

Randomized Algorithms

Antonio Carzaniga

Faculty of Informatics
Università della Svizzera italiana

May 8, 2012

- Examples
- Las Vegas and Monte Carlo algorithms
- More examples

Remember Quick-Sort?

Remember Quick-Sort?

QUICKSORT($A, \textit{begin}, \textit{end}$)

1 **if** $\textit{begin} < \textit{end}$

2 $q = \mathbf{PARTITION}(A, \textit{begin}, \textit{end})$

3 **QUICKSORT**($A, \textit{begin}, q - 1$)

4 **QUICKSORT**($A, q + 1, \textit{end}$)

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )  
1  if  $begin < end$   
2       $q = \mathbf{PARTITION}(A, begin, end)$   
3      QUICKSORT( $A, begin, q - 1$ )  
4      QUICKSORT( $A, q + 1, end$ )
```

- Idea: *partition* the sequence $A[1 \dots n]$ in three parts
 - ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
 - ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
 - ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )  
1  if  $begin < end$   
2       $q = \mathbf{PARTITION}(A, begin, end)$   
3      QUICKSORT( $A, begin, q - 1$ )  
4      QUICKSORT( $A, q + 1, end$ )
```

- Idea: *partition* the sequence $A[1 \dots n]$ in three parts
 - ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
 - ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
 - ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v

2	36	5	21	1	13	11	20	5	4	8
---	----	---	----	---	----	----	----	---	---	---

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )
```

```
1  if  $begin < end$ 
```

```
2       $q = \mathbf{PARTITION}(A, begin, end)$ 
```

```
3      QUICKSORT( $A, begin, q - 1$ )
```

```
4      QUICKSORT( $A, q + 1, end$ )
```

■ Idea: *partition* the sequence $A[1 \dots n]$ in three parts

- ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
- ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
- ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v

2	36	5	21	1	13	11	20	5	4	8
---	----	---	----	---	----	----	----	---	---	---

$v = 8$

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )
```

```
1  if  $begin < end$ 
```

```
2       $q = \mathbf{PARTITION}(A, begin, end)$ 
```

```
3      QUICKSORT( $A, begin, q - 1$ )
```

```
4      QUICKSORT( $A, q + 1, end$ )
```

■ Idea: *partition* the sequence $A[1 \dots n]$ in three parts

- ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
- ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
- ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v

2	36	5	21	1	13	11	20	5	4	8
---	----	---	----	---	----	----	----	---	---	---

$v = 8$

2	4	1	5	5						
---	---	---	---	---	--	--	--	--	--	--

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )
```

```
1  if  $begin < end$ 
```

```
2       $q = \mathbf{PARTITION}(A, begin, end)$ 
```

```
3      QUICKSORT( $A, begin, q - 1$ )
```

```
4      QUICKSORT( $A, q + 1, end$ )
```

- Idea: *partition* the sequence $A[1 \dots n]$ in three parts

- ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
- ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
- ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v

2	36	5	21	1	13	11	20	5	4	8
---	----	---	----	---	----	----	----	---	---	---

$v = 8$

2	4	1	5	5	8					
---	---	---	---	---	---	--	--	--	--	--

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )
```

```
1  if  $begin < end$ 
```

```
2       $q = \mathbf{PARTITION}(A, begin, end)$ 
```

```
3      QUICKSORT( $A, begin, q - 1$ )
```

```
4      QUICKSORT( $A, q + 1, end$ )
```

- Idea: *partition* the sequence $A[1 \dots n]$ in three parts

- ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
- ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
- ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v

2	36	5	21	1	13	11	20	5	4	8
---	----	---	----	---	----	----	----	---	---	---

$v = 8$

2	4	1	5	5	8	11	20	13	36	21
---	---	---	---	---	---	----	----	----	----	----

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )
```

```
1  if  $begin < end$ 
```

```
2       $q = \mathbf{PARTITION}(A, begin, end)$ 
```

```
3      QUICKSORT( $A, begin, q - 1$ )
```

```
4      QUICKSORT( $A, q + 1, end$ )
```

- Idea: *partition* the sequence $A[1 \dots n]$ in three parts
 - ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
 - ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
 - ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v

2	36	5	21	1	13	11	20	5	4	8
---	----	---	----	---	----	----	----	---	---	---

$v = 8$

$q = 6$

2	4	1	5	5	8	11	20	13	36	21
---	---	---	---	---	---	----	----	----	----	----

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )  
1  if  $begin < end$   
2       $q = \mathbf{PARTITION}(A, begin, end)$   
3      QUICKSORT( $A, begin, q - 1$ )  
4      QUICKSORT( $A, q + 1, end$ )
```

