Master Theorem

• The Master Theorem applies to:

- T(n) = a T(n/b) + f(n)
- Where a >= 1 and b > 1

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- T(n) = a T(n/b) + f(n)
- Where a >= 1 and b > 1

- Note 1: if b < 1, the problem would get bigger, not smaller
- Note 2: if b = 1, the size of the problem would stay the same

- T(n) = a T(n/b) + f(n)
- Where a >= 1 and b > 1

- Let's first assume that f(n) = 0
- Note: this is not very realistic (but will help understand what is going on with the running time better)

- T(n) = a T(n/b)
- Let's drill down and iterate

- $T(n) = a(aT(n/b^2))$
- $T(n) = a^2 T(n/b^2)$

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- $T(n) = a^3 T(n/b^3)$

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- $T(n) = a^3 (a T(n/b^4))$
- $T(n) = a^4 T(n/b^4)$
- Do we see a pattern?

- $T(n) = a^k T(n/b^k)$
- When do we reach the base case?

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- When $n / b^k = 1$, i.e. $n = b^k$
- Then, $T(n) = a^k T(1)$

- What is the value of a ^k?
- (considering that $n = b^k$)

- $T(n) = a^k T(n/b^k)$
- What is the value of a ^k?
- $n = b^k$

• Consider $c = log_b a$

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- b to the power and log in base b are inverse functions of each other
- We take b to the power of each side, we get:
- $b^c = b^{\log}b^a = a$

- $T(n) = a^k T(1)$
- $n = b^{k}$
- $b^{c} = a$
- $a^k = (b^c)^k = b^c = b^k = (b^k)^c = n^c$

• $T(n) = n^{c} T(1)$

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- Where $c = log_b a$

- $T(n) = \Theta(n^c)$ where $c = \log_b a$
- .. for f(n) = 0 .. which is not very realistic

- T(n) = Θ (n^c) where c = $\log_b a$
- .. for f(n) = 0 .. which is not very realistic

- If b = a, then $T(n) = \Theta(n)$
- If b < a, then c > 1 (worse then $\Theta(n)$)
- If b > a, then c < 1 (better than $\Theta(n)$)

- T(n) = Θ (n^c) where c = $\log_b a$
- .. for f(n) = 0 .. which is not very realistic

- If b < a, then c > 1 (worse than $\Theta(n)$)
- Example: T(n) = 3 T(n/2)
- Problem size is divided by 2 but we have 3 recursive calls

- T(n) = Θ (n^c) where c = $\log_b a$
- .. for f(n) = 0 .. which is not very realistic

- If b > a, then c < 1 (better than $\Theta(n)$)
- Example: T(n) = 2 T(n/3)
- 2 recursive calls but problem size is divided by 3

- $T(n) = \Theta(n^c)$ where $c = \log_b a$
- .. for f(n) = 0 .. which is not very realistic
- If b > a, then c < 1 (better than $\Theta(n)$)

- Another example: T(n) = T(n/2)
- 1 recursive call but problem size is divided by 2 (Sounds similar to ??)

 Now let's look at scenarios where f(n) is NOT 0

- If f(n) is smaller than $\Theta(n^c)$
- Then $\Theta(n^c)$ is going to be the dominant factor and T(n) is going to be $\Theta(n^c)$

- If f(n) is $\Theta(n^c)$ then
- A logarithmic factor will be introduced and
- T(n) is going to be $\Theta(n^c \log n)$

- If f(n) is (strictly, from a Ω standpoint) bigger than $\Theta(n^c)$ then
- f(n) is going to be the dominant term and
- T(n) is going to be $\Theta(f(n))$