









BLOOM Filter Lightweight Data Structure

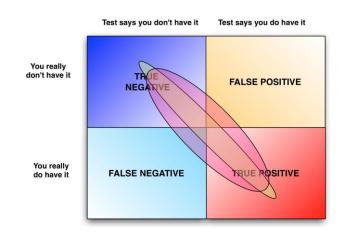


Traduction en cours

Sketching Data Structures



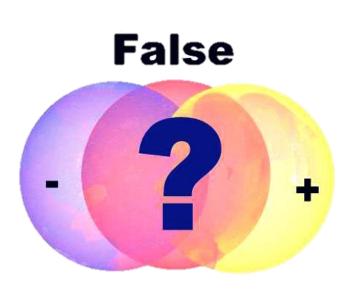
The main feature of sketching data structures is that they can answer certain questions about the data extremely efficiently, at the price of the occasional error.



False Positif



- Bloom filter allows to get away with much less storage at the price of an occasional errors.
- Errors can only be a false positive
- Bloom Filter might say that x is in A when in fact it is not.
- When BF says that x is not in A,
 - This is always true,
 - False negatives are impossible.



Bloom filters



- Bloom Filter is representing which elements are present in a set.
- Bloom filters are used to reduce expensive disk (or network) lookups for non-existent keys.
- The Bloom Filter is a data structure that compactly represents a set as a bitmap which is updated via hashing.

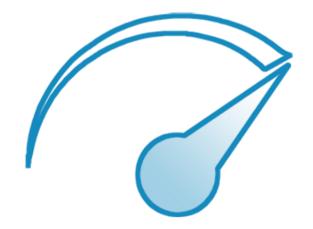


Bloom Filter common usage



- Browser tests for malicious website
- Test for weak passwords
- Propose a list of items not seen by a given user
- Bitcoin
- Web cache hit cache





Bloom Filter sub-linear space



- Bloom filter doesn't store the actual elements, it stores the "membership" of them.
- It uses sub-linear space opening the possibility for false positives, meaning there's a non-zero probability it reports an item is in the set when it's actually not.

Burton H. Bloom



The data structure was conceived by Burton H. Bloom in 1970.



Source of the exemple



- Theory and Practice of Bloom Filters for Distributed Systems
- Sasu Tarkoma, Christian Esteve Rothenberg, and Eemil Lagerspetz

Size constant



- It does not matter how many elements do we store the space will be the same.
- A bloom filter with 10³ elements will take the same amount of space as a bloom filter with 10³ elements and the same space as bloom filter with 0 elements.
- Trade of:
 - The more elements, the more uncertain is the "in the set answer".

Constant reponse time



• All the operations are taking constant time.



Exemple with alphabet letters



The set

Α	В	С	D	Ш	F	G	Η		J	K	L	М	Ν	0	Р	Q	R	S	Т	U	V	W	Х	Υ	Z
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26

Size = 27 * Short = 432 bytes

Probalistic Data Structure

Size = 2 bytes

Test if a letter is present in the set.



· · ·		
Rawid	Short	Short
1	Α	0
2	В	0
3	С	1
4	D	0
5	A B C D	0 0 1 0 0
6	F	1
7	F G H	1
2 3 4 5 6 7 8 9	Н	1
9	I	0
10	J	1 <
11	K	0
12	L	0
13	М	1
14	Z	0
15	0	0
16	Р	1
17	Q	0
18	R	0
19	S	1
20	Т	0
21	U	0
22	V	1
23	W	0
23 24 25	X	0
25	J K L M N O P Q R S T U V W X Y Z	1 < 0
26	Z	0

In the set

Size = 27 * Short * 2 = 864 bytes

Size exemples



binary-array of millions of elements

	N	bits	size∄byte	Mo
bits	77777210000000	8	7777712251000?	777777777777777 25?

char⊋per⊡vord			Char 216 16 bits	sizedndbits	Мо	Go	
words	?????? 1000000?	6	2776 00000000000000000000000000000000000	16	96100010002	77779 610002	??????? ? 6?

