

CS301 Project Report

Project Title: Sample Sort

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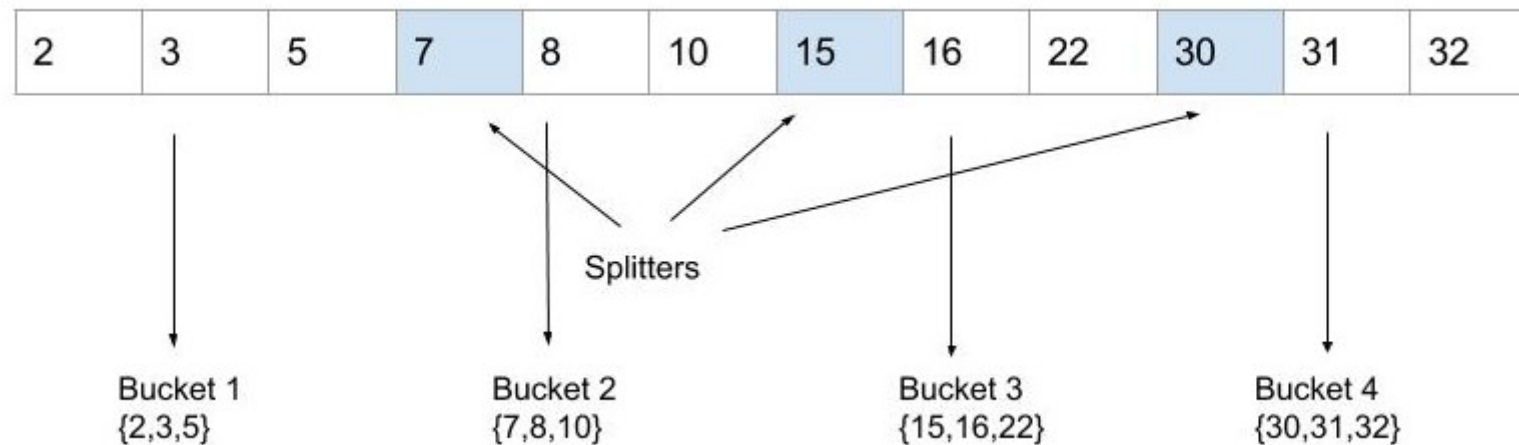
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Introduction

- Sample sort is a parallel sorting algorithm.
- Sample sort is a comparison based sorting algorithm.
- Sample sort follows PSRS algorithm design strategy. [Parallel Sorting by Regular Sampling] [explained in upcoming slides.]

Definition of Splitters

- Splitters: Splitters are the elements which divides sorted array into small blocs[chunks or buckets].
- Example:
- Here, a sorted array of 12 elements is splitted in 4 blocs [chunks or buckets] by 3 splitters.



PSRS algorithm design strategy

- This algorithm design consists of 5 stages:
 1. Local sorting
 2. Generation of local splitters by using a sampling algorithm
 3. Generation of global splitters
 4. Distributing buckets to nodes
 5. Local sorting of a bucket

Sample sort algorithm

1. Local sorting (Quicksort) .
2. Generation of local splitters. (each node selects $(p-1)$ splitters) .
3. Generation of global splitters. (root node gathers $p*(p-1)$ splitters and sorts them using quicksort and selects $(p-1)$ splitters from total $p*(p-1)$ splitters. They are transmitted to each node.)
4. Distributing buckets to nodes. (buckets are prepared by using new $(p-1)$ splitters and send to respective node. Bucket[i] goes node with rank=i. After that each node merges the buckets received.)
5. Local sorting of a bucket (Quicksort).

Algorithm example

p0								p1								p2							
32	7	13	18	2	27	10	14	20	6	10	34	15	19	21	5	16	19	33	4	21	12	5	8

2	7	10	13	14	18	27	32	5	6	10	15	19	20	21	34	4	5	8	12	16	19	21	33
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Stage 1
Local sorting

10	18	10	20	8	19
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Stage 2
Local splitters

8	10	10	18	19	20
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Stage 3
Global splitter selected at root node.

2	7	10	13	14	18	27	32	5	6	10	15	19	20	21	34	4	5	8	12	16	19	21	33
2	7	5	6	4	5	8	10	13	14	18	10	15	12	16	27	32	19	20	21	34	19	21	33

Stage 4
1. Buckets are prepared.
2. Buckets are transmitted to respective node.

p0								p1								p2							
2	4	5	5	6	7	8	10	10	12	13	14	15	16	18	19	19	20	21	21	27	32	33	34

Stage 5
Local Sorting

Theoretical analysis of parallel algorithm

Note: We have assumed that communication latency is $O(\log(p))$.

- Step 1: $O((n/p) * \log(n/p))$ [Sorting]
- Step 2: $O(p)$ [Choosing local splitters]
- Step 3: $O(p * \log(p))$ [Collecting local splitters] + $O((p^2 * \log(p)))$ [Sorting local splitters] + $O(p)$ [Choosing global splitters] + $O(p * \log(p))$ [Distributing global splitters]
- Step 4: $O(n/p)$ [Buckets generation] + $O((n/p) * \log(p))$ [Buckets transmission] + $O((n/p) * \log(p))$ [Merging buckets]
- Step 5: $O((n/p) * \log(n/p))$ [Sorting]

Theoretical analysis of parallel algorithm

- Computational complexity: $O((n/p) * \log(n/p)) + O(p^2 * \log(p)) + O((n/p) * \log(p))$
- Since $n \gg p$, $\log(n/p) \sim \log(n)$, Overall computational complexity:
 $O((n/p) * (\log(n) + \log(p)))$.
- Communication complexity: $O(p * \log(p)) + O((n/p) * \log(p))$
- Since $(n/p) \gg \log(p)$ and $(n/p) \gg p$, Overall communication complexity:
 $O(n/p)$.
- Total bound: $O((n/p) * (\log(n) + \log(p))) + O(n/p)$.
- We compare this parallel algorithm to serial randomized quicksort algorithm.
[most widely used comparison based sorting algorithm].

