## Universal Hash Functions and Perfect Hashing



## Universal Hashing

- If a malicious adversary chooses the keys to be hashed by some fixed hash function, he can choose $n$ keys $x_{i}$ such that they all hash to the same value

$$
H\left(x_{i}\right)=h \text { for } i=1,2, \ldots, n
$$

- This implies that the hash table will have $\Theta(n)$ retrieval time
- Any fixed hash function would have this worst-case behavior
- The only effective way to improve the situation is to choose the hash function randomly in a way that is independent of the keys
- This approach is called universal hashing
- It can yield provably good performance in the average, not matter which keys the adversary chooses


## Universal Hashing

- In universal hashing, we select the hash function at random from a carefully designed class of hash functions
- Randomization guarantees that no single input will evoke worst-case behavior
- This selection is done at the beginning of each execution
- Therefore, the algorithm can behave differently on each execution, even for the same input
- This will guarantee good average-case performance
- Of course, poor performance will occur when the selected hash function hashes the keys poorly
- However, the probability of this situation is small, and is the same for any set of keys of the same size


## Universal Hash Functions

- Let $\mathcal{H}$ be a finite collection of hash functions that map a given universe $U$ of keys into the range $\{0,1,2, \ldots, m-1\}$
- The set $\mathcal{H}$ is said to be universal if for each pair of keys $x, y \in U$, the number of hash functions $h \in \mathcal{H}$ for which $h(x)=h(y)$ is at most $|\mathcal{H}| / m$
- In other words, with a hash function randomly chosen from $\mathcal{H}$, the chance of collision between $x$ and $y$ is $1 / m$
- This is the same chance of collision if $h(x)$ and $h(y)$ were randomly and independently chosen from the set $\{0,1,2, \ldots, m-1\}$


## Average Case Behavior

- Suppose a hash function $h$ is randomly chosen from a universal collection of hash functions
- It is used to hash $n$ keys into a table $T$ of size $m$, with chaining as the collision resolution method
- Let $\alpha$ be the load factor, defined as $\alpha=n / m$
- If $x$ is not in the table, the expected length of the list that the key $x$ hashes into is at most $\alpha$
- If $x$ is in the table, the expected length of the list that contains the key $x$ is at most $1+\alpha$


## Average Case Behavior

- Consider a pair of keys $x$ and $y$; due to the definition of the universal hashing, the probability that they collide is

$$
P[h(x)=h(y)] \leq \frac{1}{m}
$$

- Let the random variable $R_{x y}$ take the value of 1 when $h(x)=h(y)$ and 0 otherwise; the expected value of $R_{x y}$ is

$$
E\left[R_{x y}\right]=\frac{1}{m}
$$

- Let the random variable $S_{x}$ be the number of keys other than $x$ that hash to the same slot as $x$, given as

$$
S_{x}=\sum_{\substack{y \in T \\ y \neq x}} R_{x y}
$$

## Average Case Behavior

- Therefore, we have

$$
E\left[S_{x}\right]=E\left[\sum_{\substack{y \in T \\ y \neq x}} R_{x y}\right]=\sum_{\substack{y \in T \\ y \neq x}} E\left[R_{x y}\right] \leq \sum_{\substack{y \in T \\ y \neq x}} \frac{1}{m}
$$

- If $x \notin T$, then the list length is equal to $S_{x}$ and

$$
\mid\{y: y \in T \text { and } y \neq x\} \mid=n
$$

and thus the expected list length $E\left[S_{x}\right] \leq n / m=\alpha$

- If $x \in T$, then because $x$ appears in the list $T[h(x)]$ and the count does not include $x$, we have the list length as $S_{x}+1$ and

$$
\mid\{y: y \in T \text { and } y \neq x\} \mid=n-1
$$

and thus the expected list length

$$
E\left[S_{x}\right]+1 \leq(n-1) / m+1=1+\alpha-1 / m<1+\alpha
$$

## Average Case Behavior

## Theorem

Using universal hashing and collision resolution by chaining in an initially empty table with $m$ slots, it takes $\Theta(n)$ time to handle any sequence of $n$ Insert, Find, and Delete operations containing $O(m)$ Insert operations.