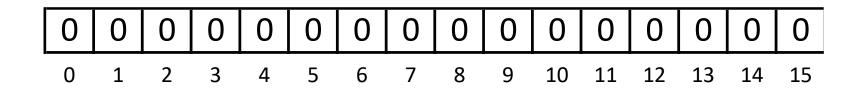
#1chars

Bit array



 A Bloom filter is an array of m bits for representing a set

 $S = \{x_1, x_2, \dots, x_n\}$ of n elements



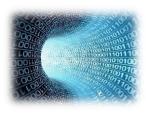
Hash functions



- The key idea is to use k hash functions,
 - $h_i(x)$, $1 \le i \le k$ to map items $x \in S$
 - To random numbers uniform in the range $\mathbf{1}, \ldots$ m.
- The hash functions are assumed to be uniform.

 We write a hash function that, instead of a single hash code, produces k hash codes for a given object.

Multiple Hash Function



 If a Bloom filter returns that an item is member of the set, there's a certain probability for a false positive.

Inserting and testing



- An element $x \in S$ is inserted into the filter by setting the bits $h_i(x)$ to one $1 \le i \le k$.
- Conversely, y is assumed a member of S if the bits h_i(y) are set, and guaranteed not to be a member if any bit h_i(y) is not set.

Exemple



16 bits map

Three hash functions are used: h₁, h₂, and h₃, being MD5, SHA1 and CRC32, respectively.

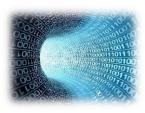
When adding an element, the values of h₁ through h₃ (modulo 16) are calculated for the element, and corresponding bit positions are set to one.

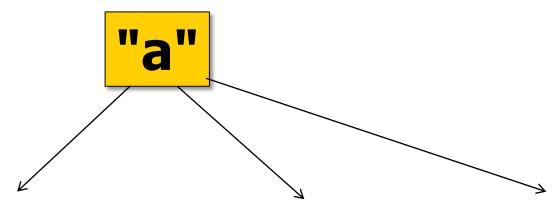
Steps



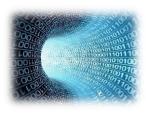
- Inserting a
- Inserting b
- Inserting I
- Inserting y

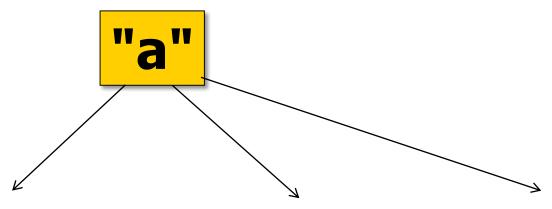
- Testing q
- Testing z





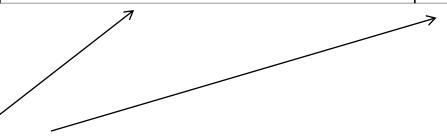
Hash	MD5	SHA1	CRC32
hi	h1	h2	h3🛚
		86f7e437faa5a7fce15d1ddcb9eaeaea377667b8	e8b7be43





Hash	MD5	SHA1	CRC32
hi	h1	h2	h3?
а	0cc175b9c0f1b6a831c399e269772661	86f7e437faa5a7fce15d1ddcb9eaeaea377667b8	e8b7be43

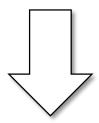
Modulo 16 ???



Binary Modulo Operation



(number)_{base} % base^y



Last y digits of (number)_{base}

Modulo 10^y base 10



Χ	Р	mod
5	10	5
12	10	2
25	10	5
36	10	6
45	10	5
55	10	5
68	10	8
74	10	4
89	10	9
92	10	2

X	Р	mo	С
68	100	68	
105	100	5	
257	100	57	
396	100	96	
458	100	58	
589	100	89	
623	100	23	
758	100	58	
899	100	99	
985	100	85	

X	Р	mod
750	1000	750
1505	1000	505
2569	1000	569
3201	1000	201
4580	1000	580
5692	1000	692
6548	1000	548
7845	1000	845
8592	1000	592
9541	1000	541

Modulo 10^y base 10



X	Р	m	od
68	10	8	
105	10	5	
257	10	7	
396	10	6	
458	10	8	
589	10	9	
623	10	3	
758	10	8	
899	10	9	
985	10	5	

Χ	Р	mod
750	100	50
1505	100	5
2569	100	69
3201	100	1
4580	100	80
5692	100	92
6548	100	48
7845	100	45
8592	100	92
9541	100	41

