

# Q100: Power of Two Integers

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## 1. Problem Understanding

- You are given a positive integer  $N$  ( $0 \leq N \leq 10^9$ ).
  - You must determine if there exist two integers  $A > 0$  and  $P > 1$  such that:
  - $A^P = N$
  - If yes print 1,
  - else print 0.
- 

## 2. Constraints

- $0 \leq N \leq 10^9$
  - $A > 0$
  - $P > 1$
  - $A$  and  $P$  are integers
  - Must fit in 32-bit signed integer
- 

## 3. Edge Cases

- $N = 0$  cannot be expressed
  - $N = 1$   $1^P = 1$  for any  $P > 1$
  - Perfect powers like 4 ( $=2^2$ ), 8 ( $=2^3$ ), 9 ( $=3^2$ ), 16 ( $=4^2$ )
  - Non-perfect powers like 10, 12, 20
- 

## 4. Examples

Input:

64

Output:

1

Explanation:

$4^3=64$  and also  $8^2=64$

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## 5. Approaches

## Approach 1: Brute Force (Iterative)

### Idea:

- Try every possible base A from 2 to  $\sqrt[n]{N}$ .
- For each A, multiply repeatedly ( $A * A$ ,  $A * A * A$ , etc.)
- until you exceed or match N.

### Java Code:

```
public static int isPower(int n) {  
    if (n == 0) return 0;  
    if (n == 1) return 1;  
  
    for (int a = 2; a * a <= n; a++) {  
        long p = a; // use long to avoid overflow  
        while (p <= n) {  
            p *= a;  
            if (p == n) return 1;  
        }  
    }  
    return 0;  
}
```

### Complexity (Time & Space):

- $\hat{O}(\sqrt[n]{N})$  Time Complexity
  - Outer loop:  $\sqrt[n]{N}$
  - Inner loop:  $\log_{\hat{O}(\sqrt[n]{N})}(N)$
  - Overall:  $O(\sqrt[n]{N} \cdot \log N)$
- $\frac{1}{4}$  Space Complexity
  - $O(1)$

## Approach 2: Using Logarithms (Mathematical)

### Idea:

- We know:
- $N = A^P \Rightarrow P = \log(A)/\log(N)$
- If P is an integer, then N can be expressed as  $A^P$ .
- Use math and rounding to check if P is close to an integer.

### Java Code:

```
public static int isPowerLog(int n) {  
    if (n == 0) return 0;  
    if (n == 1) return 1;  
  
    for (int a = 2; a * a <= n; a++) {
```

```

        double p = Math.log(n) / Math.log(a);
        if (Math.abs(p - Math.round(p)) < 1e-10 && Math.round(p) > 1)
            return 1;
    }
    return 0;
}

```

### Complexity (Time & Space):

- $\Theta(\log N)$  Time Complexity
- $O(\log N)$
- $\Theta(1)$  Space Complexity
- $O(1)$
- Note
- Floating-point precision can cause issues for large N.
- To reduce errors, use Math.round() with tolerance (1e-10).

### Approach 3: Recursive Power Check

#### Idea:

- Recursively try multiplying the base until the power exceeds N.
- For each base A, define:
  - check(A, current, N):
    - if current == N â†’ true
    - if current > N â†’ false
    - else â†’ check(A, current \* A, N)

#### Java Code:

```

public static boolean check(long base, long current, long n) {
    if (current == n) return true;
    if (current > n) return false;
    return check(base, current * base, n);
}

public static int isPowerRecursive(int n) {
    if (n == 0) return 0;
    if (n == 1) return 1;

    for (int a = 2; a * a <= n; a++) {
        if (check(a, a, n)) return 1;
    }
    return 0;
}

```

### Complexity (Time & Space):

- $\Theta(\log N)$  Time Complexity

- Similar to iterative:  $O(\sqrt{N} \sim \log N)$
- $\frac{3}{4}$  Space Complexity
  - $O(\log N)$  (due to recursion depth)

#### Approach 4: Precompute Powers (Optimization)

##### Idea:

- Instead of computing on the fly,
- store all possible powers of  $2 \leq a \leq 10^9$  that are  $a \% 10^9$  in a HashSet.
- Then simply check if  $N$  exists in the set.