- Idea: *partition* the sequence $A[1 \dots n]$ in three parts
 - ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
 - ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
 - ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v

2	36	5	21	1	13	11	20	5	4	8
---	----	---	----	---	----	----	----	---	---	---

$v = 8$

$q = 6$

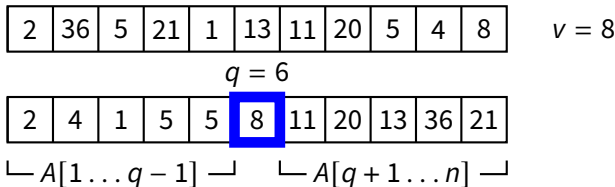
2	4	1	5	5	8	11	20	13	36	21
---	---	---	---	---	---	----	----	----	----	----

└─ $A[1 \dots q - 1]$ ─┘

Remember Quick-Sort?

```
QUICKSORT( $A, begin, end$ )
1  if  $begin < end$ 
2       $q = \mathbf{PARTITION}(A, begin, end)$ 
3      QUICKSORT( $A, begin, q - 1$ )
4      QUICKSORT( $A, q + 1, end$ )
```

- Idea: *partition* the sequence $A[1 \dots n]$ in three parts
 - ▶ $A[1 \dots q - 1]$ contain elements that are *less than or equal to* v
 - ▶ $A[q] = v$ is the “pivot” value ($v \in A$)
 - ▶ $A[q + 1 \dots n]$ contain elements that are *greater than* v



A Classic Divide-and-Conquer

- **Divide:** partition A in $A[1 \dots q - 1]$ and $A[q + 1 \dots n]$ such that...

A Classic Divide-and-Conquer

- **Divide:** partition A in $A[1 \dots q - 1]$ and $A[q + 1 \dots n]$ such that...
- **Conquer:** sort $A[1 \dots q - 1]$ and $A[q + 1 \dots n]$

A Classic Divide-and-Conquer

- **Divide:** partition A in $A[1 \dots q - 1]$ and $A[q + 1 \dots n]$ such that...
- **Conquer:** sort $A[1 \dots q - 1]$ and $A[q + 1 \dots n]$
- **Combine:** nothing to do here
 - ▶ it is all in the *partition* algorithm

A Classic Divide-and-Conquer

- **Divide:** partition A in $A[1 \dots q - 1]$ and $A[q + 1 \dots n]$ such that...
- **Conquer:** sort $A[1 \dots q - 1]$ and $A[q + 1 \dots n]$
- **Combine:** nothing to do here
 - ▶ it is all in the *partition* algorithm

```
PARTITION( $A$ ,  $begin$ ,  $end$ )
```

```
1  $q = begin$ 
```

```
2  $v = A[end]$  // we deterministically choose  $v$ 
```

```
3 for  $i = begin$  to  $end$ 
```

```
4     if  $A[i] \leq v$ 
```

```
5          $swap(A[i], A[q])$ 
```

```
6          $q = q + 1$ 
```

```
7 return  $q - 1$ 
```

Complexity of QUICKSORT

QUICKSORT($A, \textit{begin}, \textit{end}$)

1 **if** $\textit{begin} < \textit{end}$

2 $q = \mathbf{PARTITION}(A, \textit{begin}, \textit{end})$

3 **QUICKSORT**($A, \textit{begin}, q - 1$)

4 **QUICKSORT**($A, q + 1, \textit{end}$)

Complexity of QUICKSORT

```
QUICKSORT(A, begin, end)  
1  if begin < end  
2      q = PARTITION(A, begin, end)  
3      QUICKSORT(A, begin, q - 1)  
4      QUICKSORT(A, q + 1, end)
```

- *Worst case: $q = 1$ or $q = n$*

Complexity of QUICKSORT

```
QUICKSORT(A, begin, end)  
1  if begin < end  
2      q = PARTITION(A, begin, end)  
3      QUICKSORT(A, begin, q - 1)  
4      QUICKSORT(A, q + 1, end)
```

- *Worst case*: $q = 1$ or $q = n$
 - ▶ the partition transforms P of size n in P of size $n - 1$, so $T(n) = T(n - 1) + \Theta(n)$