X	Р	mod
1508	1000	508
10250	1000	250
25641	1000	641
36987	1000	987
45871	1000	871
51489	1000	489
65478	1000	478
73581	1000	581
82467	1000	467
98547	1000	547

Modulo 10^y base 10



10^y	y	=	1
-----------------------	---	---	---

X		\downarrow	Р	mod
5	0	5	10	5
12	1	2	10	2
25	2	5	10	5
36	3	6	10	6
45	4	5	10	5
55	5	5	10	5
68	6	8	10	8
74	7	4	10	4
89	8	9	10	9
92	9	2	10	2

$$10^{y} y = 2$$

	\downarrow	Р	mo	d
0	68	100	68	
1	05	100	5	
2	57	100	57	
3	96	100	96	
4	58	100	58	
5	89	100	89	
6	23	100	23	
7	58	100	58	
8	99	100	99	
9	85	100	85	
	1 2 3 4 5 6 7 8	1 05 2 57 3 96 4 58 5 89 6 23 7 58 8 99	0 68 100 1 05 100 2 57 100 3 96 100 4 58 100 5 89 100 6 23 100 7 58 100 8 99 100	0 68 100 68 1 05 100 5 2 57 100 57 3 96 100 96 4 58 100 58 5 89 100 89 6 23 100 23 7 58 100 58 8 99 100 99

$$10^{y} y = 3$$

X		\downarrow	Р	mod
750	0	750	1000	750
1505	1	505	1000	505
2569	2	569	1000	569
3201	3	201	1000	201
4580	4	580	1000	580
5692	5	692	1000	692
6548	6	548	1000	548
7845	7	845	1000	845
8592	8	592	1000	592
9541	9	541	1000	541

Last y digits of (number)_{base}

Binary Modulo Operation



Modulo 16

			Di	git							
7	6	5	4	3	2	1	0	Bin	Hex	Dec	Mod16
1	0	0	0	1	0	0	1	0b10001001	0x89	137	9
0	1	0	0	0	1	0	0	0b01000100	0x44	68	4
0	0	1	0	0	0	1	0	0b00100010	0x22	34	2
0	0	0	1	0	0	0	1	0b00010001	0x11	17	1
0	0	0	0	1	0	0	0	0b00001000	0x8	8	8