##### Java Code:

```
public static int isPowerPrecompute(int n) {
    if (n == 0) return 0;
    if (n == 1) return 1;

    HashSet<Long> powers = new HashSet<>();
    for (int a = 2; a <= 31622; a++) { // 31622^2 ≈ 10^9
        long val = a * a;
        while (val <= 1_000_000_000L) {
            powers.add(val);
            val *= a;
        }
    }
    return powers.contains((long)n) ? 1 : 0;
}
```

##### Complexity (Time & Space):

- $\pm$  Time Complexity
  - Precomputation:  $O(\sqrt{N} \sim \log N)$
  - Query:  $O(1)$
- $\frac{3}{4}$  Space Complexity
  - $O(M)$  where  $M$  = number of precomputed powers (~few thousand)

#### Approach 5 – Optimized Root-Based Method (Best One)

**Idea:** -For a given exponent  $p$ , compute the integer base  $a = \text{round}(n^{(1/p)})$ . Then verify if  $a^p == n$ .

- We only need to check exponents  $p$  from 2 to 31 (since  $2^{31} > 2,147,483,647$ ).

##### Why Approach 5 Works (Reason):

- Instead of iterating all possible bases (which is huge for large  $n$ ), we iterate over small exponents ( $2 \leq p \leq 31$ ) – just 30 iterations total.
- Computing  $n^{(1/p)}$  gives a direct estimate of the base, avoiding billions of multiplications.
- Using long and rounding ensures accuracy and no overflow.

- For max input 2147483647, it completes instantly. **Steps:**
- Handle base cases: if  $n == 0$  return 0, if  $n == 1$  return 1.
- Loop  $p$  from 2 to 31.
- Compute approximate  $a = n^{(1/p)}$  using `Math.pow`.
- Round to nearest integer, then multiply  $a$   $p$  times (using `long` to avoid overflow).
- If the result equals  $n$ , return 1.
- After loop ends, return 0.

#### Java Code:

```
public static int isPowerOptimized(int n) {
    if (n == 0) return 0;
    if (n == 1) return 1;

    for (int p = 2; p <= 31; p++) {
        double a = Math.pow(n, 1.0 / p);
        int base = (int) Math.round(a);

        long val = 1;
        for (int i = 0; i < p; i++) {
            val *= base;
            if (val > n) break;
        }

        if (val == n) return 1;
    }
    return 0;
}
```

#### Complexity (Time & Space):

- Time Complexity
  - $O(31 \cdot \log N)$  practically  $O(1)$
- Space Complexity
  - $O(1)$

---

## 6. Justification / Proof of Optimality

- Each approach ensures both  $A$  and  $P$  are integers.
  - Brute-force and recursive approaches rely purely on integer arithmetic (no rounding errors).
  - Logarithmic approach offers speed, but may have floating-point inaccuracies.
  - Precomputation is best suited for multiple queries, trading off space for speed.
- 

## 7. Variants / Follow-Ups

- Find All  $(A, P)$  pairs

- Instead of returning 1, store or print all valid pairs where  $A^P = N$ .
- Find largest P for a given N
- Find the highest exponent such that  $A^P = N$ .
- Check Power of Base K
- Restrict A to a given base K (e.g., only check if N is a power of 2).
- For multiple queries
- Use precomputation + HashSet lookup for  $O(1)$  query per number.

