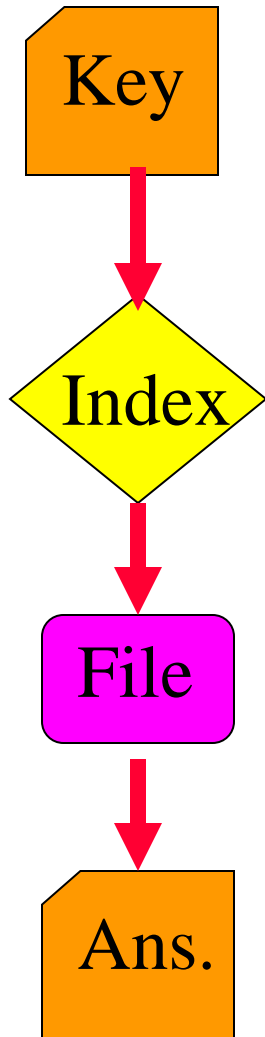


# Bloom Filters

- Differential Files
- Simple large database.
  - File of records residing on disk.
  - Single key.
  - Index to records.
- Operations.
  - Retrieve.
  - Update.
    - Insert a new record.
    - Make changes to an existing record.
    - Delete a record.

# Naïve Mode Of Operation

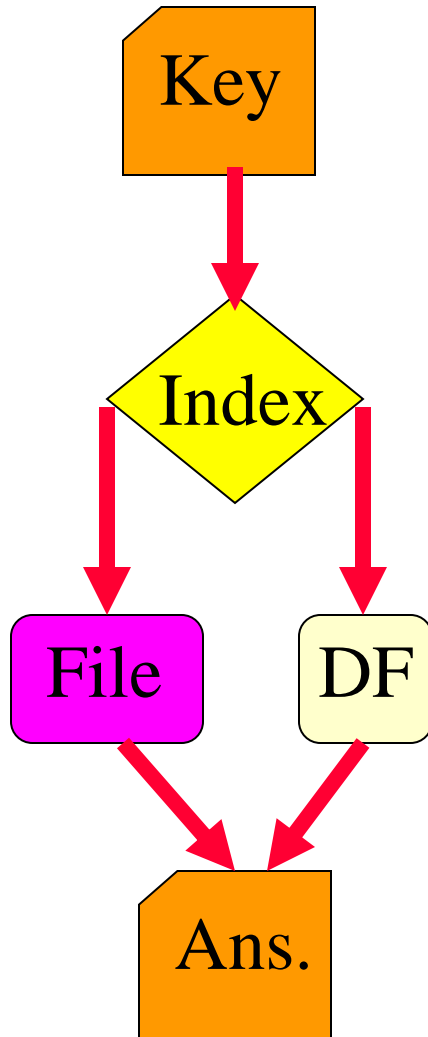


- Problems.
  - Index and File change with time.
  - Sooner or later, system will crash.
  - Recovery =>
    - Copy Master File (MF) from backup.
    - Copy Master Index (MI) from backup.
    - Process all transactions since last backup.
  - Recovery time depends on MF & MI size + #transactions since last backup.

# Differential File

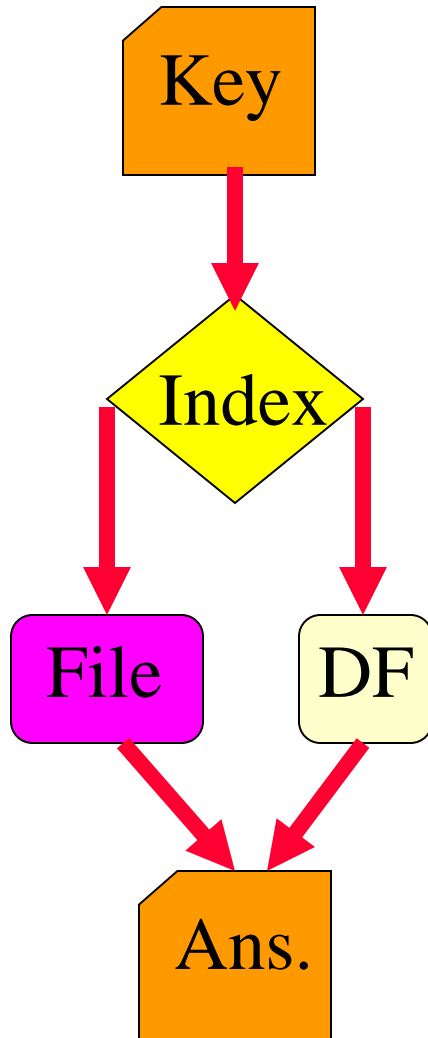
- Make no changes to master file.
- Alter index and write updated record to a new file called differential file.