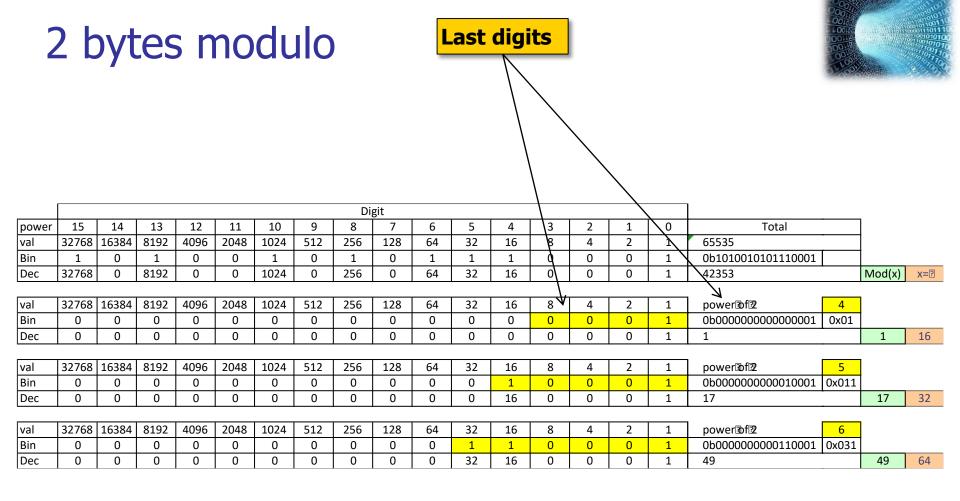
 $16 = 2^4$

Last 4 digits

y = **4**

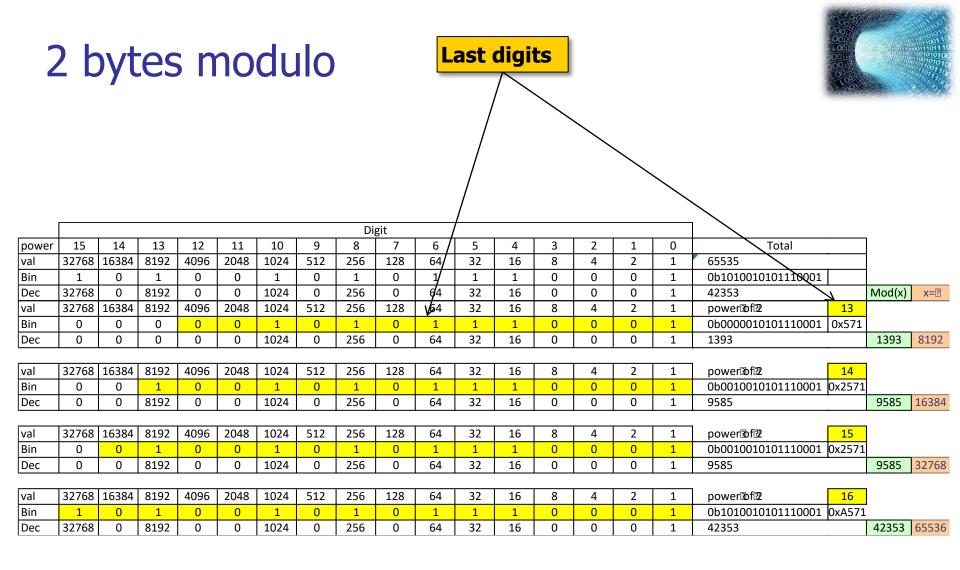
Last y digits of (number)_{base}

Last digits 1 byte modulo Mod(x)Power of 22 Bin Hex Dec Χ 0b11011001 0xD9 $\mathbf{0}$ 0b00000001 0x01 0b11011001 0xD9 $\mathbf{0}$ $\mathbf{0}$ $\mathbf{0}$ $\mathbf{0}$ 0b00000001 0x01 0b11011001 $\mathbf{0}$ 0xD9 $\mathbf{0}$ 0b00000001 0x010b11011001 0xD9 0b00001001 0x09 0b11011001 0xD9 $\mathbf{0}$ 0b00011001 0x19 0b11011001 0xD9 $\mathbf{0}$ 0b00011001 0x19 0b11011001 0xD9 0b01011001 0x59 0b11011001 0xD9 0b11011001 0xD9

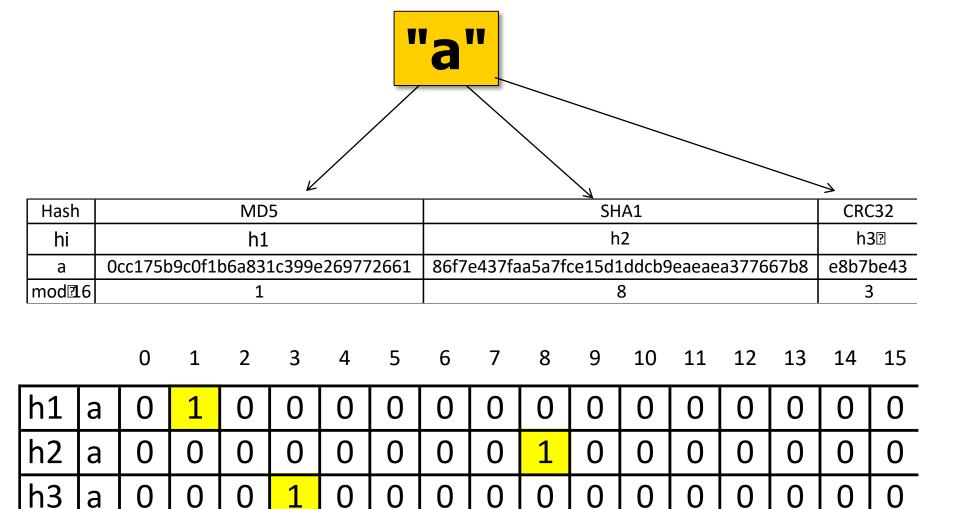


2 bytes modulo **Last digits** 32768 16384 **√**8 power 3 of 22 val Bin 0b000000001110001 0x071 Dec power of 2 val 32768 16384 0b0000000001110001 0x071 Bin Dec val power of 2 Bin 0b0000000101110001 0x171 Dec val power of 22 0b0000000101110001 0x171 Bin Dec val power of 22 0x571 Bin 0b0000010101110001 Dec power 3bf 22 val 0x571 0b0000010101110001 Bin Dec