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## 8. Tips & Observations

- Avoid using `Math.pow()` in integer problems (floating-point rounding errors).
  - Always use long while multiplying to avoid integer overflow.
  - For recursion, stop early when product exceeds N – this prunes large search space.
  - Logarithmic method is fastest for single queries but less reliable for precision-critical code.
  - Precomputation is perfect for repeated inputs in large datasets.
  - A need only go up to  $\sqrt{N}$  because  $A^2$  must be  $\leq N$  for  $P \geq 2$ .
- 

# Q12: Count All Digits of a Number

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## 1. Input, Output, & Constraints

- **Input:**

234

- **Output:**

3

**Constraints:**

- $0 \leq n \leq 5000$
  - n has no leading zeros except if  $n = 0$
- 

## 2. Approaches

Approach 1: Using Division (Iterative)

- **Idea:**
  - Divide n by 10 repeatedly, counting how many times until n becomes 0.

#### Java Code:

```
public static int countDigits(int n) {
    if (n == 0) return 1; // Edge case

    int count = 0;
    while (n > 0) {
        n /= 10;
        count++;
    }
    return count;
}
```

#### Complexity:

- Time:  $O(\log n)$  â€˜ number of digits
- Space:  $O(1)$

#### Approach 2: Using String Conversion

- **Idea:**
  - Convert the integer to a string and count the number of characters.

#### Java Code:

```
public static int countDigits(int n) {
    return String.valueOf(n).length();
}
```

#### Complexity:

- Time:  $O(\log n)$  â€˜ traverses digits to convert to string
- Space:  $O(\log n)$  â€˜ stores string representation

#### Approach 3: Using Logarithm (Math.log10)

- **Idea:**
  - The number of digits in a positive integer n is  $\text{floor}(\log_{10}(n)) + 1$ .
  - Edge case: if  $n = 0$ , the number of digits is 1.

#### Java Code:

```
public static int countDigits(int n) {
    if (n == 0) return 1; // Edge case
```

```
    return (int)(Math.log10(n)) + 1;  
}
```

#### Complexity:

- Time:  $O(1)$  â€ˆ single mathematical operation
  - Space:  $O(1)$
- 

### 3. Justification / Proof of Optimality

- Division â€ˆ simple and memory efficient
  - String â€ˆ concise and intuitive
  - Logarithm â€ˆ fastest for large numbers
- 

### 4. Variants / Follow-Ups

- Count digits in negative numbers
- Count digits in very large numbers (BigInteger in Java)
- Count digits in binary, octal, or hexadecimal representation

## Q13: Check for Perfect Number

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### 1. Understand the Problem

- **Paraphrase:** Proper divisors = all positive divisors excluding the number itself. A perfect number is equal to the sum of its proper divisors.
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### 2. Input, Output, & Constraints

- **Input:**

n

- **Output:**

Boolean true or false

#### Constraints:



- 1 ≤ n ≤ 5000
- 

### 3. Approaches

#### Approach 1: Iterating Over Divisors

- **Idea:**
  - Sum all divisors from 1 to  $n/2$  (proper divisors)
  - If sum equals  $n$ , return true; else return false

#### Java Code:

```
public static boolean isPerfectNumber(int n) {
    if (n == 1) return false; // 1 is not a perfect number

    int sum = 0;
    for (int i = 1; i <= n / 2; i++) {
        if (n % i == 0) {
            sum += i;
        }
    }
    return sum == n;
}
```

#### Complexity:

- Time:  $O(n)$  â€˜ iterate up to  $n/2$
- Space:  $O(1)$

#### Approach 2: Iterating up to $\sqrt{n}$ (Optimized)

- **Idea:**
  - Proper divisors come in pairs  $(i, n/i)$
  - Iterate  $i$  from 1 to  $\sqrt{n}$  and add both divisors to sum
  - Exclude  $n$  itself from the sum

#### Java Code:

```
public static boolean isPerfectNumber(int n) {
    if (n == 1) return false; // Edge case

    int sum = 1; // 1 is always a proper divisor
    for (int i = 2; i * i <= n; i++) {
        if (n % i == 0) {
            sum += i;
            int pair = n / i;
            if (pair != i) sum += pair; // Avoid adding sqrt twice
        }
    }
}
```

```
    return sum == n;
}
```

#### Complexity:

- Time:  $O(\sqrt{n})$  faster for larger numbers
- Space:  $O(1)$

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## 4. Justification / Proof of Optimality

- Approach 1 is simple and easy to implement.
- Approach 2 is more efficient, especially for larger  $n$ , as it avoids unnecessary iterations.

