(self, n)
    | -----
    | Methods inherited from Real:
    | __abs__(a)
    | __add__(a, b)
    | __ge__(a, b)
          Return self>=value.
    | __gt__(a, b)
          Return self>value.
```

```
__le__(a, b)
      Return self<=value.
 __lt__(a, b)
    Return self<value.
| __mul__(a, b)
| __neg__(a)
| __pow__(a, b)
| __radd__(a, b)
| __rmul__(a, b)
| __rpow__(b, a)
__rsub__(a, b)
| __rtruediv__(a, b)
| __sub__(a, b)
 __truediv__(a, b)
 disp(a)
| evstable(a, prec)
| isNegative(self)
  Data descriptors inherited from Real:
 __dict__
      dictionary for instance variables (if defined)
  __weakref__
      list of weak references to the object (if defined)
```

```
class RealExp(Real)
       RealExp(a)
      Method resolution order:
          RealExp
          Real
          builtins.object
    | Methods defined here:
    | __init__(self, a)
          Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
         Return repr(self).
    | ev(self, n)
    | isNegative()
            -----
    | Methods inherited from Real:
    | __abs__(a)
    | __add__(a, b)
      ge (a, b)
          Return self>=value.
      __gt__(a, b)
        Return self>value.
      __le__(a, b)
        Return self<=value.
    | __lt__(a, b)
         Return self<value.
      __mul__(a, b)
```

```
__neg__(a)
   __pow__(a, b)
   __radd__(a, b)
   __rmul__(a, b)
   __rpow__(b, a)
   __rsub__(a, b)
   rtruediv (a, b)
 | __sub__(a, b)
 | __truediv__(a, b)
   disp(a)
 | evstable(a, prec)
 | Data descriptors inherited from Real:
 | __dict__
       dictionary for instance variables (if defined)
   __weakref__
       list of weak references to the object (if defined)
class RealFraction(Real)
   RealFraction(a, b=None)
 | Method resolution order:
       RealFraction
       Real
       builtins.object
   Methods defined here:
```

```
| __abs__(self)
       __add__(self, b)
       __init__(self, a, b=None)
           Initialize self. See help(type(self)) for accurate
signature.
       __mul__(self, b)
     | __neg__(self)
     | radd (self, b)
       __repr__(self)
          Return repr(self).
     | __rmul__(self, b)
     | __rsub__(self, b)
     | __rtruediv__(self, b)
     | __sub__(self, b)
       __truediv__(self, b)
     | ev(self, n)
     | isDefinitelyZero(self)
       isNegative(self)
       toFraction(self)
     | Methods inherited from Real:
       __ge__(a, b)
          Return self>=value.
```

```
__gt__(a, b)
          Return self>value.
       __le__(a, b)
          Return self<=value.
      __lt__(a, b)
          Return self<value.
     | __pow__(a, b)
       __rpow__(b, a)
     | disp(a)
     | evstable(a, prec)
     | Data descriptors inherited from Real:
     | __dict__
         dictionary for instance variables (if defined)
       __weakref__
          list of weak references to the object (if defined)
    class RealLn(Real)
     | RealLn(a)
     | Method resolution order:
           RealLn
           Real
           builtins.object
     | Methods defined here:
       __init__(self, a)
           Initialize self. See help(type(self)) for accurate
signature.
```

```
__repr__(self)
      Return repr(self).
  ev(self, n)
| isDefinitelyZero(self)
| Methods inherited from Real:
| __abs__(a)
| __add__(a, b)
  __ge__(a, b)
     Return self>=value.
  __gt__(a, b)
     Return self>value.
  __le__(a, b)
     Return self<=value.
  __lt__(a, b)
     Return self<value.
| __mul__(a, b)
  ___neg___(a)
| __pow__(a, b)
  __radd__(a, b)
  __rmul__(a, b)
| __rpow__(b, a)
| __rsub__(a, b)
  __rtruediv__(a, b)
```