Theoretical analysis of serial algorithm and expected speedup

- Total bound (serial code) = $O(n \cdot \log(n))$
- Theoretical speedup =
$$O(n \cdot \log(n)) / \{O((n/p) \cdot (\log(n) + \log(p))) + O(n/p)\}.$$
- Expected speedup (predication based approach):
- To calculate approximate speedup, we simply say that parallel algorithm actually sorts twice.[Local sorting]. Thus, on p processors, calculation is p times faster but 2 times slower. Thus, predicated speedup = $p/2$. [This speedup is just based on observation.]

Scalability

- From theoretical analysis of parallel algorithm:

$$k(n,p) = O(p \cdot \log(p)) + O((n/p) \cdot \log(p))$$

- We can ignore first term in $k(n,p)$.
- Thus, $k(n,p) = O((n/p) \cdot \log(p))$.
- Therefore $T_o(n,p) = O(n \cdot \log(p))$.
- $T(n,1) = O(n \cdot \log(n))$.
- Isoefficiency function:

$$T(n,1) \geq C \cdot T_o(n,p)$$

$$n \cdot \log(n) \geq C \cdot n \cdot \log(p)$$

$$\log(n) \geq C \cdot \log(p)$$

$$n \geq p^C \quad \dots \text{equation(1)}$$

Scalability

- Scalability function:
- Since $M(n) = n$, Scalability function $M(f(p))/p = p^{(C-1)}$.
- If $C=1$ then system will perfectly scalable.

$$C = (1 - \text{efficiency})/\text{efficiency} .$$

Suppose x = maximum allowed efficiency which results into perfectly scalable system.

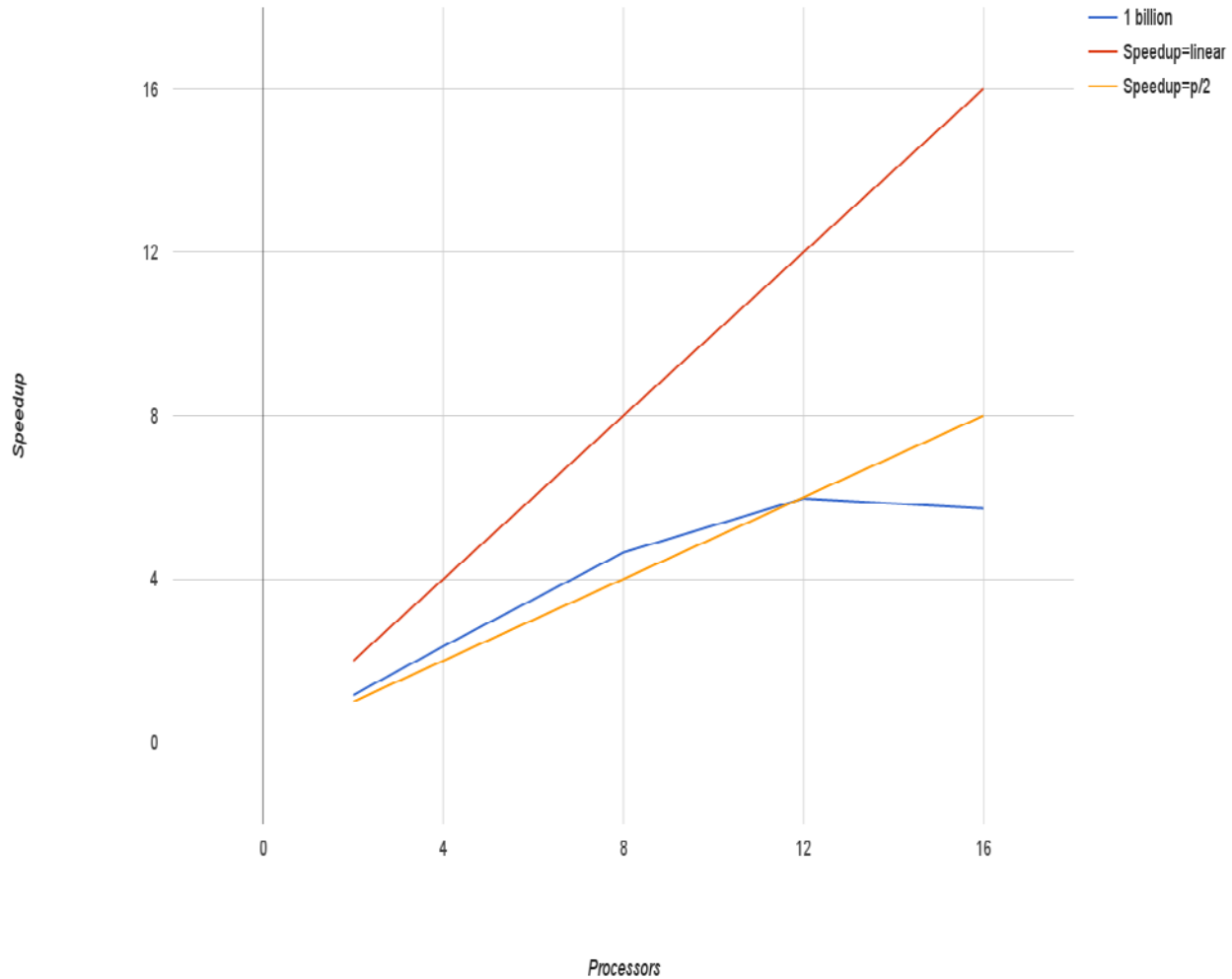
Then $C = (1-x)/x = 1$ (because system is perfectly scalable.)

