

Introduction to Intelligent Vehicles

[4. System Design]

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Announcement

□ Lecture plan

- No class on October 14
- We will consider a make-up class in December (after the midterm)

□ Homework

➤ Homework 1

- Graded
- Reminder: definition of B_i , ceiling function, connection between Q1 and Q2
- Regrade request due on October 9 (Wednesday) 11:59pm

➤ Homework 2

- Posted
- Due on October 31 (Monday) noon

□ If you have any question, please feel free to

- Post on NTU COOL or send me an email

Revisit CAN Timing Analysis

$$\underbrace{\underbrace{Q_i}_{(1)} = \underbrace{B_i}_{(2)} + \underbrace{\sum_{(\text{for all } j, P_j < P_i)} \left[\underbrace{\frac{Q_j + \tau}{T_j}}_{(4)} \right] \underbrace{C_j}_{(5)}}_{(6)}}_{(7)}$$

$$\underbrace{R_i}_{(7)} = \underbrace{Q_i}_{(7)} + \underbrace{C_i}_{(7)}$$

(1): (2)+(6)

(2): blocking time of the longest lower or same priority message

(3): index set of all higher priority messages

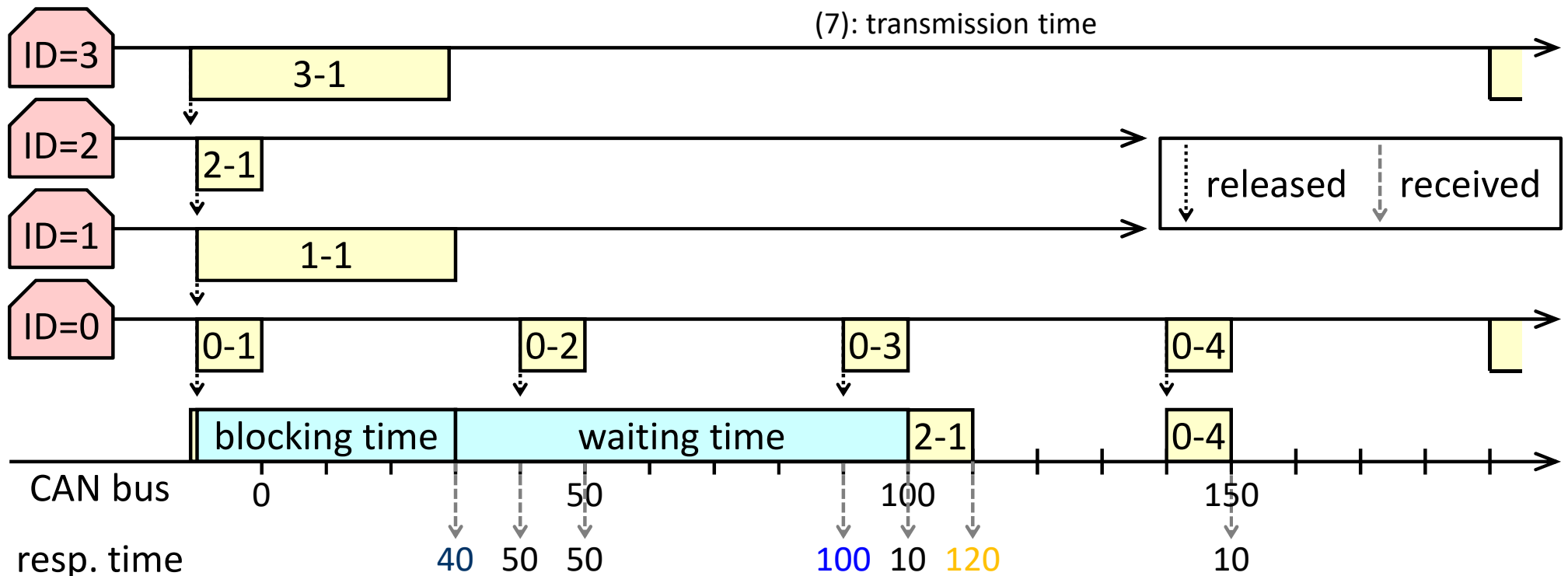
(4): max number of queued instances of message j within (1)

