

# Comprehensive Study Guide: Hashing & Bloom Filters

Exam Review Notes

## 1 The Motivation for Hashing

To maintain a set  $S$  of keys drawn from a large universe  $U$ , we compare three fundamental data structures. The goal of hashing is to combine the speed of arrays with the space efficiency of linked lists.

Approach	Space	Time (Lookup)	Trade-off
Direct Addressing (Array)	$\Theta( U )$	$O(1)$ Worst Case	<b>Fast but Impossible Space.</b> If $U$ is large (e.g., IPv6), memory explodes.
Linked List	$\Theta( S )$	$\Theta( S )$ Worst Case	<b>Space Efficient but Slow.</b> Must scan the entire list to find items.
Hash Table	$\Theta( S )$	$O(1)$ Expected	<b>The Sweet Spot.</b> Maps huge $U$ to small table size $n$ .

Table 1: Comparison of Data Structures

## 2 Birthday Paradox

Consider  $n$  people with random birthdays (assuming 365 days in a year, equally likely, and independent). We want to find the smallest integer  $n$  such that the probability of at least two people sharing a birthday is at least 50%:

$$P(\text{at least one collision}) \geq 0.5 \quad (1)$$

### 2.1 The Complement Strategy

Calculating collisions directly is complex due to the number of possible pairs. It is easier to calculate the complement event: **all birthdays are unique**.

$$\begin{aligned} P(\text{shared}) &= 1 - P(\text{all unique}) \\ 1 - P(\text{unique}) &\geq 0.5 \\ P(\text{unique}) &\leq 0.5 \end{aligned}$$

### 2.2 Exact Probability Formula

We assign birthdays to  $n$  people one by one, ensuring no collisions:

- Person 1:  $365/365$  (Any day is fine)
- Person 2:  $364/365$  (Must avoid Person 1)
- Person  $k$ :  $(365 - (k - 1))/365$  (Must avoid previous  $k - 1$  people)

The total probability is the product of these independent choices:

$$P(\text{unique}) = 1 \times \frac{364}{365} \times \frac{363}{365} \times \cdots \times \frac{365 - (n - 1)}{365} \quad (2)$$

This can be written compactly using product notation:

$$P(\text{unique}) = \prod_{k=0}^{n-1} \left( \frac{365 - k}{365} \right) = \prod_{k=0}^{n-1} \left( 1 - \frac{k}{365} \right) \quad (3)$$

### 2.3 Approximation using Taylor Series

To solve for  $n$ , we convert the product into a sum by taking the natural logarithm ( $\ln$ ) of both sides:

$$\ln(P(\text{unique})) = \sum_{k=0}^{n-1} \ln \left( 1 - \frac{k}{365} \right) \quad (4)$$

We use the Taylor Series expansion for  $\ln(1 - x)$  near  $x = 0$ :

$$\ln(1 - x) = -x - \frac{x^2}{2} - \frac{x^3}{3} - \dots$$

For small  $x$  (where  $x = k/365$ ), we approximate  $\ln(1 - x) \approx -x$ .

Substituting this into our sum:

$$\begin{aligned} \ln(P(\text{unique})) &\approx \sum_{k=0}^{n-1} \left( -\frac{k}{365} \right) \\ &= -\frac{1}{365} \sum_{k=0}^{n-1} k \end{aligned}$$

Using the arithmetic series sum formula  $\sum_{k=0}^{n-1} k = \frac{n(n-1)}{2}$ , we get:

$$\ln(P(\text{unique})) \approx -\frac{1}{365} \cdot \frac{n(n-1)}{2} = -\frac{n(n-1)}{730} \quad (5)$$

Exponentiating both sides gives the approximate probability formula:

$$P(\text{unique}) \approx e^{-\frac{n(n-1)}{730}} \quad (6)$$

### 2.4 Solving for $n$

We require  $P(\text{unique}) \leq 0.5$ .

$$\begin{aligned} e^{-\frac{n(n-1)}{730}} &\leq 0.5 \\ -\frac{n(n-1)}{730} &\leq \ln(0.5) \\ -\frac{n(n-1)}{730} &\leq -0.693 \quad (\text{since } \ln(2) \approx 0.693) \\ n(n-1) &\geq 0.693 \times 730 \\ n(n-1) &\geq 505.89 \end{aligned}$$

Approximating  $n(n - 1) \approx n^2$ :

$$\begin{aligned} n^2 &\approx 506 \\ n &\approx \sqrt{506} \approx 22.49 \end{aligned}$$

Since  $n$  must be an integer, we round up:

$$n = 23 \quad (7)$$

### 3 Modular Arithmetic & Inverses

Understanding modular inverses is a prerequisite for the Universal Hashing proof.