- The number of Insert operations is $O(m)$, thus we have $n=O(m)$ which implies $\alpha=O(1)$
- The Insert and Delete operations take constant time, and the expected time for Find operation is $O(1)$ since the expected length of the list is at most $\alpha$
- Therefore, the expected time for the entire sequence of $n$ operations is $O(n)$ since each operation takes $\Omega(1)$, the bound $\Theta(n)$ is obtained


## Designing Universal Hash Functions

- We will give 3 constructions and show them that they are universal
- The first construction is based on linear congruential arithmetic with two distinct moduli: $p$ and $m$, where $p$ is a prime
- The second construction uses a random 0-1 matrix and mod 2 arithmetic
- The third method is based the dot-product modulo $m$


## Construction of $\mathcal{H}_{p, m}$

- Select a prime $p$ that is large enough so that every possible key is in the range 0 to $p-1$
- Let $\mathcal{Z}_{p}=\{0,1,2, \ldots, p-1\}$ and $\mathcal{Z}_{p}^{*}=\{1,2, \ldots, p-1\}$
- The size of the universe of the keys is $p$ which is larger than the hash table size $m$, i.e., $p>m$
- Consider the integer $a \in \mathcal{Z}_{p}^{*}$ and $b \in \mathcal{Z}_{p}$
- Define the hash function family as

$$
h_{a, b}(x)=(a \cdot x+b \bmod p) \bmod m
$$

- The class of hash functions is defined as

$$
\mathcal{H}_{p, m}=\left\{h_{a, b} \mid a \in \mathcal{Z}_{p}^{*} \text { and } b \in \mathcal{Z}_{p}\right\}
$$

## Properties of $\mathcal{H}_{p, m}$

- An Example: $p=17$ and $m=6$, we have $h_{3,4}(8)=5$ since

$$
\begin{aligned}
h_{3,4}(8) & =((3 \cdot 8+4) \bmod 17) \bmod 6 \\
& =(28 \bmod 17) \bmod 6 \\
& =11 \bmod 6 \\
& =5
\end{aligned}
$$

- Each hash function $h_{a, b}$ maps $\mathcal{Z}_{p}$ to $\mathcal{Z}_{m}$ : the keys are in the range 0 to $p-1$, while the hash values are from 0 to $m-1$
- This family has the nice property that the table size $m$ is arbitrary, not necessarily a prime
- There are $p-1$ choices of $a$ and $p$ choices of $b$, and thus, there are $p(p-1)$ hash functions


## Proving Universality of $\mathcal{H}_{p, m}$

## Theorem

The class $\mathcal{H}_{p, m}$ of hash functions is universal.

- Consider two distinct keys $x$ and $y$ from $\mathcal{Z}_{p}$, so that $x \neq y$
- For a given hash function $h_{a, b}$, first compute

$$
\begin{aligned}
& r=(a \cdot x+b) \bmod p \\
& s=(a \cdot y+b) \bmod p
\end{aligned}
$$

- $r-s=a(x-y)$ is nonzero since $x \neq y$ and $a \neq 0$, and $p$ is prime
- Therefore, if $x \neq y$, we will always have $r \neq s$
- There will not be collision on the "mod $p$ level"


## Proving Universality of $\mathcal{H}_{p, m}$

- Moreover, each possible $p(p-1)$ pair of $(a, b)$ with $a \neq 0$ yields a different pair $(r, s)$ since

$$
\begin{aligned}
& a=(r-s)(x-y)^{-1} \bmod p \\
& b=(r-a x) \bmod p
\end{aligned}
$$

- There are $p(p-1)$ possible pairs $(r, s)$ with $r \neq s$, and thus, there is a one-to-one correspondence between pairs $(a, b)$ with $a \neq 0$ and pairs $(r, s)$ with $r \neq s$


## Proving Universality of $\mathcal{H}_{p, m}$

- Thus, for any given pairs of inputs $x$ and $y$, if we pick $(a, b)$ uniformly at random from $\mathcal{Z}_{p}^{*} \times \mathcal{Z}_{p}$, the resulting pair is equally likely to be any pair of distinct values modulo $p$
- The probability that distinct keys $x$ and $y$ collide is equal to the probability $r=s(\bmod m)$ when $r$ and $s$ are randomly chosen as distinct values modulo $p$
- Furthermore, the probability that $s$ collides with $r$ when reduced modulo $m$ is at most $1 / m$, and therefore

$$
P\left[h_{a, b}(x)=h_{a, b}(y)\right] \leq 1 / m
$$

so that $\mathcal{H}_{p, m}$ is universal

## Construction of the Matrix Method

- Assume that the keys are $u$ bits long: $x=\left(x_{u-1} \cdots x_{1} x_{0}\right)$
- The hash table size as a power of two, as $m=2^{b}$, and the hash values $z=h(x)$ are $b$-bit integers: $z=\left(z_{b-1} \cdots z_{1} z_{0}\right)$
- The hash function $h$ is computed using a 0-1 random matrix of dimension $b \times u$, denoted as $A$
- The hash operation $h(x)$ takes the key $x$ expressed as a $u$-bit binary number and multiplies with the matrix $A$ to obtain the $b$-bit hash
- All computations are done in mod 2: the Galois field GF(2)