τ : transmission of one bit (Slide P31 for the reason)

(5): transmission time of an instance of message j

(6): waiting time

(7): transmission time



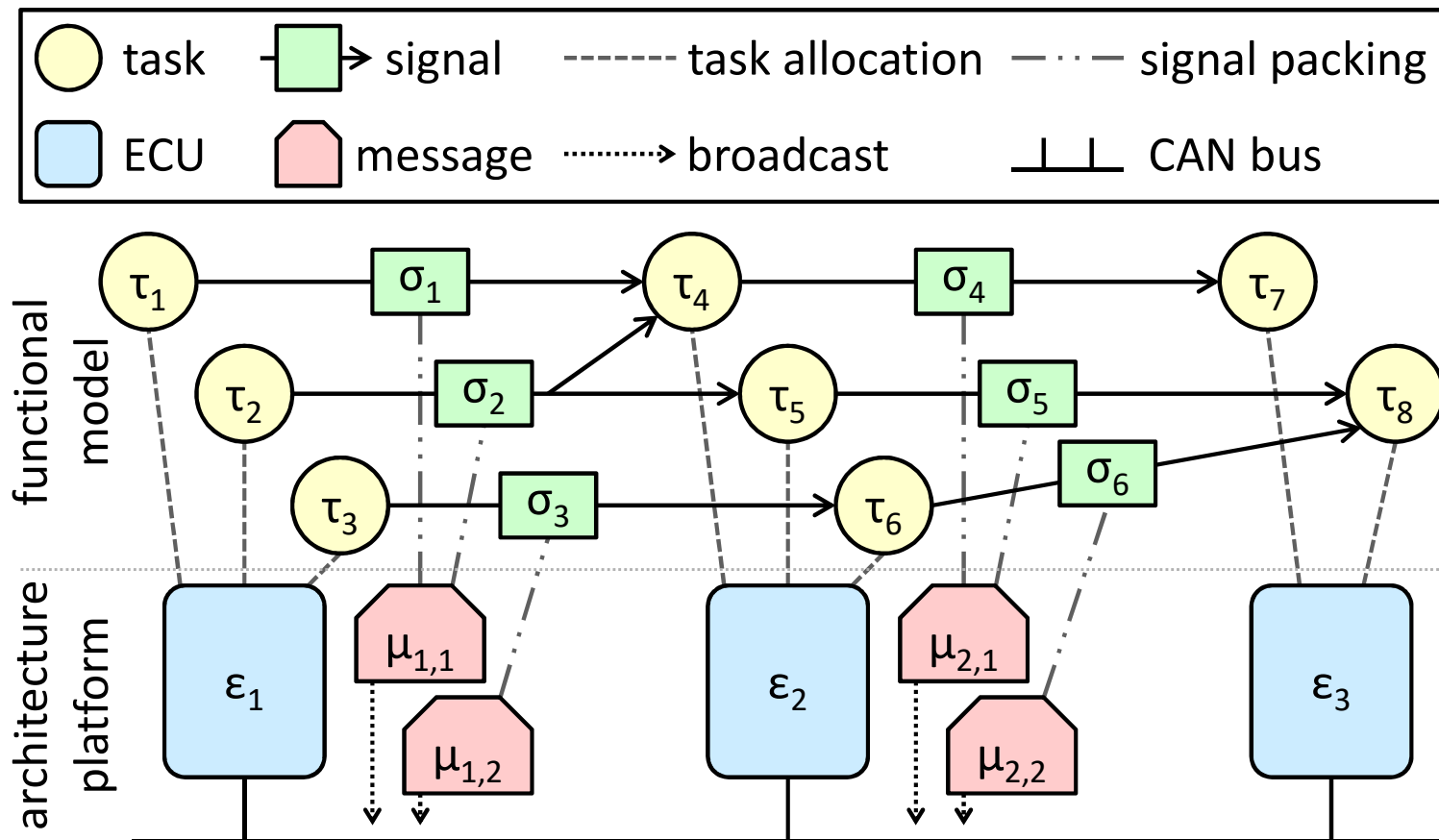
$$R_0 = 50 \quad R_1 = 100 \quad R_2 = 120 \quad R_3 = 40?$$

