

❏ Bloom Filter Calculator ❏

[Bloom filters](#) are space-efficient probabilistic data structures used to test whether an element is a member of a set.

They're surprisingly simple: take an array of **m** bits, and for up to **n** different elements, either test or set **k** bits using positions chosen using hash functions. If all bits are set, the element *probably* already exists, with a false positive rate of **p**; if any of the bits are not set, the element *certainly* does not exist.

Bloom filters find a wide range of uses, including tracking which [articles you've read](#), [speeding up Bitcoin clients](#), [detecting malicious web sites](#), and [improving the performance of caches](#).

This page will help you choose an optimal size for your filter, or explore how the different parameters interact.

n

Number of items in the filter (optionally with [SI units](#): k, M, G, T, P, E, Z, Y)

101988

p

Probability of false positives, fraction between 0 and 1 or a number indicating 1-in-p

0.0001

m

Number of bits in the filter (or a size with KB, KiB, MB, Mb, GiB, etc)

k

Number of hash functions

12

Submit

n = 101,988
p = 0.0001 (1 in 10,000)
m = 1,961,567 (239.45KiB)
k = 12

```
n = ceil(m / (-k / log(1 - exp(log(p) / k))))  
p = pow(1 - exp(-k / (m / n)), k)  
m = ceil((n * log(p)) / log(1 / pow(2, log(2))));  
k = round((m / n) * log(2));
```