## Properties of the Matrix Method

- An Example: Let $u=4$ and $b=3$, therefore, the keys are 4 -bit long $x=\left(x_{3} x_{2} x_{1} x_{0}\right)$ and the hash values are 3-bit long $z=\left(z_{2} z_{1} z_{0}\right)$
- The random 0-1 matrix is of size $b \times u=3 \times 4$
- Taking $A$ as below, the computation of $z=h(x)$ is performed using

$$
\left[\begin{array}{l}
z_{0} \\
z_{1} \\
z_{2}
\end{array}\right]=\left[\begin{array}{llll}
1 & 0 & 0 & 0 \\
0 & 1 & 1 & 1 \\
1 & 1 & 1 & 0
\end{array}\right]\left[\begin{array}{l}
x_{0} \\
x_{1} \\
x_{2} \\
x_{3}
\end{array}\right]
$$

- Let $x=\left(x_{3} x_{2} x_{1} x_{0}\right)=(0101)$, we obtain $\left(z_{2} z_{1} z_{0}\right)=(011)$ as

$$
\left[\begin{array}{l}
z_{0} \\
z_{1} \\
z_{2}
\end{array}\right]=\left[\begin{array}{llll}
1 & 0 & 0 & 0 \\
0 & 1 & 1 & 1 \\
1 & 1 & 1 & 0
\end{array}\right]\left[\begin{array}{l}
1 \\
0 \\
1 \\
0
\end{array}\right]=\left[\begin{array}{l}
1+0+0+0 \\
0+0+1+0 \\
1+0+1+0
\end{array}\right]=\left[\begin{array}{l}
1 \\
1 \\
0
\end{array}\right]
$$

## Proving Universality

## Theorem

For $x \neq y, P[h(x)=h(y)]=1 / m=2^{-b}$, therefore the class of matrix hash functions with a randomly selected 0-1 matrices is universal.

- Take an arbitrary $x$ and $y$
- They must differ in at least one bit position
- Assume that $x$ and $y$ differ in the ith bit, i.e., they are given as $\left(x_{u-1} \cdots x_{i} \cdots x_{1} x_{0}\right)$ and $\left(y_{u-1} \cdots y_{i} \cdots y_{1} y_{0}\right)$ such that $x_{i} \neq y_{i}$
- WLOG, assume $x_{i}=0$ and $y_{i}=1$
- Now choose the entire $A$ matrix except its $i$ th column


## Proving Universality

- Since this is the column that multiplies the ith bit $x$ or $y$, the hash values $h(x)$ and $h(y)$ are the same, except the contribution of the $i$ th column of $A$ is not included yet
- The length of $i$ th column is $b$, and there are $2^{b}$ different choices for this column
- Every time we change a bit in this column, we flip the corresponding bit in $h(y)$ since $y_{i}=1$
- There are exactly one in $2^{b}$ chance that $h(x)=h(y)$
- Therefore, the hash function is universal


## Proving Universality

- Consider $x=\left(x_{3} x_{2} x_{1} x_{0}\right)=(0101)$ and $y=\left(y_{3} y_{2} y_{1} y_{0}\right)=(1101)$ so that $x$ and $y$ differ only in the 3rd bit $x_{3} \neq y_{3}$