This only happens when $x = \frac{1}{2}$.

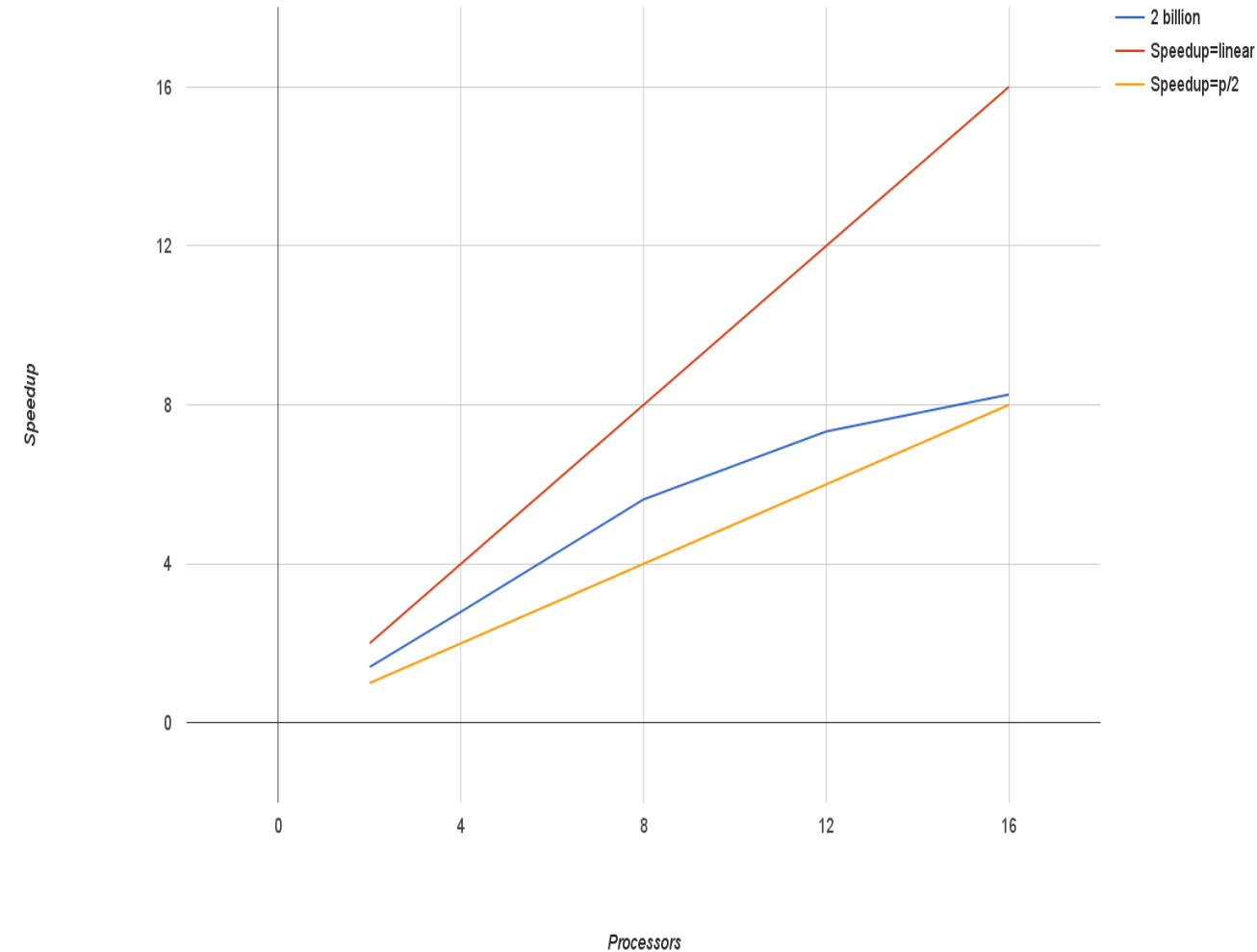
- Thus we can maintain efficiency of $1/2$ and speedup of $p/2$ by increasing n linearly with p . (i.e. put $C=1$ in equation(1) thus isoefficiency function becomes $n \geq p$).