# Differential File Operation



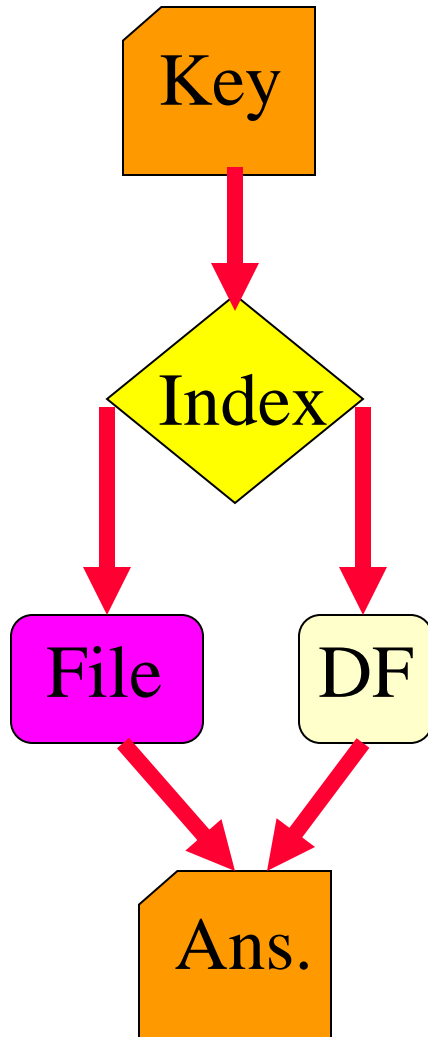
- Advantage.
  - DF is smaller than File and so may be backed up more frequently.
  - Index needs to be backed up whenever DF is. So, index should be no larger than DF.
  - Recovery time is reduced.

# Differential File Operation



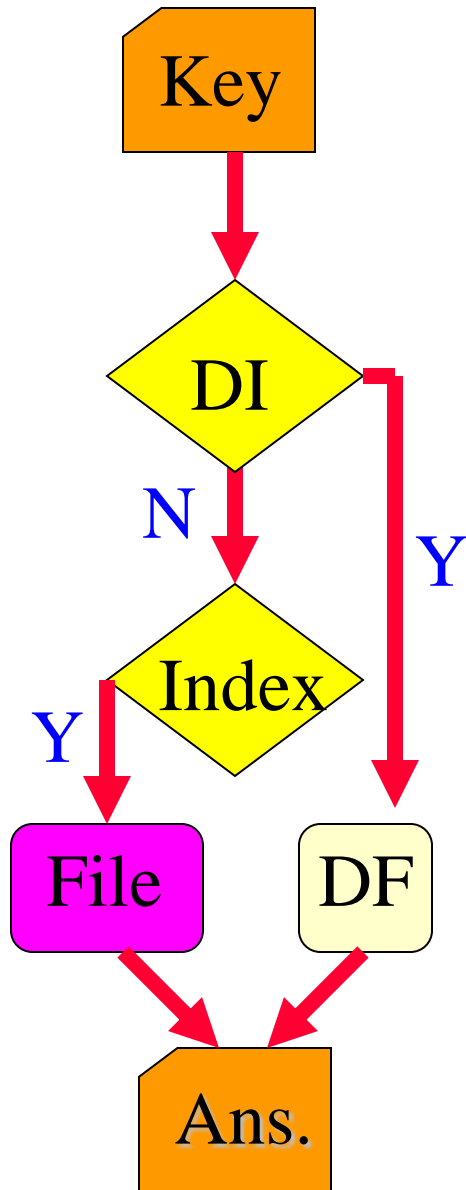
- Disadvantage.
  - Eventually DF becomes large and can no longer be backed up with desired frequency.
  - Must integrate File and DF now.
  - Following integration, DF is empty.