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## 5. Variants / Follow-Ups

- Check for abundant numbers (sum of divisors  $> n$ ) or deficient numbers (sum  $< n$ )
- Find all perfect numbers up to a given limit
- Handle very large numbers using optimized divisor sum formulas

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# Q14: GCD/HCF of Two Numbers

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## 1. Understand the Problem

- **Paraphrase:** Find the highest number that both  $n_1$  and  $n_2$  are divisible by.

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## 2. Input, Output, & Constraints

- **Input:**

```
4, 6
```

- **Output:**

```
Output: 2
```

```
Divisors of 4: 1, 2, 4
```

```
Divisors of 6: 1, 2, 3, 6
```

GCD = 2

#### Constraints:

- $1 \leq n1, n2 \leq 1000$
- 

## 3. Approaches

### Approach 1: Using Brute Force

- **Idea:**
  - Iterate from  $\min(n1, n2)$  down to 1
  - First number that divides both is the GCD

#### Java Code:

```
public static int gcdBruteForce(int n1, int n2) {
    int min = Math.min(n1, n2);
    for (int i = min; i >= 1; i--) {
        if (n1 % i == 0 && n2 % i == 0) {
            return i;
        }
    }
    return 1; // This line is never really reached because 1 always divides
}
```

#### Complexity:

- Time:  $O(\min(n1, n2))$
- Space:  $O(1)$

### Approach 2: Using Euclidean Algorithm (Optimized)

- **Idea:**
  - $\text{GCD}(a, b) = \text{GCD}(b, a \% b)$
  - Repeat until  $b = 0$ , then  $\text{GCD} = a$

#### Java Code:

```
public static int gcdEuclidean(int n1, int n2) {
    while (n2 != 0) {
        int temp = n2;
        n2 = n1 % n2;
        n1 = temp;
    }
    return n1;
}
```

### Complexity:

- Time:  $O(\log(\min(n_1, n_2)))$  â€ˆ very efficient
- Space:  $O(1)$

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## 4. Justification / Proof of Optimality

- Brute force is simple but inefficient for large numbers.
- Euclidean algorithm is optimal, widely used, and handles large inputs efficiently.

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## 5. Variants / Follow-Ups

- Find LCM using GCD:  $\text{LCM}(a, b) = (a * b) / \text{GCD}(a, b)$
- Extend to more than two numbers
- Find GCD of an array using pairwise GCD

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# Q15: LCM of Two Numbers

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## 1. Understand the Problem

- **Paraphrase:** Find the least number that both  $n_1$  and  $n_2$  divide evenly. Can be computed efficiently using GCD:  $\text{LCM}(a, b) = (a * b) / \text{GCD}(a, b)$

---

## 2. Input, Output, & Constraints

- **Input:**

4, 6

- **Output:**

12

Multiples of 4: 4, 8, 12, ...

Multiples of 6: 6, 12, 18, ...

LCM = 12

---

## 3. Approaches

### Approach 1: Using Formula $LCM = (n1 * n2) / GCD$

- **Idea:**
  - Compute GCD first using Euclidean algorithm
  - Then  $LCM = (n1 * n2) / GCD(n1, n2)$

#### Java Code:

```
public static int lcm(int n1, int n2) {
    int a = n1, b = n2;
    while (b != 0) {
        int temp = b;
        b = a % b;
        a = temp;
    }
    int gcd = a;
    return (n1 * n2) / gcd;
}
```

#### Complexity:

- Time:  $O(\log(\min(n1, n2)))$  for computing GCD
- Space:  $O(1)$

### Approach 2: Brute Force Multiples (Less Efficient)

- **Idea:**
  - Start from  $\max(n1, n2)$  and check each number incrementally until divisible by both