Practical analysis: Amdahl's Law

Speedup vs No of cores for n = 1 billion

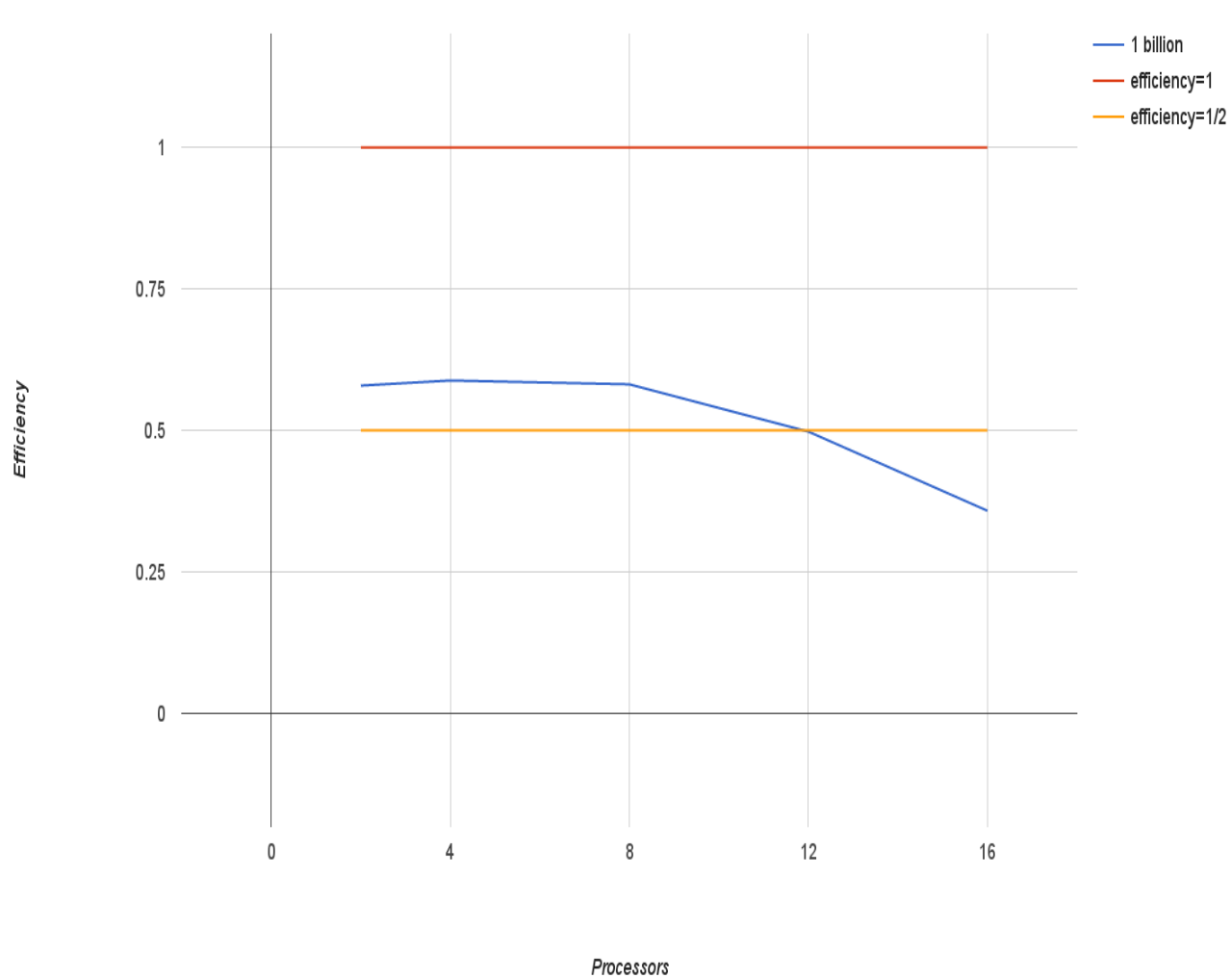