#### 3.1 Modular Inverses

An integer  $u$  is said to be **invertible modulo  $n$**  if there exists an integer  $v$  (often written as  $u^{-1}$ ) such that:

$$u \cdot v \equiv 1 \pmod{n} \quad (8)$$

- **Existence Condition:** An inverse exists **if and only if**  $\gcd(u, n) = 1$ .
- **Prime Modulus Property:** If  $n = p$  (a prime number), then every non-zero integer  $u \in \{1, \dots, p - 1\}$  is invertible. This is because a prime number shares no factors with numbers smaller than itself.

#### 3.2 Computing Inverses: Extended Euclidean Algorithm (EEA)

To find the inverse of  $u \pmod{n}$ , we solve Bézout's Identity:  $s \cdot u + t \cdot n = 1$ . Reducing this modulo  $n$  gives  $s \cdot u \equiv 1 \pmod{n}$ .

**Example: Find  $11^{-1} \pmod{26}$**

**Step 1: Forward Pass (Euclidean Division)**

$$\begin{aligned} 26 &= 2 \cdot 11 + 4 \\ 11 &= 2 \cdot 4 + 3 \\ 4 &= 1 \cdot 3 + 1 \quad \leftarrow \text{GCD is 1, so inverse exists.} \end{aligned}$$

**Step 2: Backward Pass (Substitution)** Express 1 as a linear combination of 11 and 26.

$$\begin{aligned} 1 &= 4 - 1 \cdot 3 \\ &= 4 - 1 \cdot (11 - 2 \cdot 4) \quad (\text{Substitute } 3 = 11 - 2 \cdot 4) \\ &= 3 \cdot 4 - 1 \cdot 11 \\ &= 3 \cdot (26 - 2 \cdot 11) - 1 \cdot 11 \quad (\text{Substitute } 4 = 26 - 2 \cdot 11) \\ &= 3 \cdot 26 - 6 \cdot 11 - 1 \cdot 11 \\ &= 3 \cdot 26 - 7 \cdot 11 \end{aligned}$$

**Step 3: Result** Reducing modulo 26, the term  $3 \cdot 26$  becomes 0:

$$\begin{aligned} -7 \cdot 11 &\equiv 1 \pmod{26} \\ -7 &\equiv 19 \pmod{26} \end{aligned}$$

**Conclusion:**  $11^{-1} \equiv 19 \pmod{26}$ .

## 4 Universal Hashing

### 4.1 Definition

A family of hash functions  $\mathcal{H}$  mapping  $U \rightarrow \{0, \dots, n-1\}$  is **universal** if for any two distinct keys  $x \neq y$ :

$$\Pr_{h \in \mathcal{H}}[h(x) = h(y)] \leq \frac{1}{n} \quad (9)$$

*Note:* Per-key uniformity (probability  $1/n$  of hitting a specific bucket) is necessary but **not sufficient**. We specifically require the **pairwise collision probability** to be low.

### 4.2 Proof: Hashing IP Addresses

**The Setup:**

- **Keys:** IP addresses decomposed into 4 parts:  $x = (x_1, x_2, x_3, x_4)$  where  $x_i \in \{0, \dots, n-1\}$ .
- **Buckets:** A prime number  $n$ .
- **Hash Family:** Defined by a random vector  $a = (a_1, a_2, a_3, a_4)$  where each  $a_i$  is uniform in  $\{0, \dots, n-1\}$ .
- **Function:**  $h_a(x) = \sum_{i=1}^4 a_i x_i \pmod{n}$ .