Emmanuel Fuchs



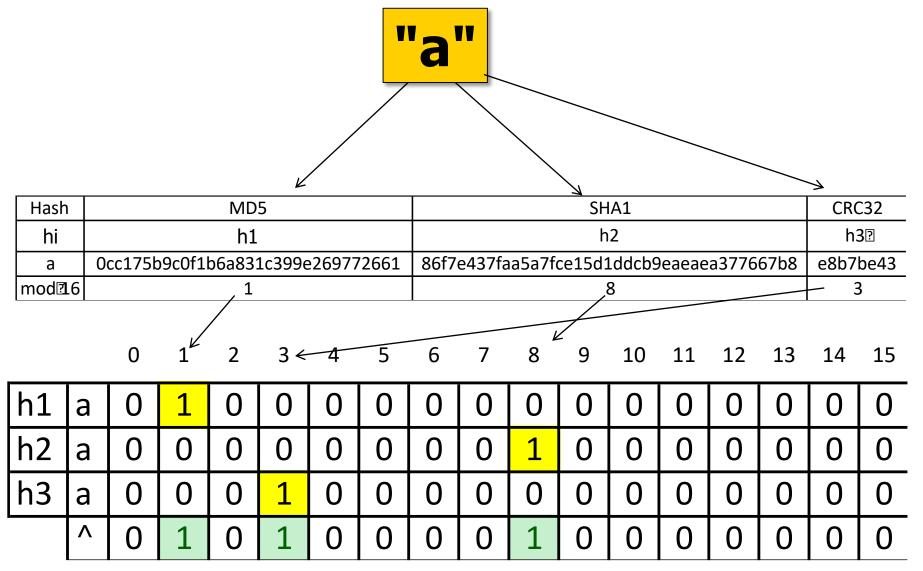




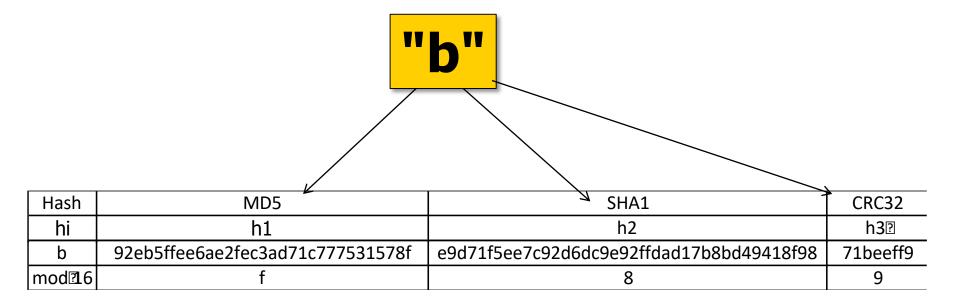
()