Speedup vs No of cores for n = 2 billion

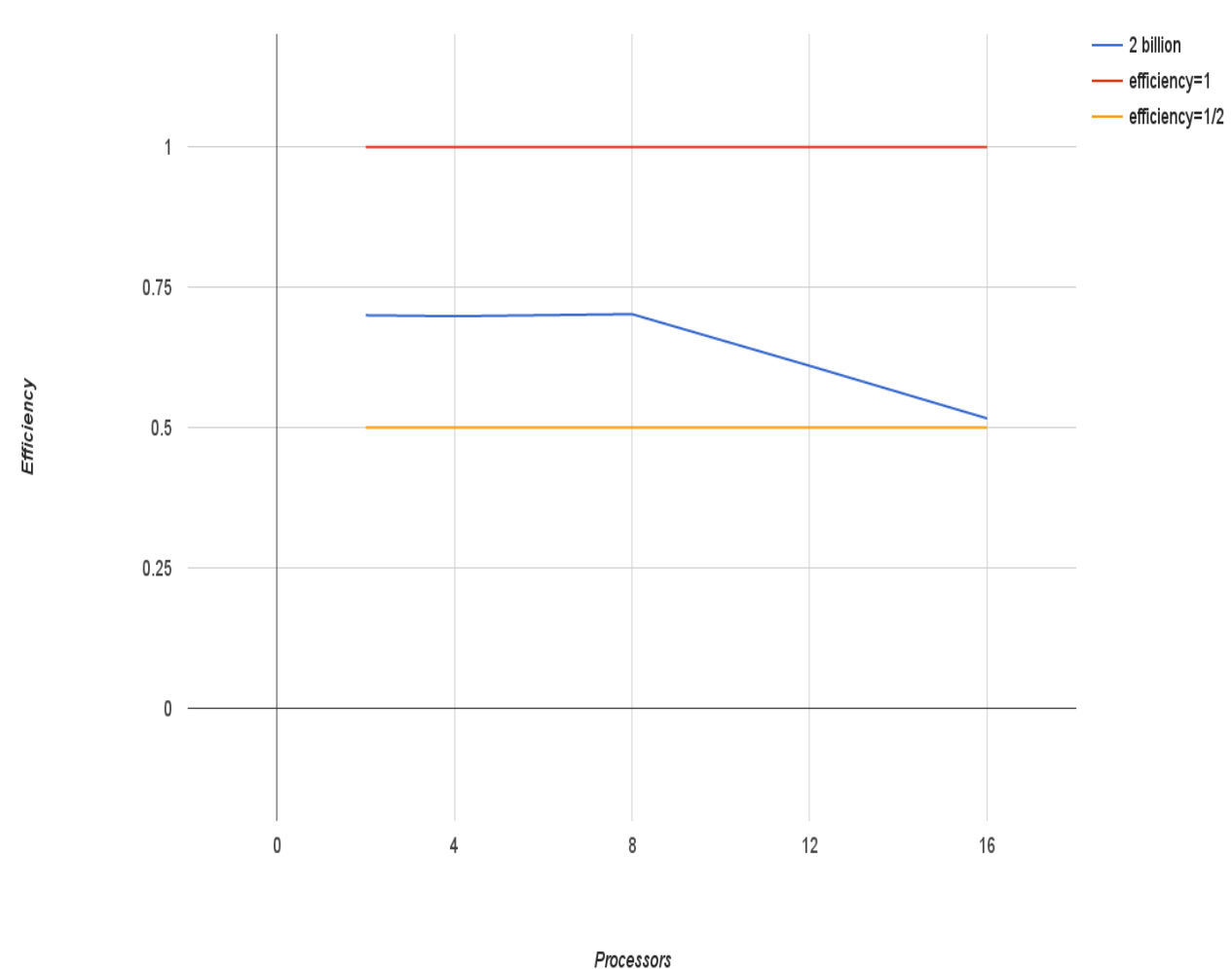


Practical analysis: Amdahl's Law

