

# 17. Randomized Algorithms: Why Randomness Helps

## Phase 6: Real-World Applications & Limitations

⌚ ~40 minutes | 🧠 Deterministic Algorithms Can Fail | 🚀 Randomness as a Feature

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## What Problem This Solves

You encounter:

- Hash table collisions in adversarial inputs (DOS attack)
- Quicksort worst-case  $O(n^2)$  behavior
- Load balancing across servers causing hotspots
- Random sampling for analyzing huge datasets
- "Why is randomness ever good? Shouldn't we be deterministic?"
- Distributed systems where randomness reduces collisions

**Without randomized algorithm intuition**, you either avoid randomness ("I want reproducible results!") or use it arbitrarily. You're vulnerable to adversarial inputs that exploit deterministic patterns.

**With randomized thinking**, you leverage randomness to break worst-case inputs, balance loads, and solve hard problems with good expected behavior.

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## Intuition & Mental Model

### The Core Insight: Randomness Breaks Patterns

Adversarial Scenario:

- Attacker sends hash collisions → hash table slow
- Adversary orders input → quicksort degrades
- Requests concentrate on one server → that server overloaded

Randomized Solution:

- Hash function randomized → attacker can't predict collisions
- Pivot selection randomized → unlikely to always pick bad pivots
- Request routing randomized → load spreads naturally
- Attacker's advantage disappears!

### Mental Model: The Adversary-Free Future

Deterministic: Input adversary can exploit → Worst case often achievable

Randomized: Adversary can't know random choices → Can't force worst case  
Only exponentially unlikely inputs fail

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## Core Concepts

### 1. Randomized Quicksort

```

function quicksortDeterministic(arr, low = 0, high = arr.length - 1) {
  // Deterministic pivot (e.g., first element)
  // Adversary sends reversed array →  $O(n^2)$  behavior

  if (low >= high) return;

  const pivot = arr[low]; // Naive choice (vulnerable!)
  let i = low + 1, j = high;

  while (i <= j) {
    while (i <= j && arr[i] <= pivot) i++;
    while (i <= j && arr[j] > pivot) j--;
    if (i < j) [arr[i], arr[j]] = [arr[j], arr[i]];
  }

  [arr[low], arr[j]] = [arr[j], arr[low]];

  quicksortDeterministic(arr, low, j - 1);
  quicksortDeterministic(arr, j + 1, high);
}

function quicksortRandomized(arr, low = 0, high = arr.length - 1) {
  // Randomized pivot (pick random element)
  // No adversarial input →  $O(n \log n)$  expected behavior

  if (low >= high) return;

  // Pick random pivot
  const randomIdx = low + Math.floor(Math.random() * (high - low + 1));
  const pivot = arr[randomIdx];
  [arr[low], arr[randomIdx]] = [arr[randomIdx], arr[low]]; // Swap to front

  let i = low + 1, j = high;

  while (i <= j) {
    while (i <= j && arr[i] <= pivot) i++;
    while (i <= j && arr[j] > pivot) j--;
    if (i < j) [arr[i], arr[j]] = [arr[j], arr[i]];
  }

  [arr[low], arr[j]] = [arr[j], arr[low]];

  quicksortRandomized(arr, low, j - 1);
  quicksortRandomized(arr, j + 1, high);
}

// Adversary's attack: Sorted/reverse array
const attack = Array.from({ length: 1000 }, (_, i) => i).reverse();

console.time('Deterministic');
quicksortDeterministic([...attack]);

```

```

console.timeEnd('Deterministic'); // ~100ms (quadratic!)

console.time('Randomized');
quicksortRandomized([...attack]);
console.timeEnd('Randomized'); // ~2ms (linear-ish!)

```

### Why It Works:

Deterministic pivot: Adversary picks input to make it worst  
 Randomized pivot: Pivot likely good (random choice)  
     Expected:  $O(n \log n)$  even against adversary  
     Probability of bad pivot: exponentially small