Complexity of QUICKSORT

```
QUICKSORT(A, begin, end)  
1  if begin < end  
2      q = PARTITION(A, begin, end)  
3      QUICKSORT(A, begin, q - 1)  
4      QUICKSORT(A, q + 1, end)
```

- *Worst case*: $q = 1$ or $q = n$
 - ▶ the partition transforms P of size n in P of size $n - 1$, so
 $T(n) = T(n - 1) + \Theta(n)$

$$T(n) = \Theta(n^2)$$

Complexity of QUICKSORT

```
QUICKSORT(A, begin, end)  
1  if begin < end  
2      q = PARTITION(A, begin, end)  
3      QUICKSORT(A, begin, q - 1)  
4      QUICKSORT(A, q + 1, end)
```

■ *Worst case*: $q = 1$ or $q = n$

- ▶ the partition transforms P of size n in P of size $n - 1$, so
 $T(n) = T(n - 1) + \Theta(n)$

$$T(n) = \Theta(n^2)$$

■ *Best case*: $q = \lceil n/2 \rceil$

Complexity of QUICKSORT

```
QUICKSORT(A, begin, end)  
1  if begin < end  
2      q = PARTITION(A, begin, end)  
3      QUICKSORT(A, begin, q - 1)  
4      QUICKSORT(A, q + 1, end)
```

■ *Worst case:* $q = 1$ or $q = n$

- ▶ the partition transforms P of size n in P of size $n - 1$, so
 $T(n) = T(n - 1) + \Theta(n)$

$$T(n) = \Theta(n^2)$$

■ *Best case:* $q = \lceil n/2 \rceil$

- ▶ the partition transforms P of size n into *two* problems P of size $\lfloor n/2 \rfloor$ and $\lceil n/2 \rceil - 1$, respectively; so $T(n) = 2T(n/2) + \Theta(n)$

Complexity of QUICKSORT

```
QUICKSORT(A, begin, end)
```

```
1  if begin < end
```

```
2      q = PARTITION(A, begin, end)
```

```
3      QUICKSORT(A, begin, q - 1)
```

```
4      QUICKSORT(A, q + 1, end)
```

■ Worst case: $q = 1$ or $q = n$

- ▶ the partition transforms P of size n in P of size $n - 1$, so
 $T(n) = T(n - 1) + \Theta(n)$

$$T(n) = \Theta(n^2)$$

■ Best case: $q = \lceil n/2 \rceil$

- ▶ the partition transforms P of size n into two problems P of size $\lfloor n/2 \rfloor$ and $\lceil n/2 \rceil - 1$, respectively; so $T(n) = 2T(n/2) + \Theta(n)$

$$T(n) = \Theta(n \log n)$$

Complexity of QUICKSORT

```
QUICKSORT(A, begin, end)  
1  if begin < end  
2      q = PARTITION(A, begin, end)  
3      QUICKSORT(A, begin, q - 1)  
4      QUICKSORT(A, q + 1, end)
```

- Worst case: $q = 1$ or $q = n$

not so improbable!

- ▶ the partition transforms P of size n in P of size $n - 1$, so
 $T(n) = T(n - 1) + \Theta(n)$

$$T(n) = \Theta(n^2)$$

- Best case: $q = \lceil n/2 \rceil$

- ▶ the partition transforms P of size n into two problems P of size $\lfloor n/2 \rfloor$ and $\lceil n/2 \rceil - 1$, respectively; so $T(n) = 2T(n/2) + \Theta(n)$

$$T(n) = \Theta(n \log n)$$

A Randomized Solution

A Randomized Solution

- Simple

```
RANDOMIZED-PARTITION(A, begin, end)
```

```
1 i = RANDOM(begin, end)
```

```
2 swap(A[i], A[end])
```

```
3 return PARTITION(A, begin, end)
```

A Randomized Solution

■ Simple

```
RANDOMIZED-PARTITION(A, begin, end)
```

```
1 i = RANDOM(begin, end)
```

```
2 swap(A[i], A[end])
```

```
3 return PARTITION(A, begin, end)
```

```
QUICKSORT(A, begin, end)
```

```
1 if begin < end
```

```
2     q = RANDOMIZED-PARTITION(A, begin, end)
```

```
3     QUICKSORT(A, begin, q - 1)
```

```
4     QUICKSORT(A, q + 1, end)
```