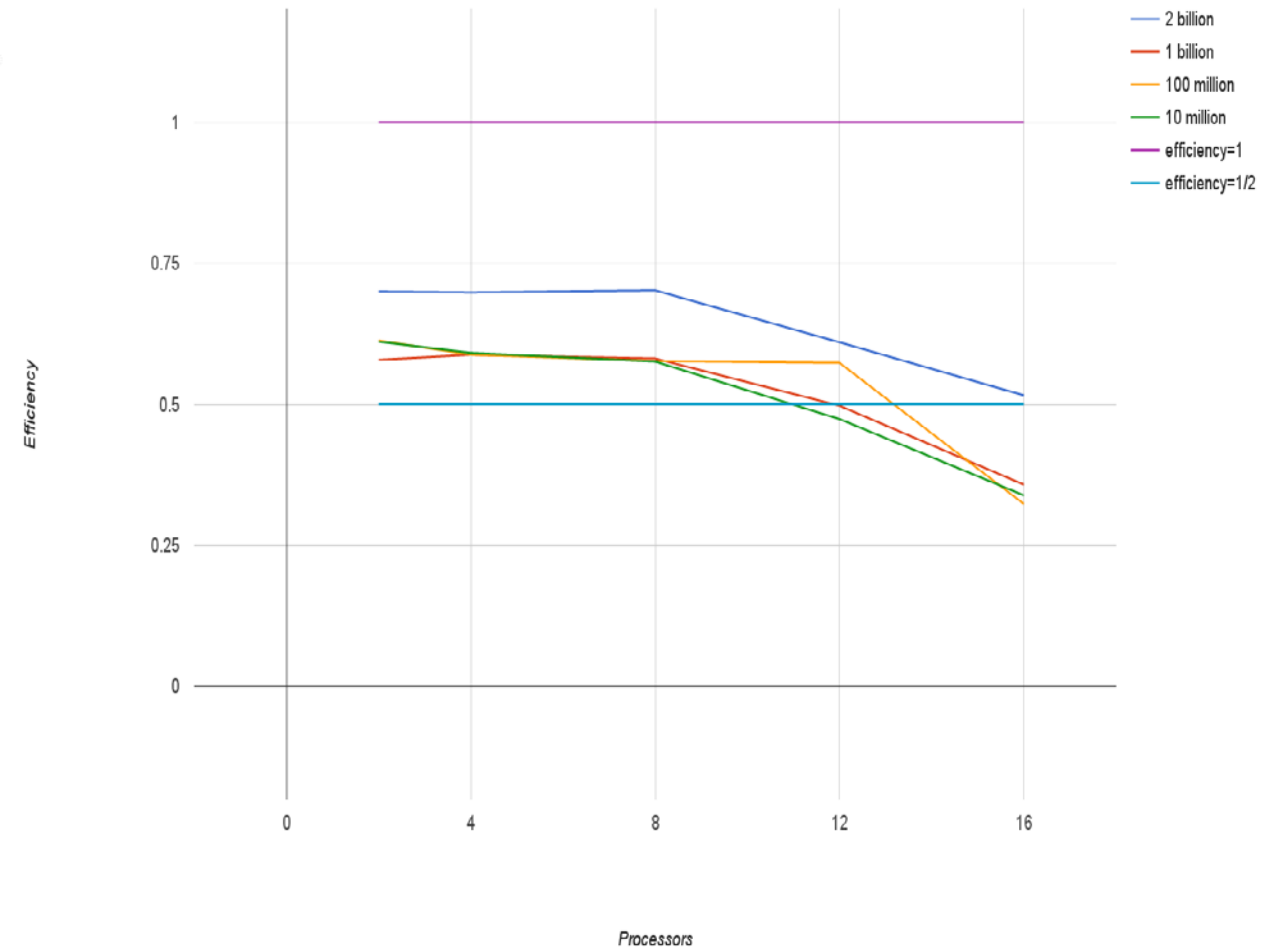
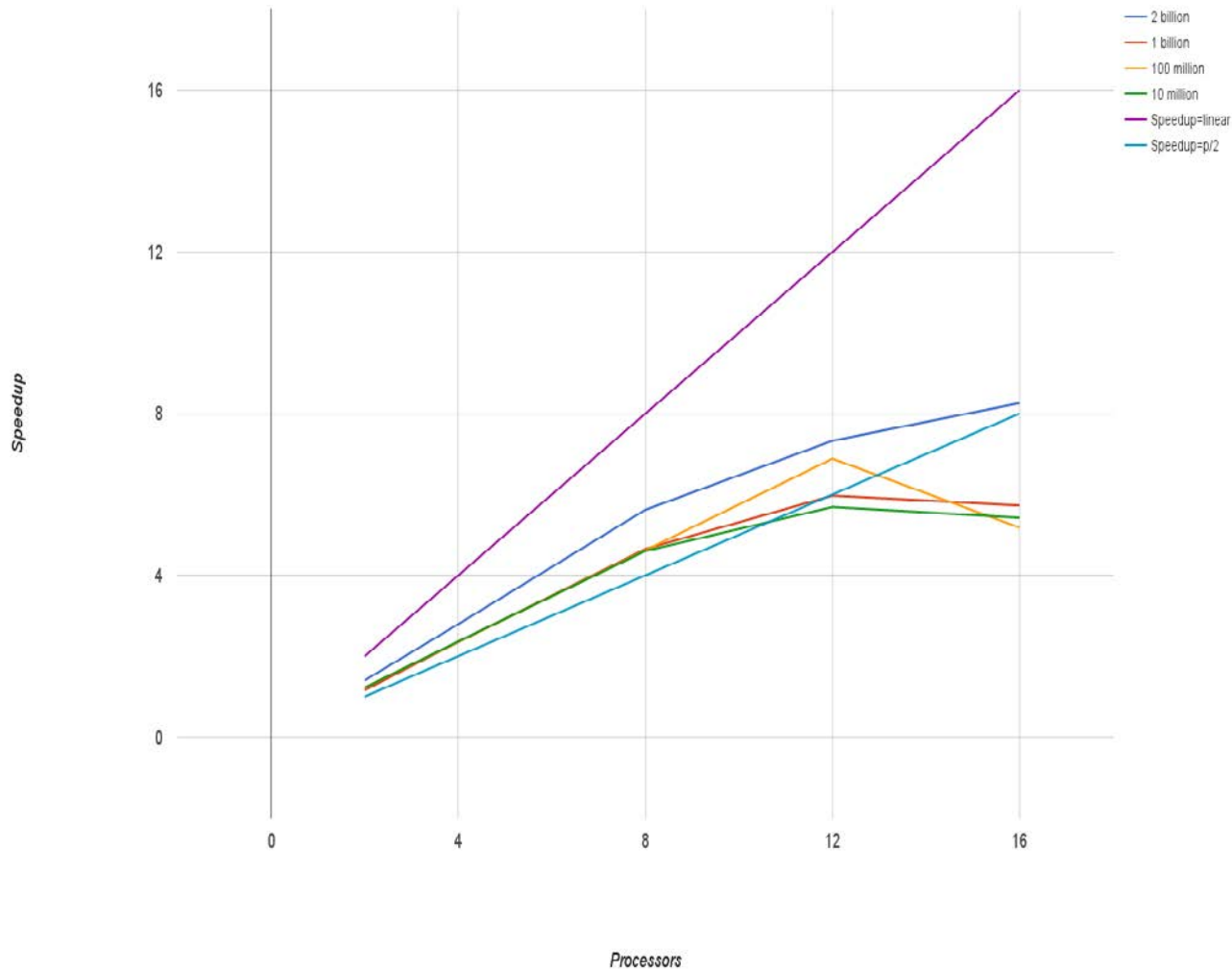
Efficiency vs No of cores for n = 1 billion