$$
\begin{gathered}
{\left[\begin{array}{l}
z_{0} \\
z_{1} \\
z_{2}
\end{array}\right]=\left[\begin{array}{llll}
1 & 0 & 0 & \mathbf{0} \\
0 & 1 & 1 & \mathbf{1} \\
1 & 1 & 1 & \mathbf{0}
\end{array}\right]\left[\begin{array}{l}
1 \\
0 \\
1 \\
\mathbf{0}
\end{array}\right] \text { and }\left[\begin{array}{l}
z_{0}^{\prime} \\
z_{1}^{\prime} \\
z_{2}^{\prime}
\end{array}\right]=\left[\begin{array}{llll}
1 & 0 & 0 & \mathbf{0} \\
0 & 1 & 1 & \mathbf{1} \\
1 & 1 & 1 & \mathbf{0}
\end{array}\right]\left[\begin{array}{l}
1 \\
0 \\
1 \\
\mathbf{1}
\end{array}\right]} \\
{\left[\begin{array}{l}
z_{0} \\
z_{1} \\
z_{2}
\end{array}\right]} \\
=\left[\begin{array}{l}
(1 \cdot 1+0 \cdot 0+0 \cdot 1)+\mathbf{0} \cdot 0 \\
(0 \cdot 1+1 \cdot 0+1 \cdot 1)+\mathbf{1} \cdot 0 \\
(1 \cdot 1+1 \cdot 0+0 \cdot 1)+\mathbf{0} \cdot 0
\end{array}\right] \\
{\left[\begin{array}{l}
z_{0}^{\prime} \\
z_{1}^{\prime} \\
z_{2}^{\prime}
\end{array}\right]}
\end{gathered}=\left[\begin{array}{l}
(1 \cdot 1+0 \cdot 0+0 \cdot 1)+\mathbf{0} \cdot 1 \\
(0 \cdot 1+1 \cdot 0+1 \cdot 1)+\mathbf{1} \cdot 1 \\
(1 \cdot 1+1 \cdot 0+0 \cdot 1)+\mathbf{0} \cdot 1
\end{array}\right] \$
$$

## Proving Universality

- The contribution of the first three columns of the $A$ matrix to the hash value is the same, and the difference occurs in the contribution of the last column

$$
\left[\begin{array}{l}
z_{0} \\
z_{1} \\
z_{2}
\end{array}\right]=\left[\begin{array}{l}
(1)+\mathbf{0} \cdot 0 \\
(1)+\mathbf{1} \cdot 0 \\
(1)+\mathbf{0} \cdot 0
\end{array}\right] \text { and }\left[\begin{array}{l}
z_{0}^{\prime} \\
z_{1}^{\prime} \\
z_{2}^{\prime}
\end{array}\right]=\left[\begin{array}{l}
(1)+\mathbf{0} \cdot 1 \\
(1)+\mathbf{1} \cdot 1 \\
(1)+\mathbf{0} \cdot 1
\end{array}\right]
$$

- As we use $A$ matrices each of which is different in the last column (there are 8 such columns), we obtain different $\left[z_{0}^{\prime}, z_{1}^{\prime}, z_{2}^{\prime}\right]^{T}$ vectors

$$
\left[\begin{array}{l}
z_{0} \\
z_{1} \\
z_{2}
\end{array}\right]=\left[\begin{array}{l}
1 \\
1 \\
1
\end{array}\right] \text { and }\left[\begin{array}{l}
z_{0}^{\prime} \\
z_{1}^{\prime} \\
z_{2}^{\prime}
\end{array}\right]=\left[\begin{array}{l}
(1)+\mathbf{0} \cdot 1 \\
(1)+\mathbf{1} \cdot 1 \\
(1)+\mathbf{0} \cdot 1
\end{array}\right]
$$

## Proving Universality

- Only in 0 case in which the last column is $[0,0,0]^{T}$, we will obtain $\left[z_{0}^{\prime}, z_{1}^{\prime}, z_{2}^{\prime}\right]^{T}=\left[z_{0}, z_{1}, z_{2}\right]^{T}$, which is the case when the last column of $A$ is selected as $[0,0,0]^{T}$

