

Phase Transition in the Number Partitioning Problem

Leena Salmela

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Outline

The Number Partitioning Problem (NPP)

Algorithms for NPP

Phase Transition in NPP

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Number Partitioning Problem (NPP)

- ▶ Input: a list a_1, a_2, \dots, a_N of positive integers
- ▶ Partition: a subset $A \subset \{1, \dots, N\}$
- ▶ Discrepancy of partition A :

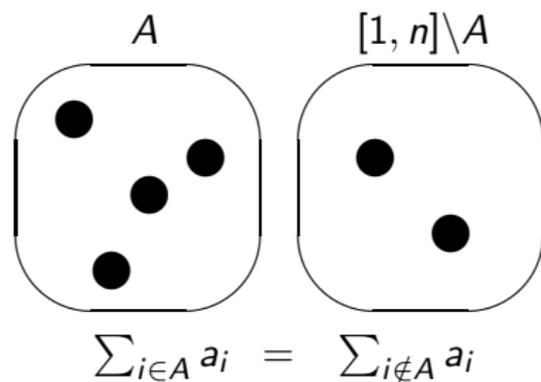
$$E(A) = \left| \sum_{i \in A} a_i - \sum_{i \notin A} a_i \right|$$

- ▶ Minimize the discrepancy

NPP: Example

Multi-processor scheduling:

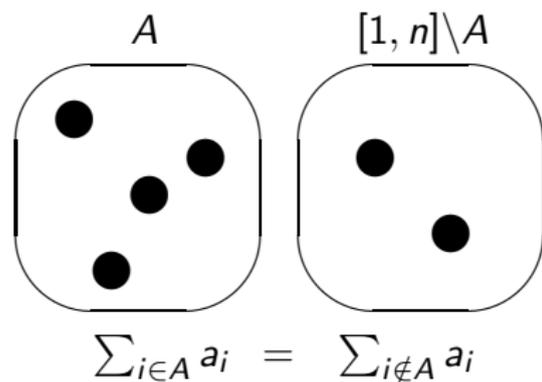
- ▶ 2 processors
- ▶ 5 tasks with runtimes 8, 7, 6, 5, 4
- ▶ How to divide the jobs among the processors so that the tasks are completed as fast as possible?



NPP: Example

Multi-processor scheduling:

- ▶ 2 processors
- ▶ 5 tasks with runtimes 8, 7, 6, 5, 4
- ▶ How to divide the jobs among the processors so that the tasks are completed as fast as possible?
 - ▶ Processor 1 runs tasks with runtimes: 8, 7
 - ▶ Processor 2 runs tasks with runtimes: 6, 5, 4
 - ▶ Discrepancy: $|8 + 7 - 6 - 5 - 4| = 0$

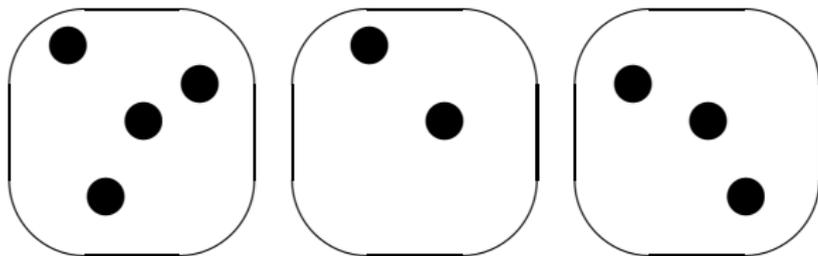


Perfect Partitions

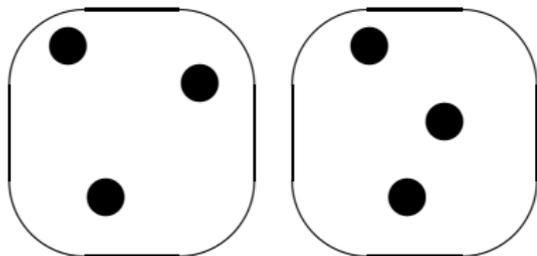
- ▶ Perfect partition:
 - ▶ Discrepancy $E(A) = 0$ if $\sum a_i$ even
 - ▶ Discrepancy $E(A) = 1$ if $\sum a_i$ odd
- ▶ The list 1, 2, 3, 4, 5, 6 has a perfect partition $A = \{2, 3, 5\}$ because $E(A) = |2 + 3 + 5 - 1 - 4 - 6| = 1$.
- ▶ Not all lists have perfect partitions.