#### Java Code:

```
public static int lcmBruteForce(int n1, int n2) {
    int lcm = Math.max(n1, n2);
    while (true) {
        if (lcm % n1 == 0 && lcm % n2 == 0) return lcm;
        lcm++;
    }
}
```

#### Complexity:

- Time:  $O(n1 * n2)$  inefficient for large numbers
  - Space:  $O(1)$
-

## 4. Justification / Proof of Optimality

- Formula using GCD is efficient and widely used.
  - Brute force is simple but slow for larger numbers.
- 

## 5. Variants / Follow-Ups

- LCM of more than two numbers (compute pairwise LCM)
- LCM using prime factorization
- LCM of large numbers using BigInteger

# Q4: Optimus Prime â€™ Print All Primes up to N

---

## 1. Understand the Problem

- **Read & Identify:** Given an integer  $n$ , print all prime numbers between 1 and  $n$  (inclusive).
  - **Goal:** Find all primes  $\leq n$ .
  - **Paraphrase:** Numbers greater than 1 that have no divisors other than 1 and themselves.
- 

## 2. Input, Output, & Constraints

- **Input:**

8

- **Output:**

2 3 5 7

### Constraints:

- $1 \leq n \leq 100,000$
- Output size  $\leq n$
- Target time complexity:  $O(n \log \log n)$  with sieve

## 3. Approaches

### Approach 1: Naive Check (Trial Division)

- **Idea:**

- For every number from 2 to  $n$ , check if it's prime by trying to divide by numbers up to  $\sqrt{\text{num}}$ .

### Pseudocode:

```
function printPrimesNaive(n):
  for i from 2 to n:
    isPrime = true
    for j from 2 to sqrt(i):
      if i % j == 0:
        isPrime = false
        break
    if isPrime:
      print i
```

### Complexity:

- Time:  $O(n^{\frac{1}{2}}n)$
- Space:  $O(1)$

### Approach 2: Sieve of Eratosthenes (Optimized)

- **Idea:**
  - Assume all numbers 2..n are prime.
  - Start from 2, mark all its multiples as non-prime.
  - Repeat for next unmarked number.
  - Remaining unmarked numbers are primes.

### Pseudocode:

```
function sieve(n):
  isPrime = array[0..n] filled with true
  isPrime[0] = false, isPrime[1] = false

  for i from 2 to sqrt(n):
    if isPrime[i]:
      for j from i*i to n step i:
        isPrime[j] = false

  for i from 2 to n:
    if isPrime[i]:
      print i
```

### Complexity:

- Time:  $O(n \log \log n)$
- Space:  $O(n)$

## 4. Justification / Proof of Optimality

- Naive method is too slow for  $n$  up to  $10^6$ .
- Sieve of Eratosthenes runs in  $O(n \log \log n)$  which is optimal for this range.
- The sieve guarantees correctness by systematically eliminating composites.

## 5. Variants / Follow-Ups

- Print primes between  $L$  and  $R$  (Segmented Sieve).
- Count primes up to  $n$  (Prime Counting Function).
- Find the  $k$ -th prime  $\leq n$ .
- Applications in number theory (Goldbach's conjecture, twin primes).

# Q5: Calculate $nCr$

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## 1. Understand the Problem

- **Read & Identify:** Given two integers,  $n$  (total items) and  $r$  (items to choose), the goal is to calculate the binomial coefficient,  $nCr$ , which represents the number of distinct combinations.
  - **Goal:** Compute the value of  $nCr$  using the standard combinatorial formula.
  - **Paraphrase:** Find the number of distinct subsets of size  $r$  from a set of size  $n$ .
- 

## 2. Input, Output, & Constraints

- **Input:**

Two non-negative integers,  $n$  and  $r$ .

- **Output:**

A single integer representing the calculated value of  $nCr$ .

### Constraints:

- $1 \leq n \leq 20$
- $1 \leq r \leq n$
- The result will fit within a standard 64-bit integer ( $\leq 10^{19}$ ).

