# Suggested Solutions - Exercises of Lecture 8 - MVP'06 

Alexandre David

## Exercise 5.1-Amdahl's law

$$
S=\frac{W}{T_{P}}=\frac{W}{W_{S}+\frac{W-W_{S}}{p}} \leq \frac{W}{W_{S}} \forall p>0
$$

## Exercise 5.2

On a single processor, 11 arcs are traversed by FDS before the solution is found. On two processors fours arcs are traversed by $P_{1}$ before it finds the solution. In Figure 5.3 (page 201), the cost was on counting the states explored, here we count the arcs. The speedup is $11 / 4=2.75$. The anomaly is due to the fact that in Figure 5.10 (b) (page 228), the algorithm being executed is not the same as in Figure $5.10(\mathrm{a})$. The parallel algorithm performs less overall work than the serial algorithm. The algorithm in Figure $5.10(\mathrm{~b})$ is not a true DFS. If a single processor alternates between the nodes of the left and right sub-trees (which emulates the parallel algorithm), then the serial algorithm traverses only 8 arcs and the speedup becomes 2 .

## Exercise 5.3

As a reminder the degree of concurrency here corresponds the the maximal degree of concurrency in chapter 3 . Table gives the answers. The overhead function is computed by using $E=\frac{W}{p T_{P}} \Longrightarrow$

| Graph | (a) | (b) | (c) | (d) |
| :--- | :---: | :---: | :---: | :---: |
| Degree of concurrency (DC) | $2^{n-1}$ | $2^{n-1}$ | $n$ | $n$ |
| Longest path | $\log (n)$ | $\log (n)$ | $2 n-1$ | $n$ |
| Number of processors | $2^{n}-1$ | $2^{n}-1$ | $n^{2}$ | $\frac{n(n+1)}{2}$ |
| Maximum possible speedup | $\frac{2^{n}-1}{\log (n)}$ | $\frac{2^{n}-1}{\log (n)}$ | $\frac{n^{2}}{2 n-1}$ | $\frac{n+1}{2}$ |
| Longest path if $p=\mathrm{DC} / 2$ | $\log (n)+1$ | $\log (n)+1$ | $\frac{n}{2}+\frac{n}{2}+n+n-2$ | $\frac{n}{2}+\frac{n}{2}+\frac{n}{2}$ |
| Speedup for $p=\mathrm{DC} / 2$ | $\frac{2^{n}-1}{\log (n)+1}$ | $\frac{2^{n}-1}{\log (n)+1}$ | $\frac{n^{2}}{3 n-2}$ | $\frac{n+1}{3}$ |
| Efficiency for $p=\mathrm{DC} / 2$ | $\frac{\log (n)+1}{\log (n)+1}$ | $\frac{1}{3 n-2}$ | $\frac{2}{3 n}$ |  |
| Overhead $\left(T_{0}\right)$ for $p=\mathrm{DC} / 2$ | $\left(2^{n}-1\right) \log (n)$ | $\left(2^{n}-1\right) \log (n)$ | $3 n^{2}(n-1)$ | $\frac{n(n+1)(3 n-2)}{4}$ |

Table 1: Characteristics of graphs of Figure 5.11 (page 229).
$p T_{P}=\frac{W}{E}$ in $T_{0}=p T_{p}-W=W\left(\frac{1}{E}-1\right)$.

## Exercise 5.9

We use the expression $T_{P}=n / p-1+11 \log (p)$. We have the constraint

$$
512 \geq \frac{n}{p}-1+11 \log (p) \Longrightarrow(512-11 \log (p)) p \geq n
$$

that gives us the largest $n$ corresponding to a $p$ in Table. In general it is not possible to solve an arbitrarily large problem with a fixed amount of time provided an unlimited number of processors. For any parallel system with an isoefficiency function greater than $\Theta(p)$, a plot between $p$ and

| $p$ | 1 | 4 | 16 | 64 | 256 | 1024 | 4096 |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\log (p)$ | 0 | 2 | 4 | 6 | 8 | 10 | 12 |
| $n$ | 513 | 1964 | 7504 | 28608 | 108800 | 412672 | 1560576 |

Table 2: Largest $n$ for varying $p$.
the size of the largest problem that can be solved in a given time using $p$ processors will reach a maximum. It can be shown that for cost-optimal algorithms, the problem size can be increased linearly with the number of processors while maintaining a fixed execution time iff the isoefficiency fjnction is $\Theta(p)$.