Variations of NPP

- ▶ Input can be a list of real numbers from the interval $[0, 1]$
- ▶ Multiway NPP: Partitioning to more than 2 sets



- ▶ Balanced NPP: The sets must have the same cardinality



Properties of NPP

- ▶ NP-complete
- ▶ Poor quality of heuristic algorithms
 - ▶ Average discrepancy of optimum partitions $O(\sqrt{N} \cdot 2^{-N})$ (for real-valued NPP)
 - ▶ Best heuristic algorithm yields average discrepancies $O(N^{-\alpha \log N})$ (for real-valued NPP)
- ▶ The phase transition can be analyzed rigorously analytically. (*S. Mertens: Phase Transition in the Number Partition Problem, Phys. Rev. Lett.* **81**, 4281 (1998))

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Phase Transition in NPP

The Greedy Algorithm

- ▶ Assign the largest unassigned number to the set with the smallest sum
 - ▶ Iterate until all numbers are assigned
-
- ▶ Keeps the discrepancy small with every decision
 - ▶ Yields average discrepancies $O(1/N)$ (for real-valued NPP)
(*S. Mertens: The Easiest Hard Problem: Number Partitioning, in: A.G. Percus, G. Istrate, C. Moore, eds., Computational Complexity and Statistical Physics, Oxford University Press (2006), p. 125-139*)
 - ▶ Time complexity $O(N \log N)$ (sorting)

The Greedy Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

The Greedy Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: 8 assigned to set 1
 $\{8\}\{\}$

The Greedy Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: 8 assigned to set 1
 $\{8\}\{\}$

Iter 2: 7 assigned to set 2
 $\{8\}\{7\}$

The Greedy Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: 8 assigned to set 1
 $\{8\}\{\}$

Iter 2: 7 assigned to set 2
 $\{8\}\{7\}$

Iter 3: 6 assigned to set 2
 $\{8\}\{7, 6\}$

The Greedy Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: 8 assigned to set 1
 $\{8\}\{\}$

Iter 2: 7 assigned to set 2
 $\{8\}\{7\}$

Iter 3: 6 assigned to set 2
 $\{8\}\{7, 6\}$

Iter 4: 5 assigned to set 1
 $\{8, 5\}\{7, 6\}$

The Greedy Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: 8 assigned to set 1
 $\{8\}\{\}$

Iter 2: 7 assigned to set 2
 $\{8\}\{7\}$

Iter 3: 6 assigned to set 2
 $\{8\}\{7, 6\}$

Iter 4: 5 assigned to set 1
 $\{8, 5\}\{7, 6\}$

Iter 5: 4 assigned to set 1
 $\{8, 5, 4\}\{7, 6\}$

Discrepancy: 4

The Differencing Algorithm by Karmarkar and Karp

- ▶ Key idea: Reduce the length of the number list
 - ▶ Replace two largest numbers a_i and a_j by $|a_i - a_j|$
 - ▶ Commit to placing the two largest numbers in different sets
 - ▶ Repeat until only one number left
 - ▶ The remaining number = discrepancy
-
- ▶ Achieves better results than the greedy algorithm
 - ▶ Yields average discrepancies $O(N^{-\alpha \log N})$ with $\alpha = 0.72$ (for real-valued NPP)
 - ▶ Time complexity $O(N \log N)$ (sorting)

The Differencing Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

The Differencing Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: Replace 8 and 7 by $|8 - 7| = 1$
 $\{6, 5, 4, 1\}$

The Differencing Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: Replace 8 and 7 by $|8 - 7| = 1$
 $\{6, 5, 4, 1\}$

Iter 2: Replace 6 and 5 by $|6 - 5| = 1$
 $\{4, 1, 1\}$

The Differencing Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: Replace 8 and 7 by $|8 - 7| = 1$
 $\{6, 5, 4, 1\}$