## 2. Hashing with Randomization

```

function hashTableAttack() {
  // Deterministic hash:  $h(x) = x \% \text{tableSize}$ 

  function simpleHash(x) {
    return x % 1000; // Known pattern
  }

  // Attacker sends: 1000, 2000, 3000, ...
  // All hash to 0 → Hash table becomes linked list →  $O(n)$  lookup!

  return {
    attack: 'Send keys that hash to same bucket: 1000, 2000, 3000',
    result: 'All collide →  $O(1)$  becomes  $O(n)$  (table degrades)'
  };
}

function randomizedHashing() {
  // Universal hashing: Random hash function from family

  class UniversalHash {
    constructor(tableSize = 1000) {
      this.size = tableSize;
      this.prime = 1000000007;
      this.a = 1 + Math.floor(Math.random() * (this.prime - 1));
      this.b = Math.floor(Math.random() * this.prime);
    }

    hash(x) {
      //  $h(x) = (ax + b) \bmod \text{prime} \bmod \text{tableSize}$ 
      return ((this.a * x + this.b) % this.prime) % this.size;
    }
  }

  // Same attack: 1000, 2000, 3000, ...
  const h = new UniversalHash(1000);

```

```

console.log([1000, 2000, 3000].map(x => h.hash(x)));
// Different hashes (random function chosen at init)
// Attack fails because attacker doesn't know a, b

// Key insight: Even if adversary knows inputs, hash function is random
// Expected collisions =  $O(1)$  per operation
}

```

### 3. Load Balancing with Randomization

```

function loadBalancing() {
  // Problem: How to distribute requests across servers?

  class DeterministicBalancer {
    constructor(serverCount = 10) {
      this.servers = Array(serverCount).fill(0);
      this.nextServer = 0;
    }

    route(request) {
      // Round-robin (deterministic)
      const server = this.nextServer % this.servers.length;
      this.servers[server]++;
      this.nextServer++;
      return server;
    }
  }

  class RandomizedBalancer {
    constructor(serverCount = 10) {
      this.servers = Array(serverCount).fill(0);
    }

    route(request) {
      // Pick random server (with probability proportional to capacity)
      // Or simpler: just pick random
      const server = Math.floor(Math.random() * this.servers.length);
      this.servers[server]++;
      return server;
    }
  }

  // Adversary sends: requests that align with round-robin
  // Knows pattern → Can predict which server gets each request

  const det = new DeterministicBalancer(5);
  for (let i = 0; i < 20; i++) {
    det.route(i);
  }
  console.log('Deterministic load:', det.servers); // [4, 4, 4, 4, 4]
  // Looks balanced! But if adversary sends bursts at certain times...

```

```

const rand = new RandomizedBalancer(5);
for (let i = 0; i < 20; i++) {
  rand.route(i);
}
console.log('Randomized load:', rand.servers); // [3, 4, 4, 5, 4]
// Also balanced, but adversary can't exploit pattern

// Worst case for randomized:
// Probability all 20 go to one server:  $(1/5)^{20} \approx 10^{-14}$ 
}

```

#### 4. Las Vegas vs Monte Carlo Algorithms

```

function algorithmClasses() {
  // Las Vegas: Always correct, randomized time
  // Monte Carlo: Randomized correctness, fixed time

  // Las Vegas: Randomized Quicksort
  // Always produces sorted array, but time varies
  function lasVegas(arr) {
    // Guaranteed correct, expected  $O(n \log n)$ 
    quicksortRandomized([...arr]);
    return arr; // Correct!
  }

  // Monte Carlo: Randomized Primality Test
  // Fixed time, small probability of error
  function millerRabin(n, k = 40) {
    // Runs in  $O(k \log n)$ 
    // Returns "probably prime" (error prob <  $2^{-k}$ )

    if (n < 2) return false;
    if (n === 2 || n === 3) return true;
    if (n % 2 === 0) return false;

    let r = 0, d = n - 1;
    while (d % 2 === 0) {
      r++;
      d /= 2;
    }

    for (let i = 0; i < k; i++) {
      const a = 2 + Math.floor(Math.random() * (n - 4));
      let x = Math.pow(a, d) % n; // Simplified

      if (x === 1 || x === n - 1) continue;

      let composite = true;
      for (let j = 0; j < r - 1; j++) {
        x = (x * x) % n;

```

```

        if (x === n - 1) {
            composite = false;
            break;
        }
    }

    if (composite) return false;
}

return true; // "Probably prime" (error < 2-40)
}

return {
    lasVegas: {
        correctness: 'Always correct',
        time: 'Randomized (expected O(n log n), worst O(n2) with prob 2-n)',
        example: 'Quicksort, randomized search'
    },
    monteCarlo: {
        correctness: 'Probably correct (bounded error)',
        time: 'Deterministic O(k log n)',
        example: 'Primality testing, randomized approximation'
    }
};
}

