

# TECHNIQUES FOR DIMENSIONING AGGREGATED EVENT STREAMS TO SUSTAIN GLOBAL APPLICATION'S DEMANDS

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Dimensioning  
Event Streams

Di Gregorio

Motivation

Problem  
Example

Event Streams

Definition  
Modeling  
Bounds over time

Techniques

Empirical  
Analytical

## DISCLAIMER

All opinions expressed here are those of the author individually and are not reflective or indicative of the opinions and positions of the author's employer.

## 1 OVERVIEW AND MOTIVATION

- The problem of sustaining the performance demand
- An introductory case study

## 2 THE ACHIEVABLE PERFORMANCE OF EVENT STREAMS

- What are event streams
- Modeling event streams using linear inequalities
- Time and variability in event streams

## 3 EMPIRICAL AND ANALYTICAL TECHNIQUES

- Empirical determination of network calculus bounds
- Analytical determination of sustained properties

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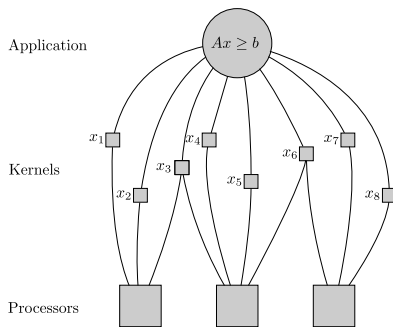
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## PROBLEM

What level of performance is demanded on every processing element, so that they **jointly** satisfy the global demand?

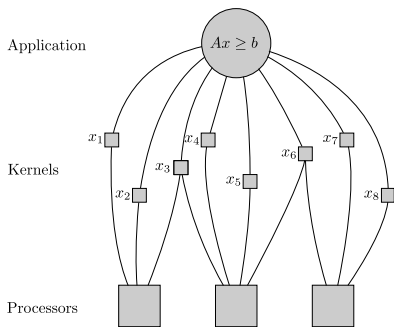
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- Demand on kernels for **aggregated streams** of computation events.
  - aggregated = many contributors to one performer.
- Demand on **processing elements** for performance in executing kernels.



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It is a standard industry practice to solve such problems

- by incremental steps and gut feeling,
- by tabling scenarios on spreadsheets,
- by profiling and tweaking the design,
- sometimes, by developing abstract architecture models.

using mathematics is usually considered not feasible for this **complexity** and too cumbersome.

### HOW COMPLEX IS “TOO COMPLEX”?

Some quite generic problems are not as complex as they appear.

- **Model** many problems by parametric linear inequalities.
- **Solve** the “architectural part” of the problem only once.
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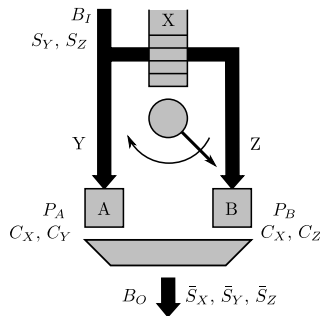
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## QUESTION

Under sufficient load, how many instructions per unit time,  $P_A$  and  $P_B$ , are demanded from A and B to **guarantee** that the bandwidth  $B_O$  of the output channel is exhausted?

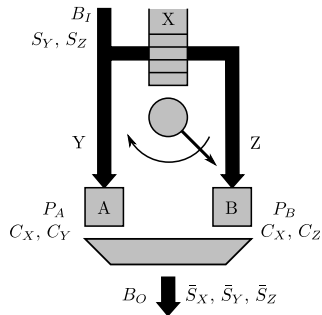
An input channel delivers frame streams Y and Z to two cores A and B, which sustain a fairly distributed traffic flow X from a queue of packets. All the processed streams X, Y and Z are multiplexed into a common output channel.



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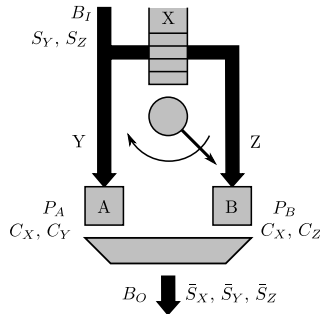
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- Processing a frame of Y costs **up to**  $C_Y$  **instructions** and changes its **maximum size** from  $S_Y$  to  $\bar{S}_Y$ .
- Similarly a frame of Z costs  $C_Z$  and changes from  $S_Z$  to  $\bar{S}_Z$ , a packet of X costs  $C_X$  and changes from  $S_X$  to  $\bar{S}_X$
- The input channel **guarantees** a bandwidth  $B_I$ .