Iter 2: Replace 6 and 5 by $|6 - 5| = 1$
 $\{4, 1, 1\}$

Iter 3: Replace 4 and 1 by $|4 - 1| = 3$
 $\{3, 1\}$

The Differencing Algorithm: An Example

Input: $\{8, 7, 6, 5, 4\}$

Iter 1: Replace 8 and 7 by $|8 - 7| = 1$
 $\{6, 5, 4, 1\}$

Iter 2: Replace 6 and 5 by $|6 - 5| = 1$
 $\{4, 1, 1\}$

Iter 3: Replace 4 and 1 by $|4 - 1| = 3$
 $\{3, 1\}$

Iter 4: Replace 3 and 1 by $|3 - 1| = 2$
 $\{2\}$

Discrepancy: 2

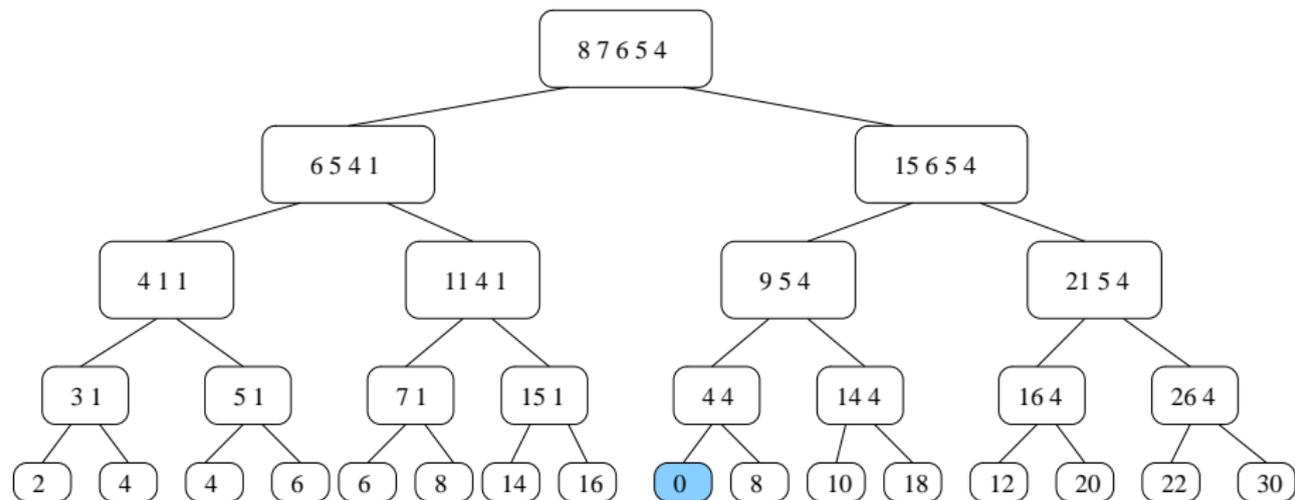
The Differencing Algorithm: An Example

Input:	$\{8, 7, 6, 5, 4\}$	
Iter 1:	Replace 8 and 7 by $ 8 - 7 = 1$ $\{6, 5, 4, 1\}$	$\{7, 5, 4\}\{8, 6\}$
Iter 2:	Replace 6 and 5 by $ 6 - 5 = 1$ $\{4, 1, 1\}$	$\{5, 4\}\{6, 1\}$
Iter 3:	Replace 4 and 1 by $ 4 - 1 = 3$ $\{3, 1\}$	$\{4\}\{1, 1\}$
Iter 4:	Replace 3 and 1 by $ 3 - 1 = 2$ $\{2\}$	$\{3\}\{1\}$
Discrepancy:	2	$\{2\}\{\}$

Complete Algorithms

- ▶ Complete greedy algorithm
 - ▶ First try to put the largest number to the set with smaller sum.
 - ▶ Then try to put it to the other set.
 - ▶ I.e. search a tree of all 2^N possible partitions
- ▶ Complete differencing algorithm
 - ▶ First try to put the two largest numbers a_i and a_j to different sets
 \implies Replace them by $|a_i - a_j|$
 - ▶ Then try to put them into the same set
 \implies Replace them by $a_i + a_j$