```
__sub__(a, b)
       __truediv__(a, b)
       disp(a)
     | evstable(a, prec)
       isNegative(self)
     | Data descriptors inherited from Real:
       dict
           dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
   class RealLogGammaInt(Real)
       RealLogGammaInt(a)
     | Method resolution order:
           {\tt RealLogGammaInt}
           Real
           builtins.object
       Methods defined here:
       __init__(self, a)
           Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
           Return repr(self).
     | ev(self, n)
```

```
Methods inherited from Real:
| __abs__(a)
__add__(a, b)
  __ge__(a, b)
      Return self>=value.
  __gt__(a, b)
     Return self>value.
  __le__(a, b)
    Return self<=value.
  __lt__(a, b)
     Return self<value.
 __mul__(a, b)
| __neg__(a)
| __pow__(a, b)
  __radd__(a, b)
  __rmul__(a, b)
 __rpow__(b, a)
| __rsub__(a, b)
  __rtruediv__(a, b)
| __sub__(a, b)
  __truediv__(a, b)
  disp(a)
  evstable(a, prec)
```

```
| isNegative(self)
    | Data descriptors inherited from Real:
     | __dict__
         dictionary for instance variables (if defined)
     __weakref__
         list of weak references to the object (if defined)
   class RealMultiply(Real)
     | RealMultiply(a, b)
     | Method resolution order:
           RealMultiply
           Real
           builtins.object
     | Methods defined here:
     | __init__(self, a, b)
          Initialize self. See help(type(self)) for accurate
signature.
     | __repr__(self)
        Return repr(self).
     | ev(self, n)
     | isDefinitelyZero(self)
     | mul(a, b)
     | Methods inherited from Real:
     | __abs__(a)
     __add__(a, b)
```

```
__ge__(a, b)
      Return self>=value.
  __gt__(a, b)
      Return self>value.
  __le__(a, b)
      Return self<=value.
  __lt__(a, b)
      Return self<value.
 __mul__(a, b)
| __neg__(a)
| __pow__(a, b)
  __radd__(a, b)
 __rmul__(a, b)
 __rpow__(b, a)
 __rsub__(a, b)
 __rtruediv__(a, b)
  __sub__(a, b)
  __truediv__(a, b)
  disp(a)
  evstable(a, prec)
  isNegative(self)
| Data descriptors inherited from Real:
```

```
dict
          dictionary for instance variables (if defined)
       __weakref__
          list of weak references to the object (if defined)
   class RealNegate(Real)
       RealNegate(a)
    | Method resolution order:
          RealNegate
          Real
          builtins.object
    | Methods defined here:
       __init__(self, a)
          Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
         Return repr(self).
    | ev(self, n)
      ______
    | Methods inherited from Real:
    | __abs__(a)
    | __add__(a, b)
    | __ge__(a, b)
         Return self>=value.
      __gt__(a, b)
         Return self>value.
      __le__(a, b)
          Return self<=value.
```

```
__lt__(a, b)
      Return self<value.
 | __mul__(a, b)
   __neg__(a)
   __pow__(a, b)
   __radd__(a, b)
   __rmul__(a, b)
 | __rpow__(b, a)
   __rsub__(a, b)
 | __rtruediv__(a, b)
 | __sub__(a, b)
   __truediv__(a, b)
   disp(a)
   evstable(a, prec)
 | isNegative(self)
 | Data descriptors inherited from Real:
 | __dict__
       dictionary for instance variables (if defined)
   __weakref__
       list of weak references to the object (if defined)
class RealPi(Real)
 | RealPi(fraction=1, consistent=False)
```

```
Method resolution order:
           RealPi
           Real
           builtins.object
      Methods defined here:
       __init__(self, fraction=1, consistent=False)
           Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
          Return repr(self).
     | ev(self, n)
     | Methods inherited from Real:
     | __abs__(a)
     | __add__(a, b)
       __ge__(a, b)
          Return self>=value.
       __gt__(a, b)
          Return self>value.
       __le__(a, b)
         Return self<=value.
       __lt__(a, b)
         Return self<value.
     | __mul__(a, b)
     | __neg__(a)
       __pow__(a, b)
```