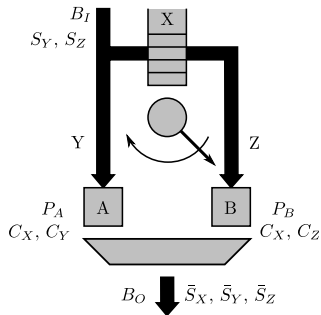


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 \bar{S}_X x_X + \bar{S}_Y x_Y + \bar{S}_Z x_Z &\geq B_O
 \end{aligned}$$

$x_Y$  and  $x_Z$  represent respectively the total frames per unit time of the streams Y and Z. Similarly,  $x_X$  represents the total packets per unit time of the flow X.



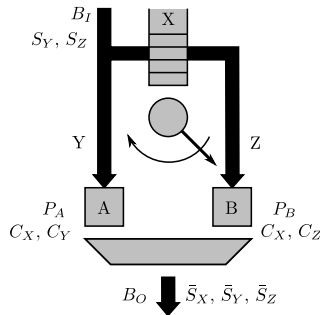
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## ANSWER

we must determine  $P_A$  and  $P_B$  such that the last inequality is satisfied for all  $x_X$ ,  $x_Y$  and  $x_Z$ .



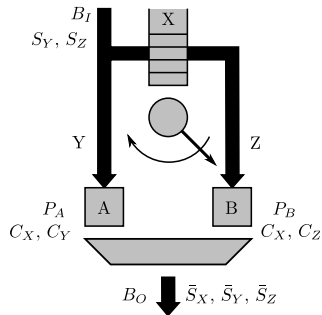
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## SOLUTION

The inequality  $\bar{S}_X x_X + \bar{S}_Y x_Y + \bar{S}_Z x_Z \geq B_O$  is **redundant** if:

$$\left[ -K - \frac{\bar{S}_Z}{S_Z}, -K - \frac{\bar{S}_X}{C_X}, K, 1 \right] \begin{bmatrix} B_I \\ P_A \\ P_B \\ B_O \end{bmatrix} \leq 0$$

where

$$K \doteq \frac{\bar{S}_Y - \bar{S}_X \frac{C_Y}{C_X} - S_Y \frac{\bar{S}_Z}{S_Z}}{C_Y + S_Y}$$

## BEFORE PROCEEDING WITH THE MATH ...

What can be modeled by linear inequalities?

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## WHAT ARE EVENT STREAMS

- An **event** is the crossing of one control flow through one selected device and instruction. Example: one function call.
- The **type** of an event is its equivalence class with respect to multiple objectives. Example: all lookups in a table.
- A **stream** is a counter over time of events of the same type.

A stream does not need to be localized, events can be anywhere.

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What aggregations can be modeled by linear inequalities?

$$\alpha_1 x_1 + \dots + \alpha_n x_n \geq \beta$$

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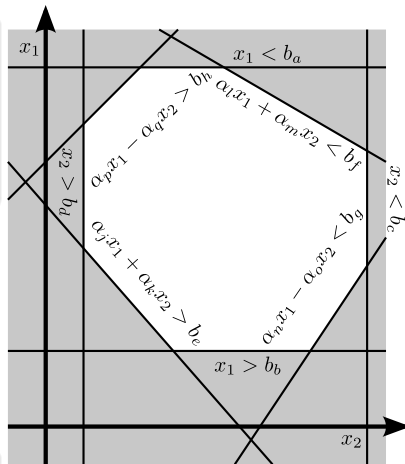
What aggregations can be modeled by linear inequalities?

$$Ax \geq b$$

- It is always an n-dimensional polytope.
- It must be convex, it can be open, it can imply equalities.

## ... TURNS INTO QUESTION

What do facets model?

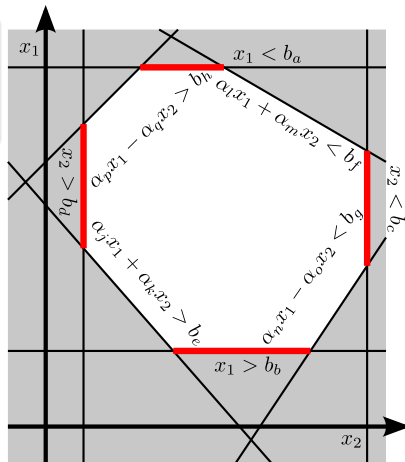


### INDIVIDUAL WORST & BEST

$$b_b < x_1 < b_a$$

$$b_d < x_2 < b_c$$

- Absolute best and worst cases for the event streams.
- Worst cases can be caused by bottlenecks like channel capacity.
- Best cases can be intended “under available workload”.