Complete Differencing Method: Example

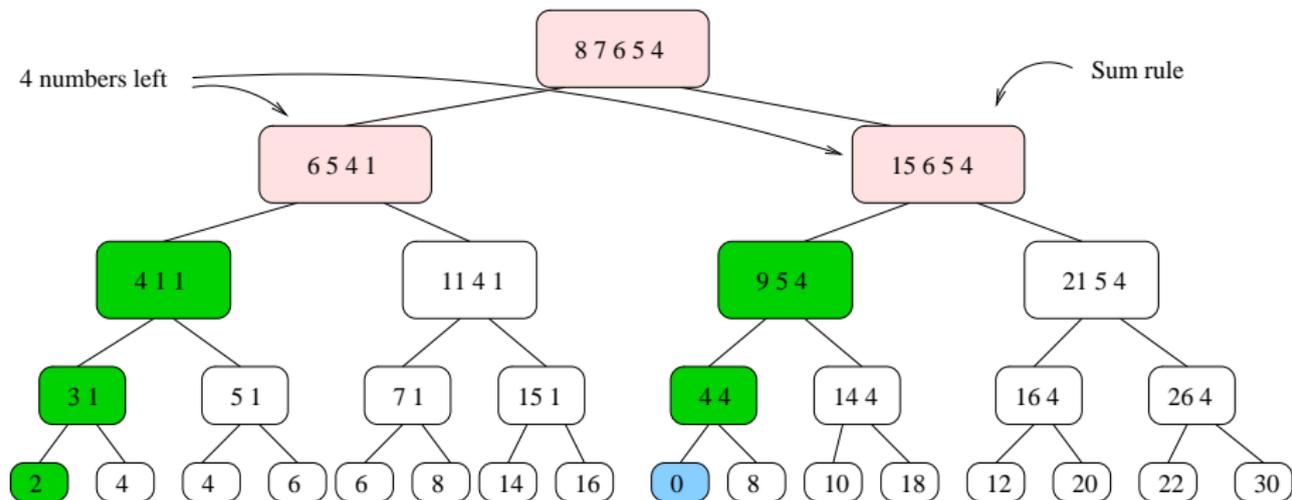


Complete Differencing Method with Pruning

Pruning rules for the complete differencing method:

- ▶ If less than 5 numbers left, take the left branch
- ▶ If largest number \geq sum of the rest of the numbers, place the largest number in one set and the others in the other set
- ▶ If a perfect partition has been found, stop.

Complete Differencing Method with Pruning: Example



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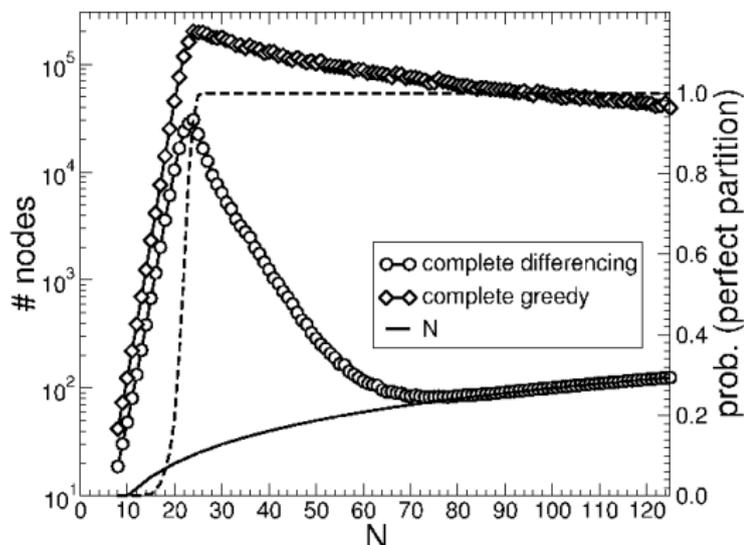
Phase Transition in NPP

Parameterization of NPP Instance Ensemble

- ▶ NPP instance: N random numbers a_i
- ▶ $a_i < A$
- ▶ $A = 2^{\kappa N} \implies$ worst case complexity of any (known) algorithm exponential in N for all $\kappa > 0$
- ▶ Typical complexity of NPP depends on κ

Phase Transition

- ▶ Instances: N random 20-bit integers
- ▶ $A = 2^{20} \implies \kappa = 20/N$
- ▶ Phase transition at $\kappa_c \approx 20/24 \approx 0.83$



(From: *S. Mertens: The Easiest Hard Problem: Number Partitioning*, in: *A.G. Percus, G. Istrate, C. Moore, eds., Computational Complexity and Statistical Physics*, Oxford University Press (2006), p. 125-139)

Pseudo Polynomiality of NPP

- ▶ $a_i < A$
- ▶ Discrepancy has at most NA different values
- ▶ The search space can be explored in time $O(N^2A)$
- ▶ A polynomial algorithm?