Other Examples

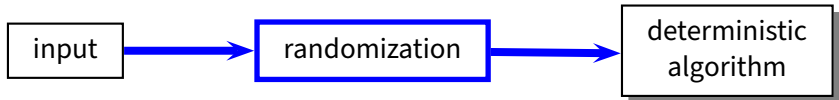
TREE-RANDOMIZED-INSERT(t, z)

```
1  if  $t = \text{NIL}$ 
2      return  $z$ 
3   $r =$  uniformly random value from  $\{1, \dots, \text{size}(t) + 1\}$ 
4  if  $r = 1$                                 //  $\text{Pr}[r = 1] = 1/(\text{size}(t) + 1)$ 
5       $\text{size}(z) = \text{size}(t) + 1$ 
6      return TREE-ROOT-INSERT( $t, z$ )
7  if  $\text{key}(z) < \text{key}(t)$ 
8       $\text{left}(t) =$  TREE-RANDOMIZED-INSERT( $\text{left}(t), z$ )
9  else  $\text{right}(t) =$  TREE-RANDOMIZED-INSERT( $\text{right}(t), z$ )
10  $\text{size}(t) = \text{size}(t) + 1$ 
11 return  $t$ 
```

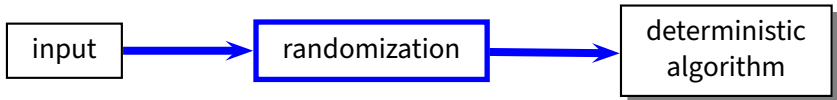

- We develop *randomized variants* of algorithms

- We develop *randomized variants* of algorithms
- The idea is to make the algorithm run with their *good complexity* ***with high probability*** on every input

- We develop *randomized variants* of algorithms
- The idea is to make the algorithm run with their *good complexity* **with high probability** on every input
- The algorithm does not change that much, though



- We develop *randomized variants* of algorithms
- The idea is to make the algorithm run with their *good complexity* **with high probability** on every input
- The algorithm does not change that much, though



- However, we can do a lot more with randomized algorithms...

■ Problem: *find a zero bit*

- ▶ *Input*: an array A , of n bits, containing more or less the same number of 1s and 0s (*hamming weight* is roughly $n/2$)
- ▶ *Output*: i such that $A[i] = 0$, or NIL if none exists

■ Problem: *find a zero bit*

- ▶ *Input*: an array A , of n bits, containing more or less the same number of 1s and 0s (*hamming weight* is roughly $n/2$)
- ▶ *Output*: i such that $A[i] = 0$, or NIL if none exists

■ Obvious solution

FIND-A-ZERO-BIT(A)

```
1  for  $i = 1$  to  $|A|$ 
2      if  $A[i] == 0$ 
3          return  $i$ 
4  return NIL
```

■ Problem: *find a zero bit*

- ▶ *Input*: an array A , of n bits, containing more or less the same number of 1s and 0s (*hamming weight* is roughly $n/2$)
- ▶ *Output*: i such that $A[i] = 0$, or NIL if none exists

■ Obvious solution

FIND-A-ZERO-BIT(A)

```
1  for  $i = 1$  to  $|A|$ 
2      if  $A[i] == 0$ 
3          return  $i$ 
4  return NIL
```

■ Problems?

- ▶ what if A is sorted in reverse order? (all 1-bits before the 0-bits)

■ Problem: *find a zero bit*

- ▶ *Input*: an array A , of n bits, containing more or less the same number of 1s and 0s (*hamming weight* is roughly $n/2$)
- ▶ *Output*: i such that $A[i] = 0$, or NIL if none exists

■ Obvious solution

FIND-A-ZERO-BIT(A)

```
1  for  $i = 1$  to  $|A|$ 
2      if  $A[i] == 0$ 
3          return  $i$ 
4  return NIL
```

■ Problems?