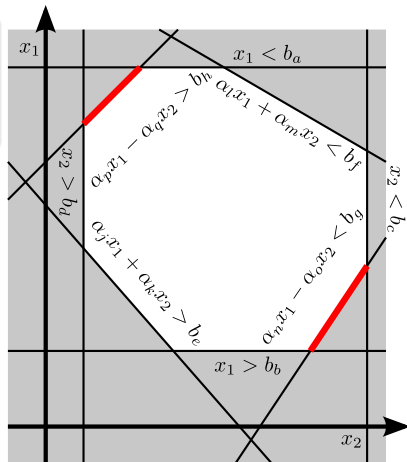


## MAX &amp; MIN SHIFT

$$\alpha_n x_1 - \alpha_o x_2 < b_g$$

$$\alpha_p x_1 - \alpha_q x_2 > b_h$$

- One stream cannot be increased above a given limit, without increasing also the other one.
- Reasons can be data hazards, but also workload characteristics like frame types in an MPEG stream.

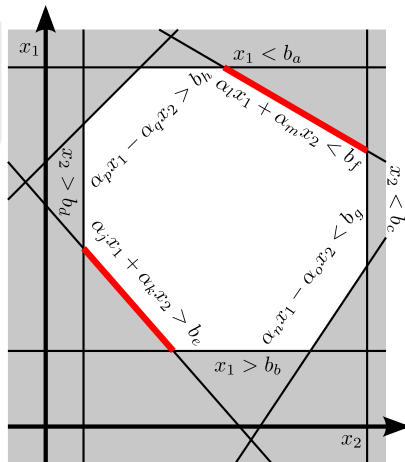


### COMMON WORST & BEST

$$\alpha_j x_1 + \alpha_k x_2 > b_e$$

$$\alpha_l x_1 + \alpha_m x_2 < b_f$$

- Upper limit of the system's performance under peak load and improvements over the individual worst cases.
- The system will process more events of one type if it less loaded from events of the other type.



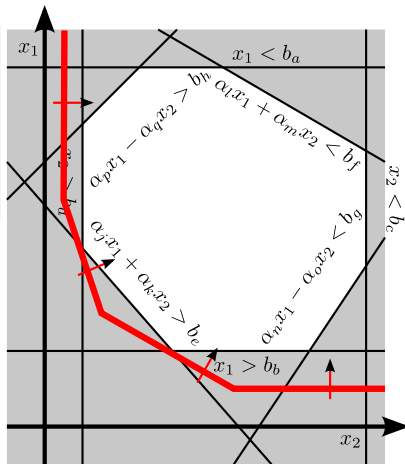
### GUARANTEE & LIMITATION

$$Ax \geq b \rightsquigarrow D^+ x \leq r^+$$

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$\rightsquigarrow$  means “is a dependency set for the infeasibility of”

- The demand is guaranteed:  $D^+ x \geq r^+$  is always satisfied.
- The limitation is never violated:  $D^\pm x \leq r^+$  is always satisfied.



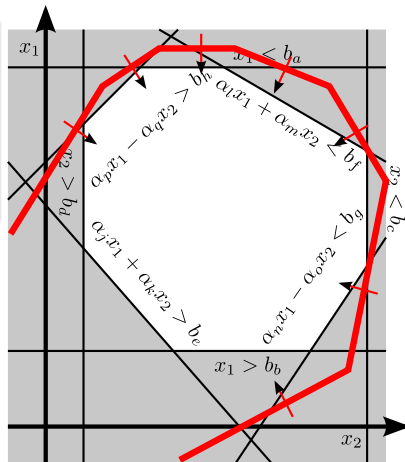
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## PERFORMANCE OVER TIME

Average performance at module level is not indicative.

- “phase behaviors” play a major role in programs.
- Applications settle down on these phases.

Disregarding such effects leads to grossly wrong estimations: the bounds for the event counters must be **time dependent**.