## 3. Approaches

### Approach 1: Direct Factorial Calculation (Naive)

- **Idea:**



- Directly compute the factorials for  $n$ ,  $r$ , and  $(n-r)!$ , then perform the division.  $nCr = n! / (r! \cdot (n-r)!)$

#### Pseudocode:

```
function Calculate_nCr(n, r):
    // Requires a data type that can hold 20! (long long/64-bit integer)
    numerator = Factorial(n)
    denominator = Factorial(r) * Factorial(n - r)
    return numerator / denominator
```

#### Complexity:

- Time:  $O(n)$
- Space:  $O(1)$  space.

#### Approach 2: Optimized Multiplicative Formula (Preferred)

- **Idea:**
  - Simplify the fraction before calculation to minimize the size of intermediate numbers, which is crucial for larger  $n$ . The simplified form cancels out the largest factorial term,  $(n-r)!$ .

#### Pseudocode:

```
function Calculate_nCr(n, r):
    // Use nCr = nC(n-r) property for fewer iterations
    if r > n / 2:
        r = n - r

    result = 1
    // Loop r times
    for i from 1 to r:
        // Current calculation: result * (n-i+1) / i
        result = result * (n - i + 1)
        result = result / i // Division is guaranteed to be exact
    return result
```

#### Complexity:

- Time:  $O(\min(r, n-r))$ , which is  $O(n)$
- Space:  $O(1)$  space

## 4. Justification / Proof of Optimality

- Approach 2 is the better solution. While both approaches are  $O(n)$  time complexity, Approach 2 keeps the intermediate values much smaller, which minimizes the risk of overflow. It directly computes  $nCr$  step-by-step, ensuring each partial product is a valid integer combination value, which is inherently

safer than computing three separate, massive factorials (as in Approach 1) and hoping their ratio fits the integer type.

## 5. Variants / Follow-Ups

- Large Constraints ( $n, r \leq 10^9$ ): When  $n$  and  $r$  are very large and the answer must be computed modulo  $p$ . This requires number theory techniques like Lucas Theorem or calculating factorials and their modular inverses using Fermat's Little Theorem.
- Dynamic Programming (Pascal's Identity): For scenarios where many  $nCr$  values are needed, the relation  $nCr = (n-1)Cr + (n-1)C(r-1)$  allows filling a table (Pascal's Triangle) in  $O(n^2)$  time.
- Permutations ( $nPr$ ): The problem of calculating  $nPr = n! / (n-r)!$  involves a similar multiplicative approach, simply stopping before dividing by  $r!$ .

# Q6: Binary To Decimal Conversion

---

## 1. Input, Output, & Constraints

- **Input:**

1011

- **Output:**

11

### Constraints:

- Binary number contains only 0 and 1.
- Length of binary number  $\leq 32$  (or as per system integer limit).

## 2. Approaches

### Approach 1: Positional Value (Iterative)

- **Idea:**
  - Each binary digit represents a power of 2. Starting from the least significant bit (LSB), multiply each bit by  $2^{\text{position}}$  and sum them.

### Pseudocode:

```
function binaryToDecimal(binary):
    decimal = 0
    length = len(binary)
    for i in range(0, length):
```

```
        bit = int(binary[length - 1 - i])
        decimal += bit * (2i)
    return decimal
```

**Complexity:**

- Time:  $O(n)$ ,  $n$  = number of bits
- Space:  $O(1)$

**Approach 2: Left-to-Right Multiplication (Accumulation)**

- **Idea:**
  - Traverse the binary number from left to right, multiply accumulated result by 2, then add current bit.
  - Works for very large binary numbers.
  - Slightly more efficient than computing powers explicitly.

