ACM/CS 114 Parallel algorithms for scientific applications

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Amortizing cost: containers

- ▶ the flexible solution is more expensive
 - this is sensible and, at some level, inevitable: flexibility implies runtime decisions
 - so we should be prepared to tolerate some slow down
- part of bad performance is caused by executing the *overhead* multiple times
- can we amortize the overhead over N, the total number of points?

Using a container for the sampling points

the base class, in PointCloud.py

```
class PointCloud(object):

"""

The abstract base class for point generators

interface

def points(self, n, box):

"""

Generate {n} random points on the interior of {box}

parameters:

(n): the number of points to generate
(box): a pair of points on the plane that specify the major diagonal of rectangular region

"""

raise NotImplementedError(

"class {.__name__!r} should implement 'points'".format(type(self)))
```

Filling a container with random points

in Mersenne.py

```
import random
from PointCloud import PointCloud
class Mersenne (PointCloud):
   # interface
   def points(self, n, box):
      # unfold the bounding box
      intervals = tuple(zip(*box))
      # create the container for the sample
      sample = []
      # loop over the sample size
      while n > 0.
         p = [ random.uniform(left, right) for left, right in intervals ]
         sample.append(p)
         n -= 1
      # return the samples
      return sample
```

Shape requires little modification

in Shape.py

```
1 class Shape(object):
2     """
3     The abstract base class for representations of geometrical regions
4     """
5     # interface
6     def interior(self, points):
8          """
9          Examine each point in {points} and return a container of only the interior points
1          """
2          raise NotImplementedError(
```

The updated interior of a circle, in Disk.py

```
from Shape import Shape
class Disk(Shape):
   def interior(self, points):
      r2 = self.radius**2
      x0, y0 = self.center
      keep = []
      for point in points:
         x, y = point
         dx = x - x0
         dy = y - y0
         if dx*dx + dy*dy <= r2:
            keep.append(point)
      return keep
   def init (self, radius=1.0, center=(0.0, 0.0)):
      self.radius = radius
      self.center = center
      return
```

The driver for the container based solution

```
def gauss():
   from Disk import Disk
   from Mersenne import Mersenne
   N = 10 * * 5
   box = [(0,0), (1,1)]
   generator = Mersenne()
   # the region of integration
   disk = Disk(center=(0,0), radius=1)
   # the integration algorithm
   sample = generator.points(N, box)
   interior = len(disk.interior(sample))
   # print out the estimate of \pi
   print("pi: {0:.8f}".format(4*interior/N))
   return
```

The performance of the container solution

	C++	python	naïve OO	containers
N	t(sec)	t(sec)	t(sec)	t(sec)
10^{0}	.002	.014	.014	.014
10^{1}	.002	.014	.014	.014
10^{2}	.002	.014	.014	.015
10^{3}	.002	.015	.020	.019
10^{4}	.004	.027	.078	.063
10^{5}	.026	.144	.625	.504
10^{6}	.230	1.265	6.242	5.925
10^{7}	2.277	12.624	61.583	188.318
10^{8}	22.749	130.430		
10^{9}	227.735			

Generators

- the container solution has a certain elegance
 - ▶ the creation of the container and its use are separated
 - we pay less overhead per point
 - with our points in containers, we can use the fast functional routines to operate on them
- but, we have now uncovered a new source of cost: managing memory
 - ▶ it appears that building large lists is expensive
 - can we avoid building the container all together?
- generators!

Generating random points

in Mersenne.py

```
import random
2 from PointCloud import PointCloud
 class Mersenne (PointCloud):
    def points(self, n, box):
       # unfold the bounding box
       intervals = tuple(zip(*box))
        # loop over the sample size
       while n > 0.
           p = [ random.uniform(*interval) for interval in intervals ]
           vield p
           n -= 1
        return
```

The modified Disk.py

```
from Shape import Shape
class Disk(Shape):
   def interior(self, points):
      r2 = self.radius**2
      x0. v0 = self.center
      for point in points:
         x, y = point
         dx = x - x0
         dy = y - y0
         if dx*dx + dy*dy <= r2:
            yield point
      return
   def init (self, radius=1.0, center=(0.0, 0.0)):
      self.radius = radius
      self.center = center
      return
```

The driver for the generator based solution

```
def gauss():
   from Disk import Disk
   from Mersenne import Mersenne
   N = 10 * *7
   box = [(0,0), (1,1)]
   generator = Mersenne()
   disk = Disk(center=(0,0), radius=1)
   sample = generator.points(N. box)
   interior = count(disk.interior(sample))
   # print out the estimate of \pi
   print("pi: {0:.8f}".format(4*interior/N))
   return
def count (iterable):
   counter = 0
   for item in iterable:
      counter += 1
   return counter
```

The performance of the generator solution

	C++	python	naïve OO	containers	generators
N	t(sec)	t(sec)	t(sec)	t(sec)	t(sec)
10^{0}	.002	.014	.014	.014	.014
10^{1}	.002	.014	.014	.014	.014
10^{2}	.002	.014	.014	.015	.014
10^{3}	.002	.015	.020	.019	.018
10^{4}	.004	.027	.078	.063	.053
10^{5}	.026	.144	.625	.504	.401
10^{6}	.230	1.265	6.242	5.925	3.780
10^{7}	2.277	12.624	61.583	188.318	38.242
10^{8}	22.749	130.430			
10^{9}	227.735				

Representing integrands

in Functor.py

```
class Functor(object):
    """
    The abstract base class for function objects
    """
    # interface
    def eval(self, points):
    """
    Evaluate the function at the supplied points
    """
    raise NotImplementedError(
    "class {.__name__!r} should implement 'eval'".format(type(self)))
```

Constant functions

in Constant.py

```
from Functor import Functor
class Constant (Functor):
   # interface
   def eval(self, points):
      # cache the constant
      constant = self.constant
     # return the constant regardless of the evaluation point
      for point in points: yield constant
      # all done
      return
   # meta methods
   def __init__(self, constant):
      self.constant = constant
      return
```

A non-trivial integrand

```
from Functor import Functor
3 class Gaussian (Functor):
     An implementation of the normal distribution with mean \mu and variance \sigma^2
     def eval(self, points):
        from math import exp, sqrt, pi
        mean = self.mean
        spread = self.spread
        normalization = 1 / sqrt(2*pi) / spread
        scaling = 2 * spread**2
        for p in points:
           r2 = sum((p_i - mean_i) **2 for p_i, mean_i in zip(p, mean))
           vield normalization * exp(- r2/scaling)
        return
     def init (self, mean, spread):
        self.mean = mean
        self.spread = spread
        return
```

The Monte Carlo integrator

```
1 import operator, functools
  def gauss():
     from Disk import Disk
     from Gaussian import Gaussian
     from Mersenne import Mersenne
     N = 10 * *7
     box = [(-1,-1), (1,1)]
     B = functools.reduce(operator.mul, ((right-left) for left,right in zip(*box)))
     generator = Mersenne()
     disk = Disk(center=(0,0), radius=1)
     qaussian = Gaussian (mean=(0,0), spread=1/3)
     sample = generator.points(N, box)
     interior = disk.interior(sample)
     integral = B/N * sum(gaussian.eval(interior))
     print("integral: {0:.8f}".format(integral))
     return
```