# Differential File Operation



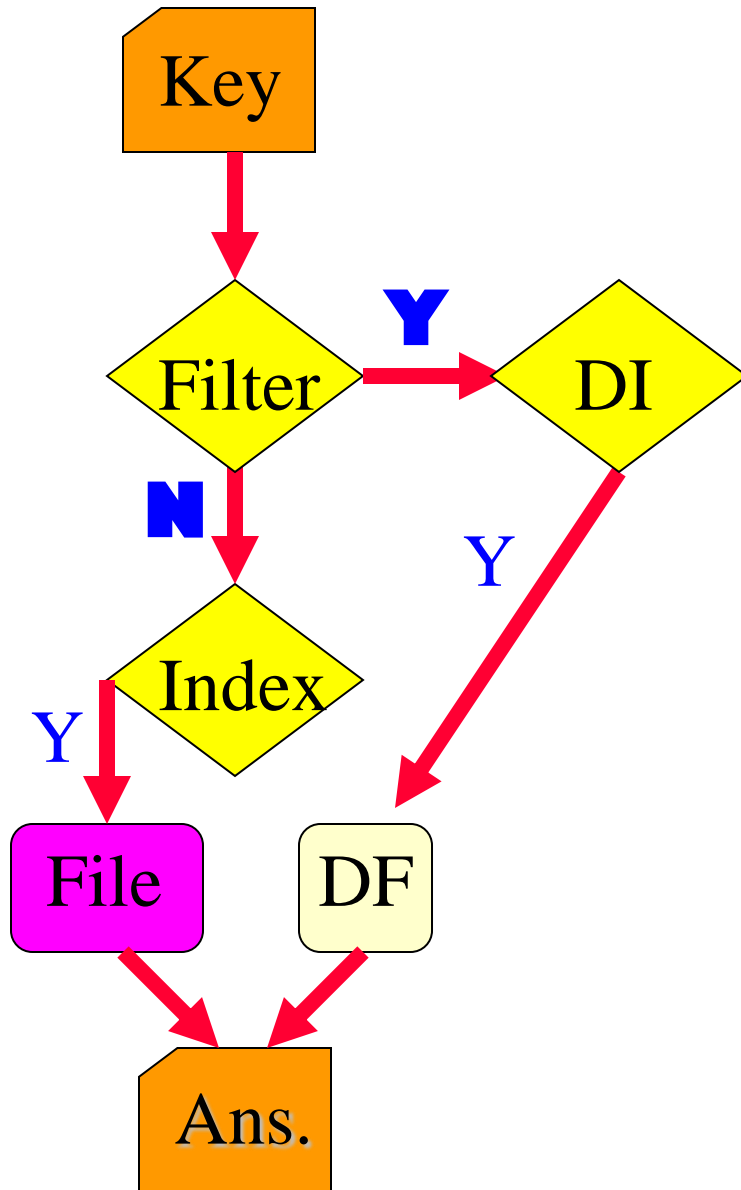
- Large Index.
  - Index cannot be backed up as frequently as desired.
  - Time to recover current state of index & DF is excessive.
  - Use a differential index.
    - Make no changes to Index.
    - DI is an index to all deleted records and updated records in DF.

# Differential File & Index Operation



- Performance hit.
  - Most queries search both DI and Index.
  - Increase in # of disk accesses/query.
- Use a filter to decide whether or not DI should be searched.