**Pseudocode:**

```
function binaryToDecimal(binary):
    decimal = 0
    for bit in binary:
        decimal = decimal * 2 + int(bit)
    return decimal
```

**Complexity:**

- Time:  $O(n)$
- Space:  $O(1)$

### 3. Justification / Proof of Optimality

- Optimality: Approach 2 is optimal in terms of simplicity and efficiency.
- Comparison:
- Approach 1: Direct calculation using powers  $\hat{=}$  more verbose.
- Approach 2: Accumulative method  $\hat{=}$  elegant, in-place.

### 4. Variants / Follow-Ups

- Decimal  $\hat{=}$  Binary conversion
- Binary  $\hat{=}$  Hexadecimal conversion
- Large binary strings beyond integer limit  $\hat{=}$  Use BigInt or string manipulation
- Summing multiple binary numbers efficiently

## Q7: Decimal to Binary Conversion

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# 1. Input, Output, & Constraints

- **Input:**

11

- **Output:**

1011

**Constraints:**

- Decimal number  $\neq 0$
- Decimal number  $\leq$  maximum integer limit of language

## 2. Approaches

### Approach 1: Repeated Division by 2

- **Idea:**
  - Keep dividing the decimal number by 2. The remainder at each step forms the binary digits from least significant bit (LSB) to most significant bit (MSB).

**Pseudocode:**

```
function decimalToBinary(n):  
    if n == 0:  
        return "0"  
    binary = ""  
    while n > 0:  
        remainder = n % 2  
        binary = str(remainder) + binary  
        n = n // 2  
    return binary
```

**Complexity:**

- Time:  $O(\log n)$
- Space:  $O(\log n)$  (for storing binary digits)

### Approach 2: Using Bit Manipulation

- **Idea:**
  - Extract bits from decimal number using bitwise AND and right shift operations.

**Pseudocode:**

---

```

function decimalToBinary(n):
    if n == 0:
        return "0"
    binary = ""
    while n > 0:
        bit = n & 1
        binary = str(bit) + binary
        n = n >> 1
    return binary

```

#### Complexity:

- Time:  $O(\log n)$
- Space:  $O(\log n)$

### 3. Justification / Proof of Optimality

- Optimality: Approach 1 & 2 are optimal and educational.
- Comparison:
- Approach 1: Classic method using division â€ easy to understand.
- Approach 2: Bitwise method â€ faster in low-level operations.

### 4. Variants / Follow-Ups

- Binary â€ Decimal conversion
- Decimal â€ Hexadecimal conversion
- Decimal â€ Binary for negative numbers (2â€™s complement)
- Fast conversion using recursion or stack

## Q8: Print Continuous Character Pattern

---

### 1. Input, Output, & Constraints

- **Input:**

5

- **Output:**

```

A
BC
CDE
DEFG
EFGHI

```

## 2. Approaches

### Approach 1: Using ASCII Values

- **Idea:**
  - Use ASCII values of characters. Start from 'A' (ASCII 65), and for each row, print consecutive letters using  $(\text{ASCII value}) \% 26 + 65$  to handle cyclic behavior.

#### Pseudocode:

```
function printPattern(n):
    for row in range(1, n+1):
        start_char = 65 + (row - 1)    # 'A' = 65
        for col in range(row):
            char_to_print = chr(65 + ((start_char - 65 + col) % 26))
            print(char_to_print, end="")
        print()    # New line after each row
```

#### Complexity:

- Time:  $O(n^2)$  â†’ Each row has up to  $n$  letters
- Space:  $O(1)$  â†’ Only loop variables

### Approach 2: Using String Arithmetic (Optional)

- **Idea:**
  - Pre-generate the alphabet string "ABCDEFGHIJKLMNOPQRSTUVWXYZ" and use slicing with modulo to handle cyclic letters.