Emmanuel Fuchs

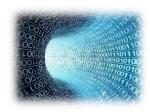


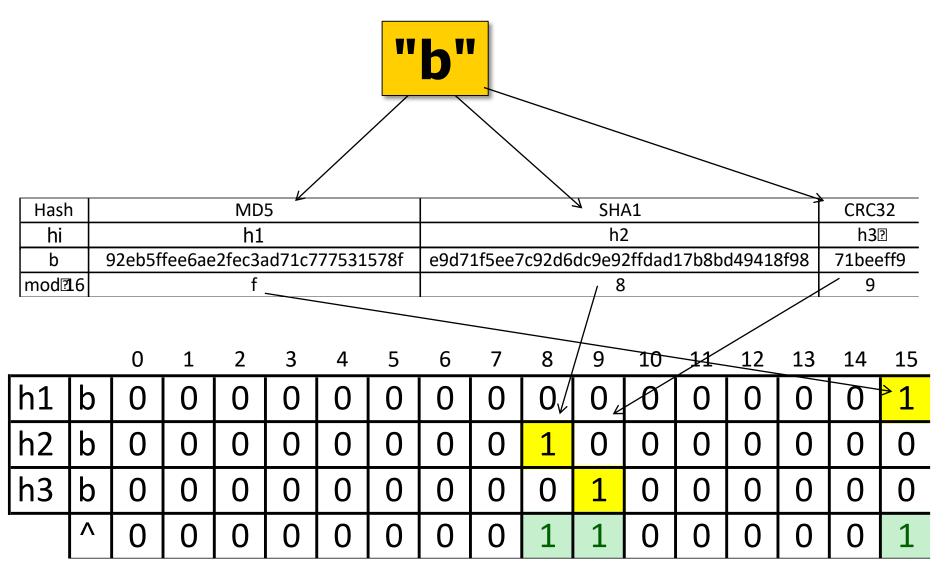






		0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
h1	b	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
h2	b	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0
h3	b	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0
	٨	0	0	0	0	0	0	0	0	1	1	0	0	0	0	0	1





a b Λ

Insert y



Hash				MD5	,						SH	A1				CR	C32	
hi				h1							h	2				h	3?	
У	41	.52907	695944	460e2e	48592	22904f	345d	95cb0	bfd29	77c76	1298d	9624e	4b4d4	c72a3	9974a	a fbdb2615		
mod16	5			d							, ;	a				5		
	•	0 1 2 3 4 5 v 0 0 0 0 0							7	8	9	10	11	12	13	14	15	
h1	У	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	
h2	У	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	
h3	У	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0	
	۸	0	0	0	0	0	1	0	0	0	0	1	0	0	1	0	0	

Emmanuel Fuchs

Insert I



Hash	n			MD	5						Sŀ	HA1				CR	C32
hi				h1							ł	12				h	3?
I		2db95e8	3e1a92	.67b7a	11885	56b20	13b33	07c3	42be6	e560e	7f4384	12e2e2	21b774	e61d8	5f047	9606	5c2fe
mod₫	16			3								7				ı	e
		0	5	6	7	8	9	10	11	12	13	14	15				
h1		0	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0
h2		0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0
h3		0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0
	^	0	0	0	1	0	0	0	1	0	0	0	0	0	0	1	0

Emmanuel Fuchs

Set map



У	0	0	0	0	0	1	0	0	0	0	1	0	0	1	0	0
	0	0	0	1	0	0	0	1	0	0	0	0	0	0	1	0
٨	0	0	0	1	0	1	0	1	0	0	1	0	0	1	1	0
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

ab	0	1	0	1	0	0	0	0	1	1	0	0	0	0	0	1
yΙ	0	0	0	1	0	1	0	1	0	0	1	0	0	1	1	0
^	0	1	0	1	0	1	0	1	1	1	1	0	0	1	1	1
•	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

Probe q and z



Hash	MD5	SHA1	CRC32
hi	h1	h2	h3?
q	7694f4a66316e53c8cdd9d9954bd611d	22ea1c649c82946aa6e479e1ffd321e4a318b1b0	f500ae27
mod16	d	0	7
Z	fbade9e36a3f36d3d676c1b808451dd7	395df8f7c51f007019cb30201c49e884b46b92fa	62d277af
mod16	7	а	f

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
	0	1	0	1	0	1	0	1	1	1	1	0	0	1	1	1
q	1							1						1		
Z								1			1					1

Not@n\set In\set

False Positive

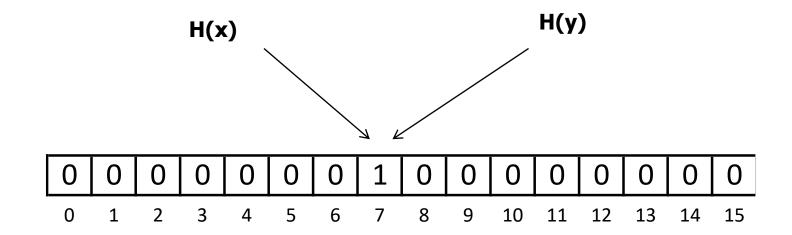


- The weak point of Bloom filters is the possibility for a false positive.
- False positives are elements that are not part of S but are reported being in the set by the filter.

Fault Positive



Collision



False Positive



If only a single hash function were used to indicate membership in the set, the probability of a false positive would be higher than using multiple hash functions.

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	_
	0	1	0	1	0	1	0	1	1	1	1	0	0	1	1	1	
q	1							1						1			Not₫r
Z								1			1					1	In \$et

n/\$et

False Positive



If only a single hash function were used to indicate membership in the set, the probability of a false positive would be higher than using multiple hash functions.