```

## 5. Randomized Data Structures

```

function skipList() {
    // Skip List: Like linked list but with "express lanes"
    // Search O(log n) expected time with randomization

    class SkipNode {
        constructor(value, level) {
            this.value = value;
            this.next = Array(level).fill(null);
        }
    }

    class SkipList {
        constructor(maxLevel = 16, p = 0.5) {
            this.maxLevel = maxLevel;
            this.p = p; // Probability of promoting to next level
            this.head = new SkipNode(-Infinity, maxLevel);
            this.level = 0;
        }

        randomLevel() {
            let level = 0;
            while (Math.random() < this.p && level < this.maxLevel - 1) {
                level++;
            }
        }
    }
}

```

```

    }
    return level;
}

search(value) {
    let current = this.head;

    for (let i = this.level; i >= 0; i--) {
        while (current.next[i] && current.next[i].value < value) {
            current = current.next[i];
        }
    }

    current = current.next[0];
    return current && current.value === value;
}

insert(value) {
    const newLevel = this.randomLevel();
    const newNode = new SkipNode(value, newLevel + 1);

    // Simplified insert logic
    let current = this.head;
    const updates = Array(newLevel + 1).fill(this.head);

    for (let i = Math.min(newLevel, this.level); i >= 0; i--) {
        while (current.next[i] && current.next[i].value < value) {
            current = current.next[i];
        }
        updates[i] = current;
    }

    for (let i = 0; i <= newLevel; i++) {
        newNode.next[i] = updates[i].next[i];
        updates[i].next[i] = newNode;
    }

    if (newLevel > this.level) {
        this.level = newLevel;
    }
}

const skip = new SkipList();
[3, 7, 1, 9, 2, 5].forEach(v => skip.insert(v));

console.log(skip.search(7)); // true
console.log(skip.search(10)); // false

// Expected search time: O(log n)

```

```
// Much simpler than balanced trees!  
}
```

## 6. Shuffle and Sampling

```
function randomSampling() {  
  // Fisher-Yates Shuffle: Uniformly random permutation  
  
  function shuffle(arr) {  
    for (let i = arr.length - 1; i > 0; i--) {  
      const j = Math.floor(Math.random() * (i + 1));  
      [arr[i], arr[j]] = [arr[j], arr[i]];  
    }  
    return arr;  
  }  
  
  // Reservoir Sampling: Sample k items from stream of unknown length  
  
  function reservoirSample(stream, k) {  
    const reservoir = stream.slice(0, k);  
  
    for (let i = k; i < stream.length; i++) {  
      const j = Math.floor(Math.random() * (i + 1));  
      if (j < k) {  
        reservoir[j] = stream[i];  
      }  
    }  
  
    return reservoir;  
  }  
  
  const data = Array.from({ length: 10000 }, (_, i) => i);  
  
  // Shuffle  
  console.log(shuffle([1, 2, 3, 4, 5])); // Random order each time  
  
  // Sample 10 items from 10000 uniformly  
  console.log(reservoirSample(data, 10).length); // 10  
  
  return {  
    shuffleTime: 'O(n)',  
    reservoirTime: 'O(stream length)',  
    uniformity: 'Each permutation equally likely'  
  };  
}
```

---

## Software Engineering Connections

### 1. Hash Table Implementation



```
function cryptographicHashTable() {
  // Use a randomized seed for hash function

  const seed = Math.random() * 1e9;

  function hash(x) {
    return ((x * 2654435761 + seed) >>> 0) % 1000; // Multiply-shift hash
  }

  // Same malicious input: 1000, 2000, 3000, ...
  // Hash function different on each run → Adversary can't exploit
  // Expected collision rate: 1/tableSize regardless of input
}
```

## 2. Distributed Systems: Random Backoff

```
function exponentialBackoffRandomized() {
  // When retrying failed operation, add randomness to avoid thundering herd

  async function retryWithBackoff(operation, maxAttempts = 5) {
    for (let attempt = 0; attempt < maxAttempts; attempt++) {
      try {
        return await operation();
      } catch (error) {
        if (attempt === maxAttempts - 1) throw error;

        // Exponential backoff with jitter
        const baseDelay = Math.pow(2, attempt) * 1000; // 1s, 2s, 4s, 8s, 16s
        const jitter = Math.random() * baseDelay;
        const delay = baseDelay + jitter;

        console.log(`Retry ${attempt + 1} after ${delay.toFixed(0)}ms`);
        await new Promise(resolve => setTimeout(resolve, delay));
      }
    }
  }

  // Without jitter: All clients retry at T=1s, T=2s, T=4s → Thundering herd
  // With jitter: Retries spread out randomly → Smooth load
}
```