- ▶ what if A is sorted in reverse order? (all 1-bits before the 0-bits)
- ▶ any *deterministic* search strategy is vulnerable

A Randomized Algorithm

A Randomized Algorithm

- Take one

RANDOMIZED-FIND-A-ZERO-BIT1(A)

```
1 repeat  
2      $i = \mathbf{RANDOM}(1, |A|)$   
3 until  $A[i] == 0$   
4 return  $i$ 
```

A Randomized Algorithm

- Take one

```
RANDOMIZED-FIND-A-ZERO-BIT1(A)
```

```
1 repeat  
2      $i = \mathbf{RANDOM}(1, |A|)$   
3 until  $A[i] == 0$   
4 return  $i$ 
```

- ▶ expected iterations: 2

A Randomized Algorithm

■ Take one

RANDOMIZED-FIND-A-ZERO-BIT1(A)

```
1 repeat  
2      $i = \mathbf{RANDOM}(1, |A|)$   
3 until  $A[i] == 0$   
4 return  $i$ 
```

▶ expected iterations: 2

▶ worst-case:

A Randomized Algorithm

■ Take one

RANDOMIZED-FIND-A-ZERO-BIT1(A)

```
1 repeat  
2      $i = \mathbf{RANDOM}(1, |A|)$   
3 until  $A[i] == 0$   
4 return  $i$ 
```

- ▶ expected iterations: 2
- ▶ worst-case: unbounded!

A Randomized Algorithm

■ Take one

RANDOMIZED-FIND-A-ZERO-BIT1(A)

```
1 repeat  
2      $i = \mathbf{RANDOM}(1, |A|)$   
3 until  $A[i] == 0$   
4 return  $i$ 
```

- ▶ expected iterations: 2
- ▶ worst-case: unbounded!

■ Take two

RANDOMIZED-FIND-A-ZERO-BIT2(A)

```
1 for  $j = 1$  to  $k$   
2      $i = \mathbf{RANDOM}(1, |A|)$   
3     if  $A[i] == 0$   
4         return  $i$   
5 return NIL
```

A Randomized Algorithm

■ Take one

RANDOMIZED-FIND-A-ZERO-BIT1(A)

```
1 repeat  
2      $i = \mathbf{RANDOM}(1, |A|)$   
3 until  $A[i] == 0$   
4 return  $i$ 
```

- ▶ expected iterations: 2
- ▶ worst-case: unbounded!

■ Take two

RANDOMIZED-FIND-A-ZERO-BIT2(A)

```
1 for  $j = 1$  to  $k$   
2      $i = \mathbf{RANDOM}(1, |A|)$   
3     if  $A[i] == 0$   
4         return  $i$   
5 return NIL
```

- ▶ worst-case iterations: k

A Randomized Algorithm

■ Take one

RANDOMIZED-FIND-A-ZERO-BIT1(A)

```
1 repeat  
2      $i = \mathbf{RANDOM}(1, |A|)$   
3 until  $A[i] == 0$   
4 return  $i$ 
```

- ▶ expected iterations: 2
- ▶ worst-case: unbounded!

■ Take two

RANDOMIZED-FIND-A-ZERO-BIT2(A)

```
1 for  $j = 1$  to  $k$   
2      $i = \mathbf{RANDOM}(1, |A|)$   
3     if  $A[i] == 0$   
4         return  $i$   
5 return NIL
```

- ▶ worst-case iterations: k
- ▶ worst-case: wrong result!

A Randomized Algorithm

■ Take one

RANDOMIZED-FIND-A-ZERO-BIT1(A)

```
1 repeat  
2      $i = \mathbf{RANDOM}(1, |A|)$   
3 until  $A[i] == 0$   
4 return  $i$ 
```

- ▶ expected iterations: 2
- ▶ worst-case: unbounded!
- ▶ *Las Vegas*

■ Take two

RANDOMIZED-FIND-A-ZERO-BIT2(A)

```
1 for  $j = 1$  to  $k$   
2      $i = \mathbf{RANDOM}(1, |A|)$   
3     if  $A[i] == 0$   
4         return  $i$   
5 return NIL
```

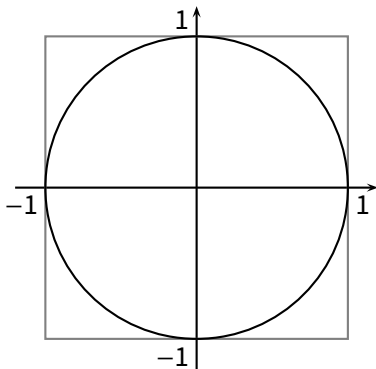
- ▶ worst-case iterations: k
- ▶ worst-case: wrong result!
- ▶ *Monte Carlo*

Simple Monte Carlo Algorithm

Simple Monte Carlo Algorithm

- Problem: *compute the surface of the unit disc*
 - ▶ you don't know the value of π —you may not even know that $S = \pi r^2$, but you know that a point