```
| __radd__(a, b)
   __rmul__(a, b)
  __rpow__(b, a)
  __rsub__(a, b)
   __rtruediv__(a, b)
 | __sub__(a, b)
   truediv (a, b)
 | disp(a)
 | evstable(a, prec)
 | isNegative(self)
 | Data descriptors inherited from Real:
      dictionary for instance variables (if defined)
   __weakref__
      list of weak references to the object (if defined)
class RealPow(Real)
   RealPow(a, b)
 | Method resolution order:
       RealPow
       Real
       builtins.object
 | Methods defined here:
   __init__(self, a, b)
```

```
Initialize self. See help(type(self)) for accurate
signature.
     | __repr__(self)
         Return repr(self).
     | ev(self, n)
     | Methods inherited from Real:
     | __abs__(a)
     | __add__(a, b)
     | __ge__(a, b)
           Return self>=value.
      __gt__(a, b)
         Return self>value.
      __le__(a, b)
         Return self<=value.
     | __lt__(a, b)
         Return self<value.
     | __mul__(a, b)
     | __neg__(a)
      __pow__(a, b)
     | __radd__(a, b)
     | __rmul__(a, b)
     | __rpow__(b, a)
     | __rsub__(a, b)
```

```
| __rtruediv__(a, b)
     | __sub__(a, b)
     | __truediv__(a, b)
     | disp(a)
     | evstable(a, prec)
     | isNegative(self)
     | Data descriptors inherited from Real:
     | __dict__
           dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
    class RealSin(Real)
     | RealSin(a)
     | Method resolution order:
           RealSin
           Real
           builtins.object
     | Methods defined here:
       __init__(self, a)
           Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
           Return repr(self).
     | ev(self, n)
     | isDefinitelyZero(self)
```

```
| Methods inherited from Real:
  __abs__(a)
 __add__(a, b)
  __ge__(a, b)
      Return self>=value.
  __gt__(a, b)
     Return self>value.
  __le__(a, b)
     Return self<=value.
  __lt__(a, b)
    Return self<value.
 __mul__(a, b)
 __neg__(a)
 __pow__(a, b)
 __radd__(a, b)
  __rmul__(a, b)
 __rpow__(b, a)
 __rsub__(a, b)
| __rtruediv__(a, b)
| __sub__(a, b)
 __truediv__(a, b)
  disp(a)
```

```
| evstable(a, prec)
    | isNegative(self)
    | Data descriptors inherited from Real:
    | __dict__
        dictionary for instance variables (if defined)
    __weakref__
        list of weak references to the object (if defined)
   class RealSqrt(Real)
    | RealSqrt(a)
    | Method resolution order:
          RealSqrt
          Real
          builtins.object
    | Methods defined here:
    | init (self, a)
        Initialize self. See help(type(self)) for accurate
signature.
    | __repr__(self)
    | Return repr(self).
    | ev(self, n)
    | -----
    | Methods inherited from Real:
    | __abs__(a)
    | __add__(a, b)
```

```
__ge__(a, b)
    Return self>=value.
  __gt__(a, b)
      Return self>value.
  __le__(a, b)
    Return self<=value.
 __lt__(a, b)
    Return self<value.
 __mul__(a, b)
| __neg__(a)
 __pow__(a, b)
 __radd__(a, b)
 __rmul__(a, b)
 __rpow__(b, a)
| __rsub__(a, b)
 __rtruediv__(a, b)
| __sub__(a, b)
  __truediv__(a, b)
  disp(a)
| evstable(a, prec)
| isNegative(self)
| Data descriptors inherited from Real:
```

