

PERFORMANCE ANALYSIS

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QUINN) accurately predict the performance of a parallel algorithm to decide whether it is worth coding/debugging it.

analyze execution time of a parallel program. (i) helps understand the barriers to higher performance

(2) predict the improvement due to increased #proc's.

AIMS

general formula for the speedup achievable by a parallel program

performance prediction formulas:

(1) Amdahl's Law:

(helps decide whether a ^{serial} program merits parallelization)

(2) Gustafson-Barsis Law:

(evaluate the performance of a parallel program)

(3) Karp-Flatt metric:

(helps decide whether the principal barrier to speedup is the amount of inherently sequential code, or parallel overhead)

(4) Isoefficiency metrics:

(evaluate the scalability of a parallel algorithm executed on a parallel computer)

SPEEDUP & EFFICIENCY

(2)

$$\text{Speedup} = \frac{\text{serial exec. time}}{\text{parallel exec. time}} \quad // \quad \text{RATIO}$$

ops performed by a parallel alg. fall into 3 categories:

- (1) serial computations
- (2) parallel " "
- (3) inter-proc. com/tion, parallel overhead

using these 3 categories we can produce a simple model for speed-up.

$\psi(n, p)$ speedup achieved solving a pb of size n on p proc^s.

$\sigma(n)$ inherently sequential (serial) portion of the computation

$\rho(n)$ portion of the computation that can be executed in parallel

$\kappa(n, p)$ time required for parallel overhead.

(3)

Serial program on 1 proc. requires time $\sigma(n) + p(n)$ to execute the computations.
(no inter-proc comm/comm required)

consider best possible parallel exec. time: (on p proc)

{ serial portion takes $\sigma(n)$ time
parallel portion takes $p(n)/p$ time (IDEAL)
+ $k(n, p)$ time for inter-proc comm/comm

ideal assumption: if parallel exec. time is larger, then speedup is smaller.

IN SUMMARY

$$\psi(n, p) \leq \frac{\sigma(n) + p(n)}{\sigma(n) + \frac{p(n)}{p} + k(n, p)} \quad \left| \right| \text{general speedup formula}$$

adding procs reduces completion of parallel portion, but increases completion time.

at some point the completion time increase is larger than the comp. decrease.
so denom $\uparrow \rightarrow \psi(n, p) \downarrow$

- computation component decreases as $\# \text{proc} \uparrow$
 - communication component increases as $\# \text{proc} \uparrow$
- FOR A FIXED PB SIZE, THERE IS AN OPTIMAL $\# \text{proc}$ MINIMIZING OVERALL

$$= \frac{\text{speedup}}{p}$$

(4)

$$\text{efficiency} = \frac{\text{serial exec. time}}{(\# \text{procs}) \times \text{parallel exec. time}}$$

(of a || program)

It is a measure of processor utilization

efficiency
of a || code
solving pb
of size n
on p proc

$$\varepsilon(n, p) \leq \frac{\sigma(n) + \rho(n)}{p[\sigma(n) + \kappa(n, p)] + \rho(n)}$$

property: $0 \leq \varepsilon(n, p) \leq 1$

AMDAHL'S LAW

since $\kappa(n, p) > 0$ we can discard it

$$\psi(n, p) \leq \frac{\sigma(n) + \rho(n)}{\sigma(n) + \rho(n)/p}$$

denote by f the inherently sequential portion of the computation

$$f = \frac{\sigma(n)}{\sigma(n) + \rho(n)} \quad \text{i.e. } \rho(n) = \sigma(n) \left(\frac{1}{f} - 1 \right)$$

$$\begin{aligned} \text{THEN } \psi(n, p) &\leq \frac{\sigma(n) / f}{\sigma(n) + \sigma(n) \left(\frac{1}{f} - 1 \right) / p} \\ &= \frac{1}{f + (1-f)/p} \end{aligned}$$

(5)

Amdahl's Law

if f is the fraction of ops in a computation that must be performed sequentially ($0 \leq f \leq 1$)

then the max speedup ψ achievable by a parallel computer with p procs

$$\text{is } \psi \leq \frac{1}{f + (1-f)/p}$$

\sim upper bound on the speedup achievable when using p procs

\sim determination of asymptotic speedup (as #procs increases)

ex1 10% \sim serial max speedup
90% \sim parallel on 8 procs?

$$\psi(n, 8) \leq \frac{1}{0.1 + (1-0.1)/8} \approx 4.7$$

ex2 25% \sim ~~serial~~ ^{possible} max speedup?

$$\psi(n, p) \leq \frac{1}{0.25 + (1-0.25)/p}$$

$$p \rightarrow \infty \\ \sim 4$$

⚠ Amdahl's law neglects parallel overhead.

Amdahl effect

(6)

typically $K(n, p)$ has smaller complexity than $f(n)$

e.g. $\Theta(n \log n)$, $\Theta(n^2)$

$f(x)$ is $\Theta(g(x))$ if $\exists x_0, M, N$ s.t.
 $\forall x > x_0, N/g(x) < |f(x)| < M/g(x)$
 $\sim \rightarrow g$ bounds f above and below for sufficiently large x
 \rightarrow gives tighter bounds on runtime than O -notation

conseq. increasing size of pb, increases the ~~computation~~ ^{computation} time ~~slower~~ ^{faster} than it increases the ~~computation~~ ^{computation} time

Amdahl effect $\sim \rightarrow$ for a fixed # procs speedup is usually an increasing fct of the pb. size.

hypothetical pb