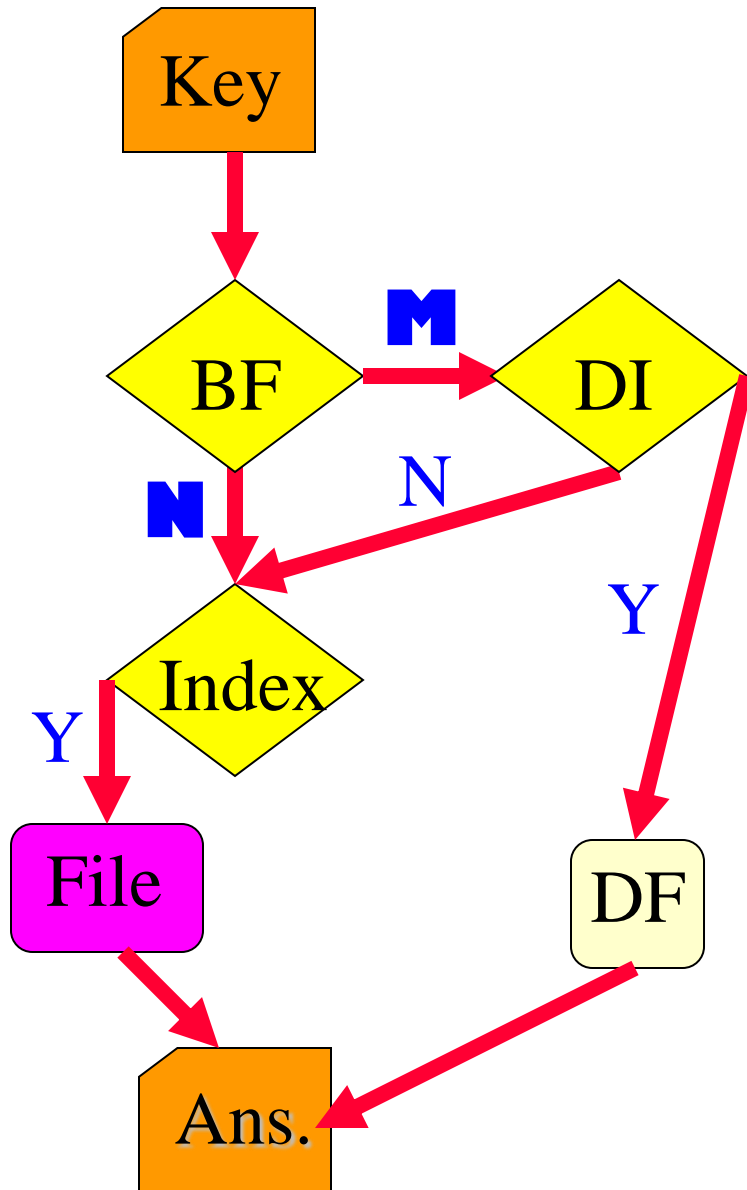
# Ideal Filter



- **Y**  $\Rightarrow$  this key is in the DI.
- **N**  $\Rightarrow$  this key is not in the DI.
- Functionality of ideal filter is same as that of DI.
- So, a filter that eliminates performance hit of DI doesn't exist.