```
| dict
           dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
    class RealSubtract(Real)
     | RealSubtract(a, b)
     | Method resolution order:
           RealSubtract
           Real
           builtins.object
     | Methods defined here:
     | __init__(self, a, b)
          Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
          Return repr(self).
     | ev(self, n)
     | Methods inherited from Real:
     | __abs__(a)
     | __add__(a, b)
     | __ge__(a, b)
          Return self>=value.
     | __gt__(a, b)
           Return self>value.
     | __le__(a, b)
           Return self<=value.
```

```
__lt__(a, b)
      Return self<value.
   __mul__(a, b)
 | __neg__(a)
 | __pow__(a, b)
   __radd__(a, b)
   __rmul__(a, b)
   __rpow__(b, a)
 | __rsub__(a, b)
 | __rtruediv__(a, b)
   __sub__(a, b)
   __truediv__(a, b)
 | disp(a)
 | evstable(a, prec)
 | isNegative(self)
 | Data descriptors inherited from Real:
 | __dict__
       dictionary for instance variables (if defined)
   __weakref__
       list of weak references to the object (if defined)
class RealTan(Real)
 | RealTan(a)
```

```
Method resolution order:
           RealTan
           Real
           builtins.object
     | Methods defined here:
     | __init__(self, a)
           Initialize self. See help(type(self)) for accurate
signature.
       __repr__(self)
         Return repr(self).
     | ev(self, n)
     | Methods inherited from Real:
     | __abs__(a)
     | __add__(a, b)
     | __ge__(a, b)
         Return self>=value.
       __gt__(a, b)
          Return self>value.
       __le__(a, b)
         Return self<=value.
      __lt__(a, b)
          Return self<value.
     | __mul__(a, b)
       __neg__(a)
       __pow__(a, b)
```

```
__radd__(a, b)
       __rmul__(a, b)
     | __rpow__(b, a)
       __rsub__(a, b)
       __rtruediv__(a, b)
     | __sub__(a, b)
       __truediv__(a, b)
       disp(a)
       evstable(a, prec)
      isNegative(self)
       Data descriptors inherited from Real:
       __dict__
           dictionary for instance variables (if defined)
        weakref
           list of weak references to the object (if defined)
   class ShapeSampler(builtins.object)
     | ShapeSampler(inshape, dx=1, dy=1)
     | Methods defined here:
       __init__(self, inshape, dx=1, dy=1)
           Builds a sampler for random numbers (in the form of
PSRNs) on or inside a 2-dimensional shape.
           inshape is a function that takes three parameters (x, y,
s) and
           returns 1 if the box (x/s,y/s,(x+1)/s,(y+1)/s) is fully
in the shape;
```

```
-1 if not; and 0 if partially.
          dx and dy are the size of the bounding box and must be
integers. Default is 1 each.
    | sample(self, rg)
          Generates a random point inside the shape, in the form
of a uniform PSRN.
    | ------
    | Data descriptors defined here:
    | dict
          dictionary for instance variables (if defined)
      weakref
          list of weak references to the object (if defined)
   class ShapeSampler2(builtins.object)
    | ShapeSampler2(inshape, dx=1, dy=1)
    | Methods defined here:
    __init__(self, inshape, dx=1, dy=1)
          Builds a sampler for random numbers on or inside a 2-
dimensional shape.
          inshape is a function that takes a box described as
[[min1, max1], ..., [minN, maxN]]
          and returns 1 if the box is fully in the shape;
          -1 if not; and 0 if partially.
          dx and dy are the size of the bounding box and must be
integers. Default is 1 each.
    | sample(self, rg)
          Generates a random point inside the shape.
    _____
    | Data descriptors defined here:
    | __dict__
          dictionary for instance variables (if defined)
```