$$
\begin{aligned}
& {\left[\begin{array}{l}
(1)+\mathbf{0} \cdot \mathbf{1} \\
(1)+\mathbf{0} \cdot 1 \\
(1)+\mathbf{0} \cdot 1
\end{array}\right]=\left[\begin{array}{l}
1 \\
1 \\
1
\end{array}\right] ;\left[\begin{array}{l}
(1)+\mathbf{0} \cdot \mathbf{1} \\
(1)+\mathbf{0} \cdot 1 \\
(1)+\mathbf{1} \cdot 1
\end{array}\right]=\left[\begin{array}{l}
1 \\
1 \\
0
\end{array}\right] ;\left[\begin{array}{l}
(1)+\mathbf{0} \cdot \mathbf{1} \\
(1)+\mathbf{1} \cdot 1 \\
(1)+\mathbf{0} \cdot 1
\end{array}\right]=\left[\begin{array}{l}
1 \\
0 \\
1
\end{array}\right] ;\left[\begin{array}{l}
(1)+\mathbf{0} \cdot \mathbf{1} \\
(1)+\mathbf{1} \cdot 1 \\
(1)+\mathbf{1} \cdot 1
\end{array}\right]=\left[\begin{array}{l}
1 \\
0 \\
0
\end{array}\right]} \\
& {\left[\begin{array}{l}
(1)+\mathbf{1} \cdot \mathbf{1} \\
(1)+\mathbf{0} \cdot 1 \\
(1)+\mathbf{0} \cdot 1
\end{array}\right]=\left[\begin{array}{l}
0 \\
1 \\
1
\end{array}\right] ;\left[\begin{array}{l}
(1)+\mathbf{1} \cdot \mathbf{1} \\
(1)+\mathbf{0} \cdot 1 \\
(1)+\mathbf{1} \cdot \mathbf{1}
\end{array}\right]=\left[\begin{array}{l}
0 \\
1 \\
0
\end{array}\right] ;\left[\begin{array}{l}
(1)+\mathbf{1} \cdot \mathbf{1} \\
(1)+\mathbf{1} \cdot \mathbf{1} \\
(1)+\mathbf{0} \cdot 1
\end{array}\right]=\left[\begin{array}{l}
0 \\
0 \\
1
\end{array}\right] ;\left[\begin{array}{l}
(1)+\mathbf{1} \cdot \mathbf{1} \\
(1)+\mathbf{1} \cdot 1 \\
(1)+\mathbf{1} \cdot 1
\end{array}\right]=\left[\begin{array}{l}
0 \\
0 \\
0
\end{array}\right]}
\end{aligned}
$$

- Therefore $h(x)=h(y)$ only in 1 out of 8 cases
- There are exactly one in $2^{b}$ chance that $h(x)=h(y)$


## Construction of the Dot-Product Mod $m$ Method

- Let $m$ be prime
- Decompose the key $x$ into $r+1$ digits each with the value in the set $\mathcal{Z}_{m}=\{0,1,2, \ldots, m-1\}$
- We have $x=\left(x_{r} x_{r-1} \cdots x_{1} x_{0}\right)$ with $x_{i} \in \mathcal{Z}_{m}$
- Let $a=\left(a_{r} a_{r-1} \cdots a_{1} a_{0}\right)$ be a random vector such that $a_{i} \in \mathcal{Z}_{m}$
- Define the hash function family as

$$
h_{a}(x)=\sum_{i=0}^{r} a_{i} x_{i} \quad(\bmod m)
$$

- The size of $\mathcal{H}$ is $m^{r+1}$


## Proving Universality

## Theorem

The set $\mathcal{H}=\left\{h_{a}\right\}$ is universal.

- Let $x=\left(x_{r} \cdots x_{1} x_{0}\right)$ and $y=\left(y_{r} \cdots y_{1} y_{0}\right)$ be two distinct keys
- Thus, they differ in at least one digit position, WLOG position 0
- For how many $h_{a} \in \mathcal{H}$ do $x$ and $y$ collide?
- The equality $h(x)=h(y)$ implies

$$
\sum_{i=0}^{r} a_{i} x_{i}=\sum_{i=0}^{r} a_{i} y_{i} \quad(\bmod m)
$$

## Proving Universality

- Equivalently we have

$$
\begin{aligned}
\sum_{i=0}^{r} a_{i}\left(x_{i}-y_{i}\right) & =0(\bmod m) \\
a_{0}\left(x_{0}-y_{0}\right)+\sum_{i=1}^{r} a_{i}\left(x_{i}-y_{i}\right) & =0(\bmod m) \\
a_{0}\left(x_{0}-y_{0}\right) & =-\sum_{i=1}^{r} a_{i}\left(x_{i}-y_{i}\right)(\bmod m)
\end{aligned}
$$

## Proving Universality

- Since $x_{0} \neq y_{0}$ and $m$ is prime, the inverse $\left(x_{0}-y_{0}\right)^{-1}(\bmod m)$ exists, which implies

$$
a_{0}=-\left(x_{0}-y_{0}\right)^{-1}\left[\sum_{i=1}^{r} a_{i}\left(x_{i}-y_{i}\right)\right] \quad(\bmod m)
$$

- Thus, for any choices of $a_{1}, a_{2}, \ldots, a_{r}$, exactly one choice of $a_{0}$ causes $x$ and $y$ collide
- How many $h_{a}$ functions cause $x$ and $y$ collide?