## PERFORMANCE UNDER VARIABILITY

Effects difficult to model in simulations:

- Same code paths run at different performance levels.
- Same initial states lead to different code paths.

Do not use averages, use **confidence intervals**.

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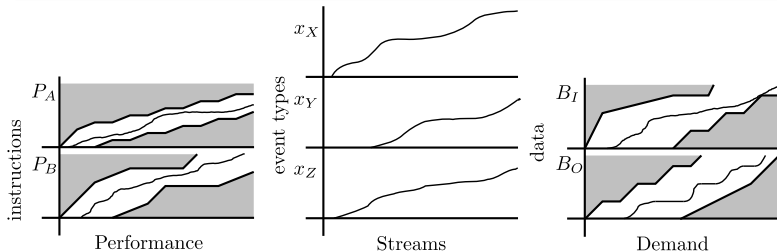
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### FUNDAMENTAL QUESTION

Which performance bounds satisfy the demand bounds?



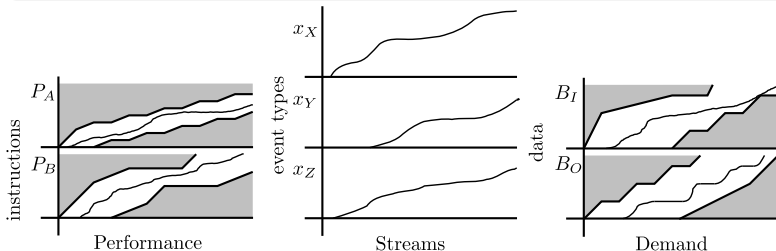
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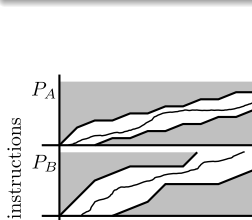
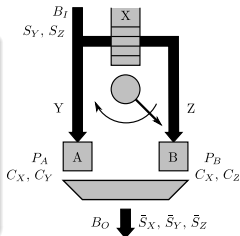
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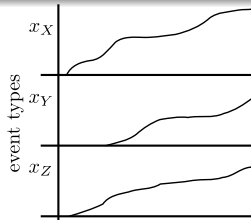
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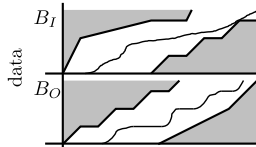
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Performance



Streams



Demand

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## SLIDING WINDOW UNDER LOAD

For every time window, determine

- the highest increment: arrival curve
- the lowest increment **under load**: service curve

## COMPLEX DETERMINATION

- must update all sliding windows at every simulation step
- must identify and skip idle conditions

## PIECEWISE LINEAR PROFILES

Rather than cycle by cycle, characterize event by event and assume linear growth between events.

- calculate and round intersections between linear pieces
- avoid accumulating rounding errors during incremental steps

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$$\left[ -K - \frac{\bar{S}_Z}{S_Z}, -K - \frac{\bar{S}_X}{C_X}, K, 1 \right] \begin{bmatrix} B_I \\ P_A \\ P_B \\ B_O \end{bmatrix} \leq 0$$

## POLAR CONE OF THE DUAL SPACE OF THE INCONSISTENT PROBLEM

$$\begin{bmatrix} 0 & C_X & C_X & \bar{S}_X \\ S_Y & C_Y & 0 & \bar{S}_Y \\ S_Z & 0 & C_Z & \bar{S}_Z \end{bmatrix} \begin{bmatrix} \pi_1 \\ \pi_2 \\ \pi_3 \\ \pi_4 \end{bmatrix} = 0, \forall \pi_1 \geq 0, \pi_2 \geq 0, \pi_3 \geq 0, \pi_4 < 0$$

- the dual space is a line: 3 equations in four variables
- the polar cone is a halfspace, the parameters vector must remain out of that halfspace: in this case just sign inversion.

## A CONVEX HULL PROBLEM

Determining the **halfspace representation** of the prohibited regions for the parameters is a **convex hull** problem.

- simplified by the null space, but still complex.
- solved by double description method (CDD by Fukuda).
- performance strictly depending on the problem formulation.
- as a rule of thumb, in general no more than 20 variables.

## INVERSE PROBLEM: WHICH DEMAND CAN BE SUSTAINED?

Determine the space of redundant inequalities.

- employ the subsumption cone by Lassez
- solve by Fourier-Motzkin elimination (complex!)

Example on the next slide.