```
weakref
           list of weak references to the object (if defined)
     | Data and other attributes defined here:
     | MAYBE = 0
     | N0 = -1
     | YES = 1
   class SinFunction(builtins.object)
     | Methods defined here:
     | value(self, pt)
     | Data descriptors defined here:
     | __dict__
           dictionary for instance variables (if defined)
       __weakref__
           list of weak references to the object (if defined)
FUNCTIONS
   addto1(rg)
   bernoullinum(n)
   bernsteinDiff(coeffs, diff)
   betabin(k, psi, rho, cpsi, m=5)
   betadist(b, ax=1, ay=1, bx=1, by=1, precision=53)
   betadist_geobag(b, ax=1, ay=1, bx=1, by=1)
       Generates a beta-distributed random number with arbitrary
```

```
(user-defined) precision. Currently, this sampler only
works if (ax/ay) and
         (bx/by) are both 1 or greater, or if one of these
parameters is
        1 and the other is less than 1.
        - b: Bernoulli object (from the "bernoulli" module).
        - ax, ay: Numerator and denominator of first shape
parameter.
        - bx, by: Numerator and denominator of second shape
parameter.
        - precision: Number of bits after the point that the result
will contain.
    c2a(r=None)
    c4example()
    crudelog(av)
    exchangeable_bernoulli(p, d, lamda=None)
    exp_minus_x2y(rg, f, y, pwr=2)
        B(x) \rightarrow B(exp(-x*x*y))
    exp_minus_xy(rg, f, y)
        B(x) \rightarrow B(exp(-x*y))
    forsythe_prob(rg, m, n)
    forsythe prob2(rg, x)
    forsythe_prob3(rg, x)
    fpNormalROU(mu=0, sigma=1)
    fracAreClose(a, b, n)
    fracAreCloseND(an, ad, bn, bd, n)
    fracEV(sn, sd, n)
    gammaDist2()
```

```
gen to transition(s)
    genscore(psi, rho, m=5)
    genscore mean var(mean, vari, m=5)
    genshape(rg, inshape)
        Generates a random point inside a 2-dimensional shape, in
the form of a uniform PSRN.
        inshape is a function that takes three parameters (x, y, s)
and
        returns 1 if the box (x/s,y/s,(x+1)/s,(y+1)/s) is fully in
the shape;
        -1 if not; and 0 if partially.
    geobagcompare(bag, f)
        Returns 1 with probability f(U), where U is the value that
          the given geometric bag turns out to hold, or 0 otherwise.
          This method samples bits from the geometric bag as
necessary.
        - b: Geometric bag, that is, an ordinary Python list
           that holds a list of bits from left to
           right starting with the bit immediately after the binary
point.
           An item can contain the value None, which indicates an
           unsampled bit.
        - f: Function to run, which takes one parameter, namely a
'float'.
          Currently, this method assumes f is strictly increasing or
strictly decreasing.
          Note that this may suffer rounding and other approximation
          errors as a result. A more robust implementation would
require
         the method to return an interval (as in interval
arithmetic)
          or would pass the desired level of accuracy to the
function given
          here, and would probably have the function use arbitrary-
precision
          rational or floating-point numbers rather than the fixed-
precision
```

```
'float' type of Python, which usually has 53 bits of
precision.
    iteratedPoly2(func, n)
    iteratedPoly3(func, n)
    iteratedPolyExample()
    lah(n, k)
    logbinco(n, k)
    logbinprob(n, k)
    logconcave(f, c)
    loggammahelper(n, precision)
    logpoisson(lamda, n)
    logsmall(av, n)
    minDegree(maxValue, maxDeriv, epsilon, deriv=4)
    monoSecondMoment(secondMoment, pdf)
    muth(mu)
    powerOfUniform(b, px, py, precision=53)
        Generates a power of a uniform random number.
        - px, py - Numerator and denominator of desired exponent for
the uniform
          random number.
        - precision: Number of bits after the point that the result
will contain.
    proddist(x, a, b, c, d)
    proddist2(x, a, b, c, d)
    psrn add(rg, psrn1, psrn2, digits=2)
```