## 3. A/B Testing: Random Assignment

```
function randomABTest() {
  function getUserGroup(userId, testId) {
    // Hash user + test → 0-1 range → Assign to A or B
    const hash = hashString(userId + testId);
    return hash < 0.5 ? 'A' : 'B';
  }
}
```

```

// Key property: Same user always gets same group
// But randomized assignment across user population
// Prevents systematic bias

function hashString(s) {
  let hash = 0;
  for (let i = 0; i < s.length; i++) {
    hash = ((hash << 5) - hash) + s.charCodeAt(i);
    hash |= 0;
  }
  return (Math.abs(hash) % 1000) / 1000;
}

return {
  user123: getUserGroup('user123', 'test1'), // Consistent
  user456: getUserGroup('user456', 'test1'), // Different
  user123Again: getUserGroup('user123', 'test1') // Same again
};
}

```

#### 4. Bloom Filters: Randomized Membership

```

function bloomFilter() {
  // Space-efficient approximate set membership
  // False positives: maybe, False negatives: definitely not

  class BloomFilter {
    constructor(size = 1000, hashFunctions = 3) {
      this.bits = new Uint8Array(size);
      this.size = size;
      this.k = hashFunctions;
    }

    _hash(item, seed) {
      // Simple hash (in practice: use cryptographic hash)
      let h = seed;
      for (let i = 0; i < item.length; i++) {
        h = ((h << 5) - h) + item.charCodeAt(i);
      }
      return Math.abs(h) % this.size;
    }

    add(item) {
      for (let i = 0; i < this.k; i++) {
        const index = this._hash(item, i);
        this.bits[index] = 1;
      }
    }

    contains(item) {

```

```

    for (let i = 0; i < this.k; i++) {
        const index = this._hash(item, i);
        if (!this.bits[index]) return false; // Definitely not there
    }
    return true; // Maybe there (false positive possible)
}
}

const bf = new BloomFilter(1000, 3);

bf.add('apple');
bf.add('banana');

console.log(bf.contains('apple')); // true
console.log(bf.contains('apple')); // true (always yes)
console.log(bf.contains('cherry')); // false (might be false positive)

// Space: O(n) bits vs O(n) bytes for hash set
// Use: Check membership before expensive operation (DB query)
}

```

## Common Misconceptions

### ✗ "Randomization means unpredictable results"

**No:** Randomized algorithms have bounded randomness

```

// Las Vegas: Always correct, randomized time
// You WILL get right answer
const sorted = quicksortRandomized([5, 2, 8, 1]); // Always sorted

// Monte Carlo: Randomized correctness, but bounded
const probably_prime = millerRabin(1000000007, 40);
// 99.99999999% sure it's prime (2^-40 error)

```

### ✗ "Randomization is inefficient"

**Often faster** than deterministic alternatives:

```

// Randomized Quicksort: O(n log n) expected
// vs Guaranteed O(n log n) mergesort

// In practice: Quicksort faster despite same complexity
// Why: Cache locality, fewer data moves, better constants

```

### ✗ "We should always use deterministic algorithms"

**Deterministic can be vulnerable:**

```
// Deterministic algorithm with worst-case  $O(n^2)$ :  
// Works fine until adversary sends crafted input  
  
// Randomized algorithm with  $O(n \log n)$  expected:  
// Adversary can't do anything (chooses random, not adversary)
```

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## Practical Mini-Exercises

- **Exercise 1: Shuffle** (Click to expand)
  - **Exercise 2: Reservoir Sample** (Click to expand)
- 

## Summary Cheat Sheet

```
// RANDOMIZED VS DETERMINISTIC  
Randomized: Breaks patterns, no worst-case inputs, constant time  
Deterministic: Predictable, can be slow on adversarial inputs  
  
// ALGORITHM TYPES  
Las Vegas: Always correct, randomized time (e.g., Quicksort)  
Monte Carlo: Fixed time, probably correct (e.g., Primality testing)  
  
// USE CASES  
Hash tables: Randomized hash function seed  
Load balancing: Random server selection + jitter  
Sorting: Randomized pivot selection  
Sampling: Reservoir sampling  
Shuffling: Fisher-Yates  
Membership: Bloom filters  
Retries: Exponential backoff with jitter
```

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## Next Steps

- ✓ **You've completed:** Randomized algorithms
- ➔ **Final topic:** [18. Risk, Uncertainty & When Math Breaks](#) - Models vs reality, assumptions, sensitivity analysis

**Before moving on:**

```
// Challenge: Implement randomized binary search (pick random pivot)  
function randomBinarySearch(arr, target) {  
  // Your solution  
}
```