## Proving Universality

- There are $m$ choices for each of $a_{1}, a_{2}, \ldots, a_{r}$ but once they are chosen, there is only once choice of $a_{0}$ that causes $x$ and $y$ collide
- Therefore, the number of hash functions that causes $x$ and $y$ collide is

$$
m^{r} \cdot 1=m^{r}=\frac{m^{r+1}}{m}=\frac{|\mathcal{H}|}{m}
$$

that makes $\mathcal{H}$ a universal hash function family

## Perfect Hashing

- A hashing technique is called perfect hashing if $O(1)$ memory accesses are required to perform a search in the worst case
- To create a perfect hashing, we use two levels of hashing, with universal hashing at each level



## Perfect Hashing

- The first level is the same as hashing with chaining: we hash $n$ keys into $m$ slots using a hash function $h$ from a family of universal hash functions
- However, instead of making a linked list of keys hashing to slot $j$, we use a secondary hash table $S_{j}$ with an associate hash function $h_{j}$
- By choosing the hash functions $h_{j}$ carefully, we can guarantee that there are no collisions at the secondary level
- In order to guarantee that there are no collisions on the secondary level, we need to let the size $m_{j}$ of the hash table $S_{j}$ be the square of the number $n_{j}$ of keys hashing to slop $j$


## Perfect Hashing

- Consider the key set $K=\{10,22,37,40,52,60,70,72,74\}$
- The first level hash function is

$$
h(k)=(a k+b \bmod p) \bmod m
$$

with parameters $(m, a, b, p)=(9,3,42,101)$, where $m$ is the table size

- For example, $h(75)$ is computed as

$$
\begin{aligned}
h(75) & =(3 \cdot 75+42 \bmod 101) \bmod 9 \\
& =(267 \bmod 101) \bmod 9 \\
& =65 \bmod 9 \\
& =2
\end{aligned}
$$

## Perfect Hashing

- A secondary hash table $S_{j}$ stores all keys hashing to slot $j$
- The size of hash table $S_{j}$ is $m_{j}=n_{j}^{2}$, where $n_{j}$ is the number of keys hashing to slot $j$
- The associated hash function of $S_{j}$ is

$$
h_{j}(k)=\left(a_{j} k+b_{j} \bmod p\right) \bmod m_{j}
$$



## Perfect Hashing

- On the second level, we use the hash function belonging to Slot 2, which has the parameters $\left(m_{2}, a_{2}, b_{2}\right)=(9,10,18)$ and the same prime $p=101$, therefore, we compute $h_{2}(75)$ as

$$
\begin{aligned}
h_{2}(75) & =(10 \cdot 75+18 \bmod 101) \bmod 9 \\
& =7
\end{aligned}
$$

and place the key 75 in the 7 th cell of the Slot 2 table


## Perfect Hashing Properties

- If we store $n$ keys in a hash table of size $m=n^{2}$ using a universal hash function, then the probability of collision is $1 / 2$
- There are $C(n, 2)$ pairs of different pairs of keys
- The probability that a pair collides is $1 / m$, if $h$ is chosen from $\mathcal{H}$
- Let $X$ be the number of collisions, since $m=n^{2}$, the expected value of $X$ is

$$
E[X]=C(n, 2) \cdot \frac{1}{n^{2}}=\frac{n(n-1)}{2} \cdot \frac{1}{n^{2}}<\frac{1}{2}
$$

## Perfect Hashing Properties

- Since we choose $m=n^{2}$, a hash function $h$ chosen at random from $\mathcal{H}$ is more likely not to have collisions
- Given a static set of $n$ keys, it is easy to find a collision-free hash function $h$
- When $n$ is large, a hash table of size $m=n^{2}$ is excessive
- However, in the two-level approach we only hash the entries in each slot
- On the first level the hash function $h$ hashes $n$ keys into $m=n$ slots
- Then, if $n_{j}$ keys hash to slot $j m$ we use the secondary hash table of size $m_{j}=n_{j}^{2}$ to provide a collision-free constant-time lookup


## Perfect Hashing Storage Requirement

- In the first level table size is $m=n$, and therefore, the amount of the memory used is $O(n)$ for the primary hash table
- In the secondary hash tables, each hash table $S_{j}$ is of size $n_{j}^{2}$
- To compute the total memory used in the secondary tables, we need to know the expected sum of the squares of the number of keys $n_{j}$ that hash to slot $j$, which turns out to be

$$
E\left[\sum_{j=0}^{m-1} m_{j}\right]=E\left[\sum_{j=0}^{m-1} n_{j}^{2}\right]<2 n
$$

- Therefore, the total secondary storage is also $O(n)$