#### Pseudocode:

```
alphabet = "ABCDEFGHIJKLMNOPQRSTUVWXYZ"
function printPattern(n):
    for row in range(1, n+1):
        start_index = row - 1
        for col in range(row):
            index = (start_index + col) % 26
            print(alphabet[index], end="")
        print()
```

#### Complexity:

- Time:  $O(n^2)$
- Space:  $O(1)$

## 3. Justification / Proof of Optimality

- Optimality: Both approaches are efficient; Approach 1 is straightforward using ASCII, Approach 2 is more intuitive for beginners.
- Comparison:
- ASCII arithmetic â€ˆ Less memory, direct computation
- String-based â€ˆ Easier to read and maintain, especially for cyclic operations

## 4. Variants / Follow-Ups

- Change starting letter for the first row (instead of always 'A')
- Print pattern in reverse order
- Allow lowercase letters or custom alphabet sets
- Print continuous character diamond pattern

# Q99: Sum to N

---

## 1. Problem Understanding

- You need to find how many combinations of distinct digits (1â€”9) of size k have a sum equal to n.
- Each combination must use distinct numbers, and the order doesnâ€™t matter (i.e., {1,2,4} and {2,1,4} are the same).

---

## 2. Constraints

- $1 \leq k \leq 9$
- $1 \leq n \leq 45$
- Digits available = {1, 2, 3, 4, 5, 6, 7, 8, 9}

---

## 3. Edge Cases

- $n < \text{smallest possible sum}$  â€ˆ return 0
- $n > \text{largest possible sum}$  â€ˆ return 0
- If no valid combinations exist, return 0

---

## 4. Examples

```
Input:
9 3
Output:
3
Valid combinations:
{1, 2, 6}, {1, 3, 5}, {2, 3, 4}
```

---

## 5. Approaches

### Approach 1: Recursive Backtracking

#### Idea:

- Use recursion to explore all combinations of numbers from 1 to 9.
- At each step, choose a number, reduce n by that number, and decrease k by 1.
- Stop when n == 0 and k == 0 (a valid combination found).

#### Steps:

- Define a helper function countCombinations(start, n, k)
  - start: current number to consider (ensures distinct + ascending order)
- If n == 0 and k == 0 → found a valid combination → return 1
- If n < 0 or k == 0 → invalid path → return 0
- Loop i from start to 9
  - Include i → call recursively with n - i, k - 1, i + 1
- Sum all valid recursive counts

#### Java Code:

```
public class Main {
    public static int sumOfN(int n, int k) {
        return helper(1, n, k);
    }

    private static int helper(int start, int n, int k) {
        // Base case: found valid combination
        if (n == 0 && k == 0) return 1;

        // Base case: invalid state
        if (n < 0 || k == 0) return 0;

        int count = 0;
        for (int i = start; i <= 9; i++) {
            count += helper(i + 1, n - i, k - 1);
        }
        return count;
    }

    public static void main(String[] args) {
        java.util.Scanner sc = new java.util.Scanner(System.in);
        int n = sc.nextInt();
        int k = sc.nextInt();
        System.out.println(sumOfN(n, k));
    }
}
```

Recursion Tree (Example: n = 7, k = 3)

```
helper(1,7,3)
  ↳ i=1 → helper(2,6,2)
```



```

    if i == 2: helper(3, 4, 1)
    if i == 3: helper(4, 1, 0)
    if i == 4: helper(5, 0, 0)
    if i > 4: exceeds 9
    if i > 2: other branches
    if i > 1: other branches

```

Found {1, 2, 4} total = 1

### Complexity (Time & Space):

- Time Complexity
  - Roughly  $O(2^n)$  since we explore include/exclude for 9 digits
  - Practically much less due to pruning ( $n < 0, k == 0$ )
- Space Complexity
  - $O(k)$  recursion depth (stack space)

## 6. Justification / Proof of Optimality

- This approach ensures:
- Distinct numbers (due to start parameter)
- No duplicates (combination order ignored)
- Checks all possible valid subsets efficiently

## 7. Variants / Follow-Ups

- Find actual combinations (store in list instead of counting)
- Allow repeated numbers (remove distinctness constraint)
- Use digits from 1 to n instead of 1 to 9

## 8. Tips & Observations

- Use backtracking for combination-type problems
- Always control distinctness via the start parameter
- Recursion naturally handles combination depth (k here)