Efficiency vs No of cores for n = 2 billion



Practical analysis: Amdahl's Effect



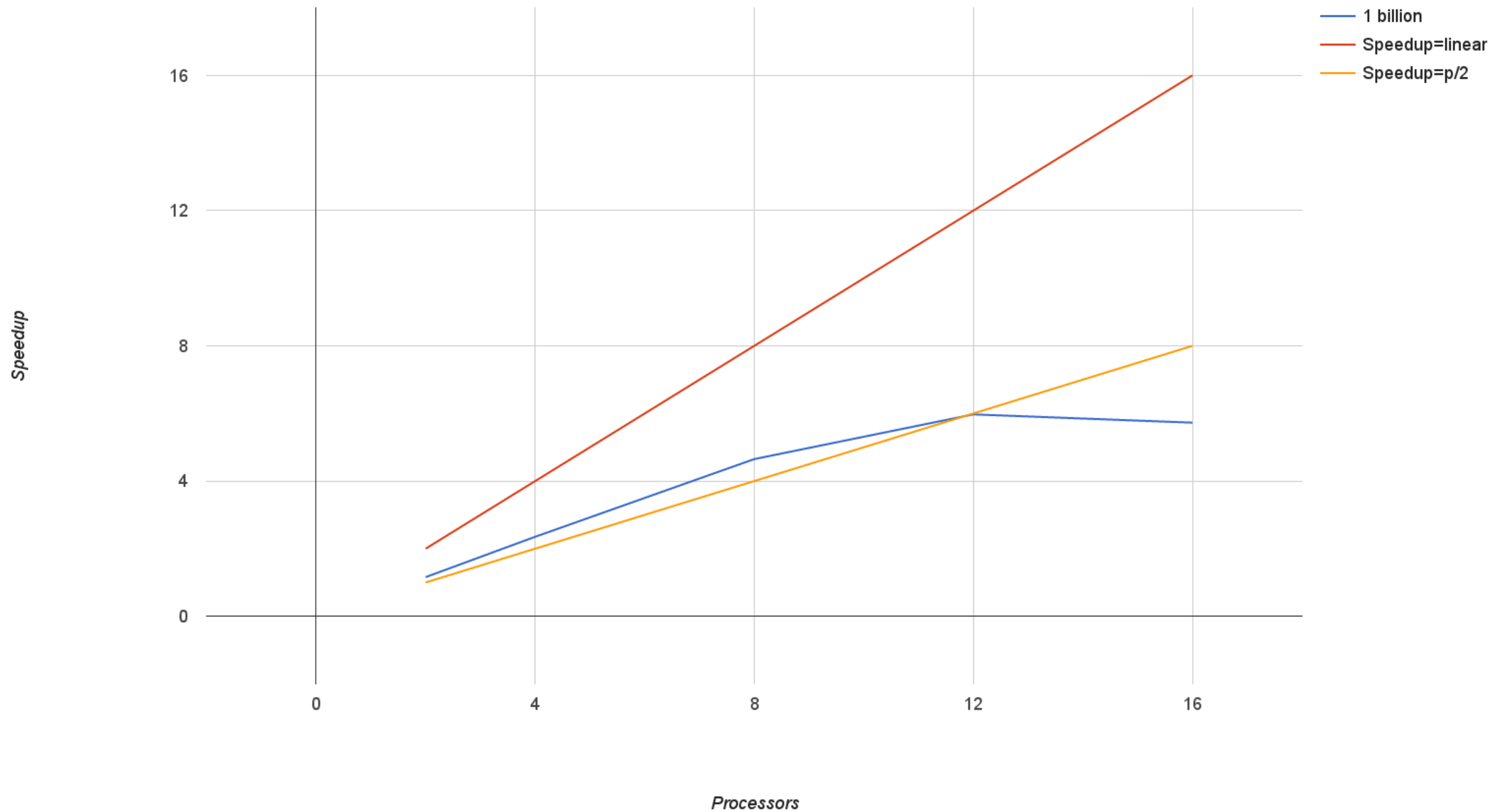
Improvements

- Based on design strategy of sample sort, several new sorting algorithms have been developed. Most of them focus on 2 things:
- (i) Better Sampling algorithm
- (ii) Decrease communication time
 - a) By using shared memory programming model
 - b) By improving communication patterns