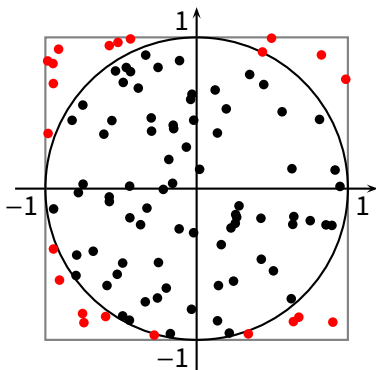
$$(x, y) \text{ is } \begin{cases} \text{outside the unit disc if} & x^2 + y^2 > 1 \\ \text{inside the unit disc if} & x^2 + y^2 \leq 1 \end{cases}$$



Simple Monte Carlo Algorithm

- Problem: *compute the surface of the unit disc*
 - ▶ you don't know the value of π —you may not even know that $S = \pi r^2$, but you know that a point

$$(x, y) \text{ is } \begin{cases} \text{outside the unit disc if} & x^2 + y^2 > 1 \\ \text{inside the unit disc if} & x^2 + y^2 \leq 1 \end{cases}$$

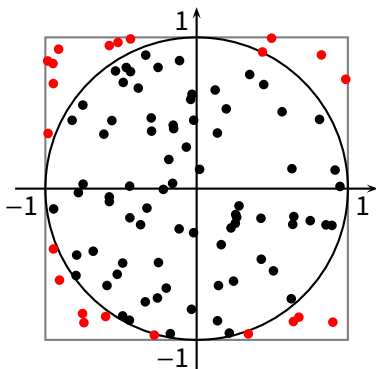


Simple Monte Carlo Algorithm

■ Problem: *compute the surface of the unit disc*

- ▶ you don't know the value of π —you may not even know that $S = \pi r^2$, but you know that a point

$$(x, y) \text{ is } \begin{cases} \text{outside the unit disc if } x^2 + y^2 > 1 \\ \text{inside the unit disc if } x^2 + y^2 \leq 1 \end{cases}$$



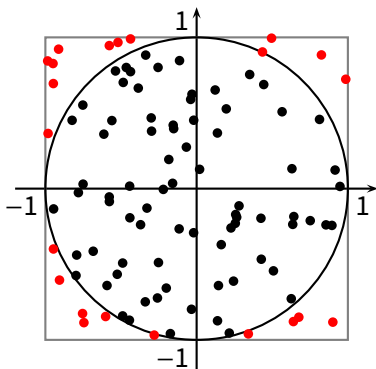
78 out of 100 points are *inside*

Simple Monte Carlo Algorithm

■ Problem: *compute the surface of the unit disc*

- ▶ you don't know the value of π —you may not even know that $S = \pi r^2$, but you know that a point

$$(x, y) \text{ is } \begin{cases} \text{outside the unit disc if } & x^2 + y^2 > 1 \\ \text{inside the unit disc if } & x^2 + y^2 \leq 1 \end{cases}$$



78 out of 100 points are *inside*

$$S = 4 \times 78/100 = 3.12$$

MONTE-CARLO-PI(n)

```
1   $i = 0$ 
2  for  $j = 1$  to  $n$ 
3       $x = \mathbf{RANDOM}(-1, 1)$ 
4       $y = \mathbf{RANDOM}(-1, 1)$ 
5      if  $x^2 + y^2 \leq 1$ 
6           $i = i + 1$ 
7  return  $4i/n$ 
```

MONTE-CARLO-PI(n)

```
1   $i = 0$ 
2  for  $j = 1$  to  $n$ 
3       $x = \text{RANDOM}(-1, 1)$ 
4       $y = \text{RANDOM}(-1, 1)$ 
5      if  $x^2 + y^2 \leq 1$ 
6           $i = i + 1$ 
7  return  $4i/n$ 
```

- The precision grows with n
 - ▶ more specifically, the *expected* precision grows with n

MONTE-CARLO-PI(n)

```
1   $i = 0$ 
2  for  $j = 1$  to  $n$ 
3       $x = \text{RANDOM}(-1, 1)$ 
4       $y = \text{RANDOM}(-1, 1)$ 
5      if  $x^2 + y^2 \leq 1$ 
6           $i = i + 1$ 
7  return  $4i/n$ 
```

- The precision grows with n
 - ▶ more specifically, the *expected* precision grows with n
- It is also easy to think about a better (adaptive) *stopping condition* other than going through n loops