**The Proof:** We wish to calculate the probability that distinct keys  $x$  and  $y$  collide.

#### 1. The Collision Equation

$$\begin{aligned} h_a(x) &= h_a(y) \\ \sum_{i=1}^4 a_i x_i &\equiv \sum_{i=1}^4 a_i y_i \pmod{n} \\ \sum_{i=1}^4 a_i (x_i - y_i) &\equiv 0 \pmod{n} \end{aligned}$$

**2. Isolate a Non-Zero Difference** Since  $x \neq y$ , they must differ in at least one position. Without loss of generality, assume  $x_4 \neq y_4$ . We isolate the term involving  $a_4$ :

$$\begin{aligned} a_4(x_4 - y_4) + \sum_{i=1}^3 a_i(x_i - y_i) &\equiv 0 \pmod{n} \\ a_4(x_4 - y_4) &\equiv -\sum_{i=1}^3 a_i(x_i - y_i) \pmod{n} \end{aligned}$$

Let the right-hand side be  $C = \sum_{i=1}^3 a_i(y_i - x_i)$ . The equation becomes:

$$a_4(x_4 - y_4) \equiv C \pmod{n} \quad (10)$$

**3. Principle of Deferred Decisions** Imagine we pick the random coefficients  $a_1, a_2, a_3$  **first**. This fixes the value of  $C$ . Now, we pick  $a_4$ .

Since  $n$  is prime and  $(x_4 - y_4) \not\equiv 0 \pmod{n}$ , the term  $(x_4 - y_4)$  has a unique modular inverse. We multiply both sides by this inverse:

$$a_4 \equiv C \cdot (x_4 - y_4)^{-1} \pmod{n} \quad (11)$$

**4. Conclusion** The equation above yields exactly **one** valid solution for  $a_4$ . Since  $a_4$  is chosen uniformly from  $n$  possibilities  $\{0, \dots, n-1\}$ , the probability of picking exactly this one solution is:

$$\Pr[\text{Collision}] = \frac{1}{n} \quad (12)$$

Thus, the family is Universal.

## 5 Bloom Filters

A Bloom Filter is a space-efficient probabilistic data structure for set membership. It yields no false negatives, but possible false positives.

### 5.1 Mathematical Derivation of False Positive Rate

**Parameters:**

- $n$ : Number of items inserted.
- $m$ : Size of the bit array.
- $k$ : Number of hash functions.

**Step 1: Probability a bit is 0** When inserting 1 element with 1 hash function, the probability a specific bit is *not* set (remains 0) is:

$$1 - \frac{1}{m}$$

After inserting  $n$  elements using  $k$  hash functions, we perform  $kn$  total writes. The probability the bit is still 0 is:

$$\left(1 - \frac{1}{m}\right)^{kn}$$

**Step 2: Taylor Series Approximation** Recall the Taylor series expansion for  $e^{-x}$  for small  $x$ :  $e^{-x} \approx 1 - x$ . Conversely, for large  $m$ :

$$1 - \frac{1}{m} \approx e^{-1/m}$$

Substituting this into our equation:

$$\Pr[\text{bit is 0}] \approx \left(e^{-1/m}\right)^{kn} = e^{-kn/m}$$

**Step 3: Probability a bit is 1**

$$\Pr[\text{bit is 1}] = 1 - \Pr[\text{bit is 0}] = 1 - e^{-kn/m}$$

**Step 4: False Positive Probability ( $\epsilon$ )** A false positive occurs when we query an item that is **not** in the set, but all  $k$  hash functions map to bits that happen to be 1. Assuming independence:

$$\epsilon = \left(1 - e^{-kn/m}\right)^k \quad (13)$$

## 6 Summary Checklist

- **Why Prime Buckets?** Prime  $n$  ensures that every non-zero difference  $(x_i - y_i)$  has a modular inverse. This guarantees the collision equation has a unique solution.
- **Invertible Slopes:** In affine hashing  $h(x) = (ax + b) \pmod{m}$ , if  $a$  is not coprime to  $m$  ( $\gcd(a, m) > 1$ ), multiple keys will collapse to the same bucket systematically. We need  $\gcd(a, m) = 1$ .
- **Universal Bound:** The gold standard is  $\Pr[\text{Collision}] \leq 1/n$ .