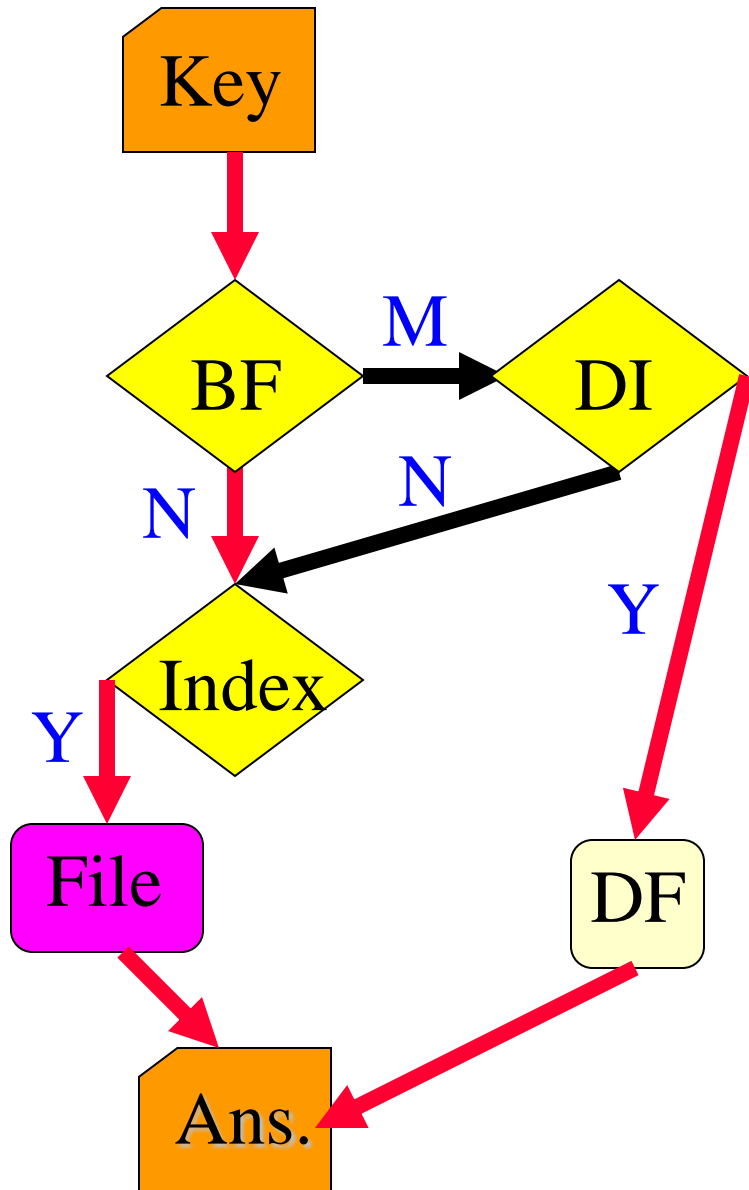


# Bloom Filter (BF)



- **N** => this key is not in the DI.
- **M** (maybe) => this key may be in the DI.
- Filter error.
  - BF says Maybe.
  - DI says No.

# Bloom Filter (BF)



- Filter error.
  - BF says Maybe.
  - DI says No.
- BF resides in memory.
- Performance hit paid only when there is a filter error.

# Bloom Filter Design

- Use  $m$  bits of memory for the BF.
- Larger  $m \Rightarrow$  fewer filter errors.
- When DI empty, all  $m$  bits = 0.
- Use  $h > 0$  hash functions:  $f_1()$ ,  $f_2()$ , ...,  $f_h()$ .
- When key  $k$  inserted into DI, set bits  $f_1(k)$ ,  $f_2(k)$ , ..., and  $f_h(k)$  to 1.
- $f_1(k)$ ,  $f_2(k)$ , ...,  $f_h(k)$  is the signature of key  $k$ .

# Example

0	1	0	0	1	0	1	0	1	0	0
0	1	2	3	4	5	6	7	8	9	

- $m = 11$  (normally,  $m$  would be much much larger).
- $h = 2$  (2 hash functions).
- $f_1(k) = k \bmod m$ .
- $f_2(k) = (2k) \bmod m$ .
- $k = 15$ .
- $k = 17$ .

# Example

0	1	0	0	1	0	1	0	1	0	0
0	1	2	3	4	5	6	7	8	9	

- DI has  $k = 15$  and  $k = 17$ .
- Search for  $k$ .
  - $f_1(k) = 0$  or  $f_2(k) = 0 \Rightarrow k$  not in DI.
  - $f_1(k) = 1$  and  $f_2(k) = 1 \Rightarrow k$  may be in DI.
- $k = 6 \Rightarrow$  filter error.

# Bloom Filter Design

- Choose **m** (filter size in bits).
  - Use as much memory as is available.
- Pick **h** (number of hash functions).
  - **h** too small  $\Rightarrow$  probability of different keys having same signature is high.
  - **h** too large  $\Rightarrow$  filter becomes filled with ones too soon.
- Select the **h** hash functions.
  - Hash functions should be relatively independent.

# Optimal Choice Of $h$

- Probability of a filter error depends on:
  - Filter size ...  $m$ .
  - # of hash functions ...  $h$ .
  - # of updates before filter is reset to 0 ...  $u$ .
    - Insert
    - Delete
    - Change
- Assume that  $m$  and  $u$  are constant.
- # of master file records =  $n \gg u$ .

# Probability Of Filter Error

- $p(u)$  = probability of a filter error after  $u$  updates  
=  $A * B$
- $A = p(\text{request for an unmodified record after } u \text{ updates})$
- $B = p(\text{filter bits are all } 1 \text{ for this request for an unmodified record})$



$A = p(\text{request for unmodified record})$

- $p(\text{update } j \text{ is for record } i) = 1/n.$
- $p(\text{record } i \text{ not modified by update } j) = 1 - 1/n.$
- $p(\text{record } i \text{ not modified by any of the } u \text{ updates})$   
 $= (1 - 1/n)^u$   
 $= A.$