## TWO THREADS WITH OVERHEAD

Thread 1 guarantees a service  $b_1$ , thread 2 guarantees a service  $b_2$ . Either thread 1 or 2 is active, so  $b_3 > b_1 + b_2$ .

Which demand  $\alpha_1 x_1 + \alpha_2 x_2 \geq \beta$  can be sustained?

## AS INEQUALITY SYSTEM $Ax \geq b$

$$\begin{array}{rcl} & +x_1 & \geq b_1 \\ x_2 & & \geq b_2 \\ x_2 + x_1 & \geq b_3 \end{array}$$

Subsumption cone:

$$\{\pi A - \alpha = 0, \pi b - \beta \geq 0, \pi \geq 0\}$$

## SUBSUMPTION CONE

$$\begin{array}{cccccc} & \pi_2 & +\pi_3 & -\alpha_1 & = & 0 \\ \pi_1 & & +\pi_3 & -\alpha_2 & = & 0 \\ b_1 \pi_1 & +b_2 \pi_2 & +b_3 \pi_3 & -\beta & \geq & 0 \\ \pi_1 & & & & \geq & 0 \\ & \pi_2 & & & \geq & 0 \\ & & \pi_3 & & \geq & 0 \end{array}$$

## FOURIER-MOTZKIN ELIMINATION

$$\begin{array}{rcl} \pi_3(b_3 - b_2 - b_1) + \alpha_1 b_2 + \alpha_2 b_1 - \beta & \geq & 0 \\ & -\pi_3 & \geq \alpha_1 \\ & -\pi_3 & \geq -\alpha_2 \\ & \pi_3 & \geq 0 \end{array}$$

The coefficient of  $\pi_3$  is  $b_3 - b_2 - b_1 \geq 0$  per construction.

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$$\begin{array}{rcl} & +x_1 & \geq b_1 \\ x_2 & & \geq b_2 \\ x_2 + x_1 & \geq & b_3 \end{array}$$

Subsumption cone:

$$\{\pi A - \alpha = 0, \pi b - \beta \geq 0, \pi \geq 0\}$$

## SUBSUMPTION CONE

$$\begin{array}{rcccccc} & & \pi_2 & +\pi_3 & -\alpha_1 & = & 0 \\ \pi_1 & & & +\pi_3 & -\alpha_2 & = & 0 \\ b_1 \pi_1 & +b_2 \pi_2 & +b_3 \pi_3 & -\beta & \geq & 0 \\ \pi_1 & & & & \geq & 0 \\ & \pi_2 & & & \geq & 0 \\ & & \pi_3 & & \geq & 0 \end{array}$$

## FOURIER-MOTZKIN ELIMINATION

$$\begin{array}{rcl} \pi_3(b_3 - b_2 - b_1) + \alpha_1 b_2 + \alpha_2 b_1 - \beta & \geq & 0 \\ & -\pi_3 & \geq \alpha_1 \\ & -\pi_3 & \geq -\alpha_2 \\ & \pi_3 & \geq 0 \end{array}$$

The coefficient of  $\pi_3$  is  $b_3 - b_2 - b_1 \geq 0$  per construction.

## TWO THREADS WITH OVERHEAD

Thread 1 guarantees a service  $b_1$ , thread 2 guarantees a service  $b_2$ . Either thread 1 or 2 is active, so  $b_3 > b_1 + b_2$ .

Which demand  $\alpha_1 x_1 + \alpha_2 x_2 \geq \beta$  can be sustained?

AS INEQUALITY SYSTEM  $Ax \geq b$ 

$$\begin{array}{rcl} & +x_1 & \geq b_1 \\ x_2 & & \geq b_2 \\ x_2 & +x_1 & \geq b_3 \end{array}$$

Subsumption cone:

$$\{\pi A - \alpha = 0, \pi b - \beta \geq 0, \pi \geq 0\}$$

## SUSTAINABLE DEMAND SPACE

$$-\alpha_2 b_1 - \alpha_1 (b_3 - b_1) + \beta \leq 0$$

$$-\alpha_2 (b_3 - b_2) - \alpha_1 b_2 + \beta \leq 0$$

$$\alpha_1 \geq 0$$

$$\alpha_2 \geq 0$$

Dimensioning  
Event Streams

Di Gregorio

Motivation

Problem  
Example

Event Streams

Definition  
Modeling  
Bounds over time

Techniques

Empirical  
Analytical

# Thank you!

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