```
Adds two uniform partially-sampled random numbers.
        psrn1: List containing the sign, integer part, and
fractional part
            of the first PSRN. Fractional part is a list of digits
            after the point, starting with the first.
        psrn2: List containing the sign, integer part, and
fractional part
            of the second PSRN.
        digits: Digit base of PSRNs' digits. Default is 2, or
binary.
    psrn_add_fraction(rg, psrn, fraction, digits=2)
    psrn complement(x)
    psrn fill(rg, psrn, precision=53, digits=2)
    psrn in range(rg, bmin, bmax, digits=2)
    psrn_in_range_positive(rg, bmin, bmax, digits=2)
    psrn less(rg, psrn1, psrn2, digits=2)
    psrn less than fraction(rg, psrn, rat, digits=2)
    psrn multiply(rg, psrn1, psrn2, digits=2)
        Multiplies two uniform partially-sampled random numbers.
        psrn1: List containing the sign, integer part, and
fractional part
            of the first PSRN. Fractional part is a list of digits
            after the point, starting with the first.
        psrn2: List containing the sign, integer part, and
fractional part
            of the second PSRN.
        digits: Digit base of PSRNs' digits. Default is 2, or
binary.
    psrn multiply b(rg, psrn1, psrn2, digits=2, testing=False)
    psrn_multiply_by_fraction(rg, psrn1, fraction, digits=2)
        Multiplies a partially-sampled random number by a fraction.
        psrn1: List containing the sign, integer part, and
```

```
fractional part
            of the first PSRN. Fractional part is a list of digits
            after the point, starting with the first.
        fraction: Fraction to multiply by.
        digits: Digit base of PSRNs' digits. Default is 2, or
binary.
    psrn_new_01()
    psrn_reciprocal(rg, psrn1, digits=2)
        Generates the reciprocal of a partially-sampled random
number.
        psrn1: List containing the sign, integer part, and
fractional part
            of the first PSRN. Fractional part is a list of digits
            after the point, starting with the first.
        digits: Digit base of PSRNs' digits. Default is 2, or
binary.
    psrn_sample(rg, psrn, digits=2)
    psrnexpo(rg)
    randBernoulli(f)
    randLnUniform()
    randMax(n=2)
    randMin(n=2)
    randUniformLessThan(val)
    randUniformPower(pwr)
    rayleighpsrn(rg, s=1)
    realCeiling(a)
    realFloor(a)
    realGamma(ml)
```

```
realIsGreater(a, b)
    realIsLess(a, b)
    realIsLessOrEqual(a, b)
    realIsNegative(a)
    realNormalROU(mu=0, sigma=1)
    recordcount(n)
    sampleIntPlusBag(rg, psrn, k)
        Return 1 with probability (x+k)/2^bitlength(k).
        Ignores PSRN's integer part and sign.
    size biased poisson ailamujia(rg, eta=1)
        Hassan, A., Dar, S.A., et al., "On size biased Poisson
Ailamujia distribution and its applications",
        Pak. J. Statistics 37(1), 19-38, 2021.
    stirling1(n, k)
    truncated_gamma(rg, bern, ax, ay, precision=53)
        ######################
    tulap(m, b, q)
DATA
    ArcTanHTable = [0, 294906490, 137123709, 67461703, 33598225,
16782680,...
    CRUDELOG = [0, -726816, -681390, -654818, -635964, -621340,
-609392, -...
    CRUDELOG ARCTANBITDIFF = 13
    CRUDELOG ARCTANFRAC = 29
    CRUDELOG_BITS = 16
    CRUDELOG LOG2BITS = 45426
    CRUDELOG LOGMIN = 9830
    LNPOLY2 = [(-28986367995118693...8591117027361355259,
10000000000000000...
    LNPOLY3 = [(-13476514299119388...8263005361644498323,
```

```
500000000000000...
    REALHALFPI = RealPi(1/2)
    REALPI = RealPi(1)
    REAL_858_1000 = RealFraction(429/500)

FILE
/home/peter/Documents/SharpDevelopProjects/peteroupc.github.io/betad:
```