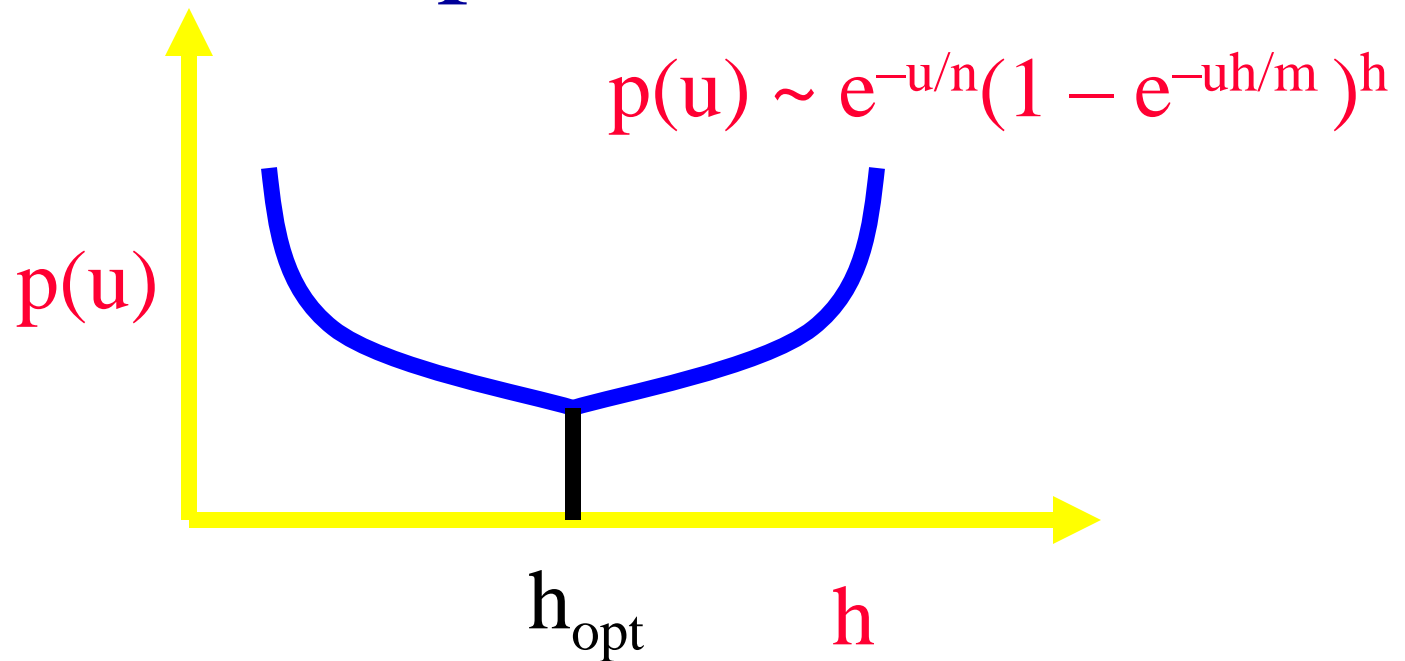
$B = p(\text{filter bits are all 1 for this request})$

- Consider an update with key  $K$ .
- $p(f_j(K) \neq i) = 1 - 1/m$ .
- $p(f_j(K) \neq i \text{ for all } j) = (1 - 1/m)^h$ .
- $p(\text{bit } i = 0 \text{ after one update}) = (1 - 1/m)^h$ .
- $p(\text{bit } i = 0 \text{ after } u \text{ updates}) = (1 - 1/m)^{uh}$ .
- $p(\text{bit } i = 1 \text{ after } u \text{ updates}) = 1 - (1 - 1/m)^{uh}$ .
- $p(\text{signature of } K \text{ is } 1 \text{ after } u \text{ updates})$   
 $\quad = [1 - (1 - 1/m)^{uh}]^h$   
 $\quad = B.$

# Probability Of Filter Error

- $p(u) = A * B$   
 $= (1 - 1/n)^u * [1 - (1 - 1/m)^{uh}]^h$
- $(1 - 1/x)^q \sim e^{-q/x}$  when  $x$  is large.
- $p(u) \sim e^{-u/n}(1 - e^{-uh/m})^h$
- $d p(u)/dh = 0 \Rightarrow h = (\ln 2)m/u \sim 0.693m/u$ .

# Optimal h



- $h \sim 0.693m/u$ .
- $m = 10^6, u = 10^6/2$ 
  - $h \sim 1.386$
  - Use  $h = 1$  or  $h = 2$ .

- $m = 2 \cdot 10^6, u = 10^6/2$ 
  - $h \sim 2.772$
  - Use  $h = 2$  or  $h = 3$ .