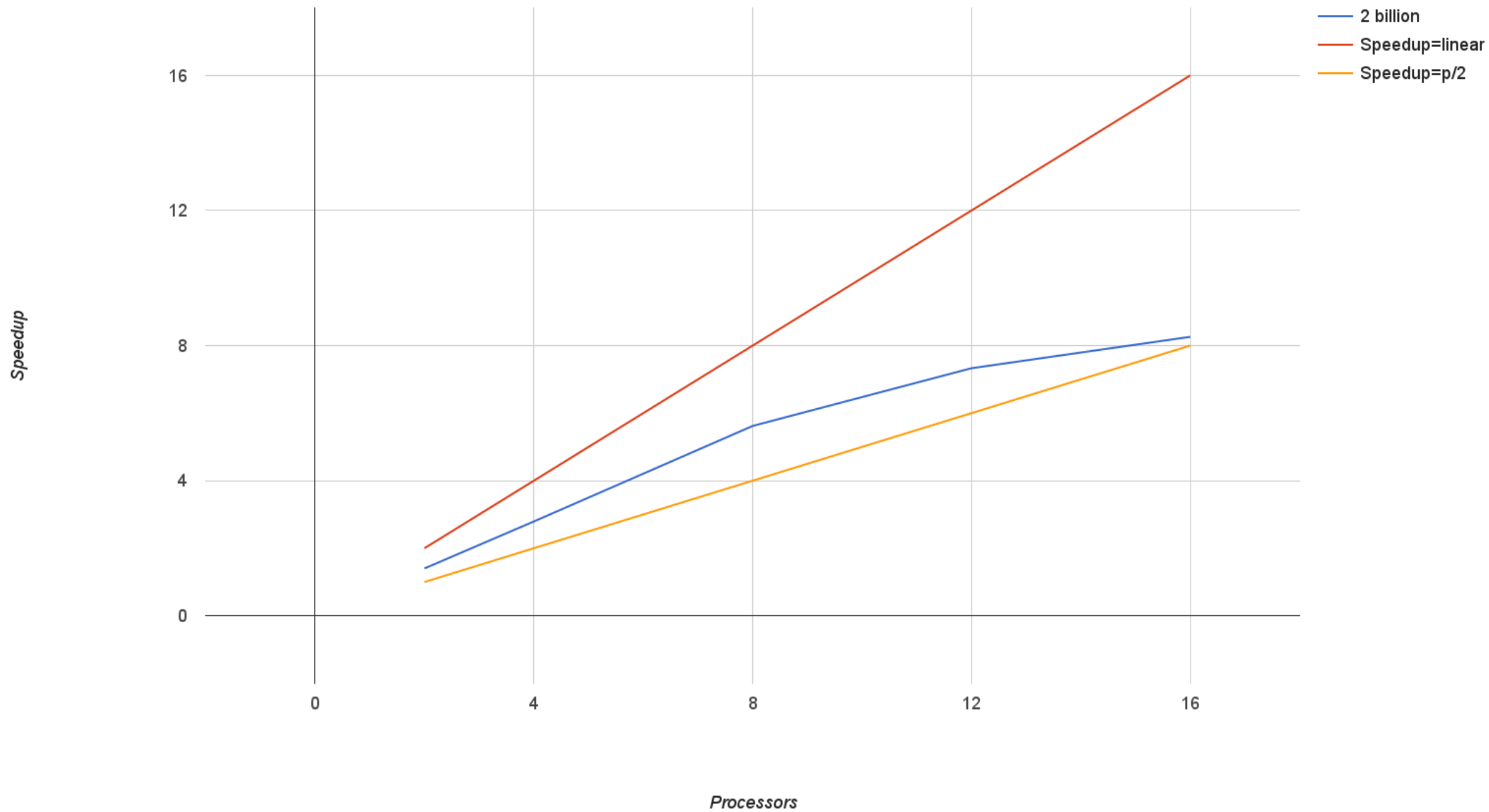
Optional Slides

Below slides have same graphs as shown in slide number 12,13,14. If they are not clearly visible in above slides , you can find them in upcoming slides.

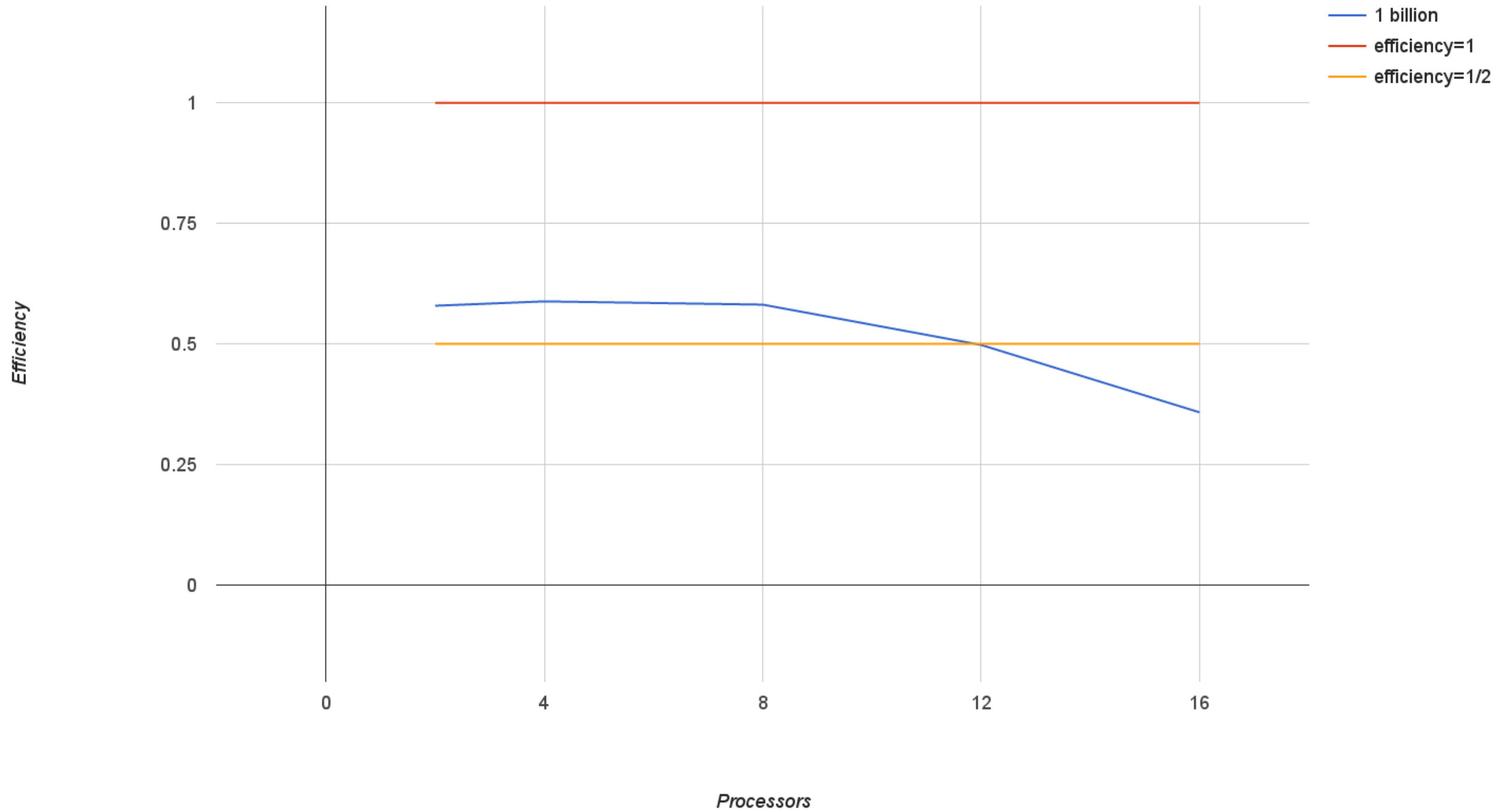
Speedup vs No of cores for n = 1 billion



Speedup vs No of cores for n = 2 billion



Efficiency vs No of cores for n = 1 billion



Efficiency vs No of cores for n = 2 billion