Pseudo Polynomiality of NPP

- ▶ $a_i < A$
- ▶ Discrepancy has at most NA different values
- ▶ The search space can be explored in time $O(N^2A)$
- ▶ A polynomial algorithm?
 - ▶ Input size $O(N \log A)$
 - ⇒ Running time is exponential in the size of the input!
- ▶ NP-hardness: input numbers allowed to be exponentially large in N
- ▶ If N is increased for constant A , NPP is eventually not NP-hard anymore

Analytical Results

(*S. Mertens: Phase Transition in the Number Partition Problem, Phys. Rev. Lett.* **81**, 4281 (1998))

- ▶ Analyze the average number of perfect partitions $\Omega(0)$
- ▶ It can be shown that

$$\log_2 \Omega(0) = N(\kappa_c - \kappa)$$

with

$$\kappa_c = 1 - \frac{\log_2 N}{2N} - \frac{1}{2N} \log_2 \left(\frac{\pi}{6} \right)$$

- ▶ $\kappa < \kappa_c \implies$ Exponential number of perfect partitions \implies Easy phase
- ▶ $\kappa > \kappa_c \implies$ No perfect partitions \implies Hard phase

Characteristics of the Easy Phase

- ▶ Exponential number of perfect partitions
- ▶ κ decreases \implies the number of perfect partitions increases
- ▶ For small κ perfect partitions found by heuristic algorithms

Characteristics of the Hard Phase

- ▶ Exponentially small probability of perfect partitions \implies optimum almost always unique
- ▶ Heuristic algorithms no better than blind search

Summary

- ▶ Number partitioning problem
- ▶ Two heuristic algorithms
 - ▶ Greedy
 - ▶ Differencing
- ▶ Phase transition in average complexity \iff
Phase transition in probability of perfect partitions
 - ▶ Easy phase where heuristic algorithms are effective
 - ▶ Hard phase where heuristic algorithms are no better than blind search

Analytical Results – Proof

(*S. Mertens: The Easiest Hard Problem: Number Partitioning*, in: *A.G. Percus, G. Istrate, C. Moore, eds., Computational Complexity and Statistical Physics*, Oxford University Press (2006), p. 125-139)

- ▶ Code a partition A with binary variables $s_i = \pm 1$:
 - ▶ $s_i = +1 \implies i \in A$ and $s_i = -1 \implies i \notin A$
- ▶ The cost function: $E = |D(s)|$ where

$$D(s) = \sum_{i=1}^N a_i s_i$$

- ▶ D can be interpreted as distance to the origin of a random walker in one dimension who takes steps to the right ($s_i = +1$) or left ($s_i = -1$) with random step sizes (a_i).
- ▶ Average number of walks ending at D :

$$\Omega(D) = \sum_{\{s_i\}} \left\langle \delta \left(D - \sum_{i=1}^N a_i s_i \right) \right\rangle$$

Analytical Results – Proof (continued)

- ▶ For a fixed walk $\{s_i\}$ and large N the distance to the origin is Gaussian with mean

$$\langle D \rangle = M \langle a \rangle$$

and variance

$$\langle D^2 \rangle - \langle D \rangle^2 = M^2 (\langle a^2 \rangle - \langle a \rangle^2)$$

where $M = \sum_j s_j$.

- ▶ Averaging over the random walk $\{s_i\}$:

$$[\langle D \rangle] = [M] \langle a \rangle = 0$$

$$[\langle D^2 \rangle] - [\langle D \rangle]^2 = [M^2] \langle a^2 \rangle = N \langle a^2 \rangle$$

- ▶ The probability of ending the walk at distance D

$$p(D) = \frac{1}{\sqrt{2\pi N \langle a^2 \rangle}} \exp\left(-\frac{D^2}{2N \langle a^2 \rangle}\right)$$

Analytical Results – Proof (continued)

- ▶ Note: The walk ends at even (odd) numbers for $\sum a_i$ even (odd)
- ▶ Average number of walks ending at D

$$\Omega(D) = 2^N 2p(D) = \frac{2^{N+1}}{\sqrt{2\pi N \langle a^2 \rangle}} \exp\left(-\frac{D^2}{2N \langle a^2 \rangle}\right)$$

- ▶ Substituting

$$\langle a^2 \rangle = \frac{1}{3} 2^{2\kappa N} (1 - O(2^{-\kappa N}))$$

we get

$$\log_2 \Omega(0) = N(\kappa_c - \kappa)$$

with

$$\kappa_c = 1 - \frac{\log_2 N}{2N} - \frac{1}{2N} \log_2 \left(\frac{\pi}{6}\right)$$