System Design

- ☐ What is system design?
- ☐ Why is system design needed?
- ☐ When is system design done?
- ☐ Where is system design done?
- ☐ Who performs system design?
- ☐ How to perform system design?

Example (Piece) of System Design

- ❑ What is system design? / Why is system design needed?
- ❑ When / Where is system design done?
- ❑ Who / How to perform(s) system design?



Cores of Model-Based Design

□ Modeling

- Work (design and analyze) upon models rather than real systems

□ Design

- Optimize some objectives (performance, robustness, security, etc.)
- Satisfy some constraints

□ Analysis

- Check if there is any flaw or how good the designs are?

□ (Implementation)

- Make it real

Outline

- ❑ Mixed Integer Linear Programming (MILP)
- ❑ Simulated Annealing
- ❑ Mapping Problem
 - It is just a piece of system design

Linear Programming

□ Linear Programming (LP)

➤ Maximize

- $2x_1 - x_2$

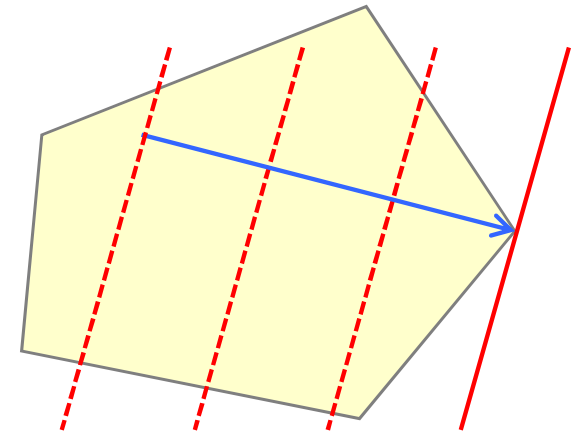
➤ Subject to

- $x_1 - x_2 \leq 1$

- $2x_2 \leq 3$

- $x_1, x_2 \geq 0$

- x_1, x_2 are real numbers



□ Canonical form

$$\min \quad \mathbf{c}^T \mathbf{x}$$

$$\text{subject to} \quad \mathbf{Ax} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0}, \mathbf{x} \text{ in } \mathbb{R}$$

□ Simplex algorithm

➤ https://en.wikipedia.org/wiki/Simplex_algorithm

Integer (Linear) Programming (ILP)

□ Integer (Linear) Programming (ILP)

➤ Maximize

- $2x_1 - x_2$

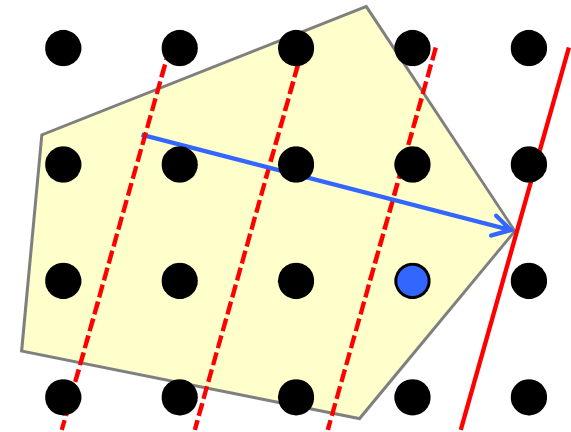
➤ Subject to

- $x_1 - x_2 \leq 1$

- $2x_2 \leq 3$

- $x_1, x_2 \geq 0$

- x_1, x_2 are integers



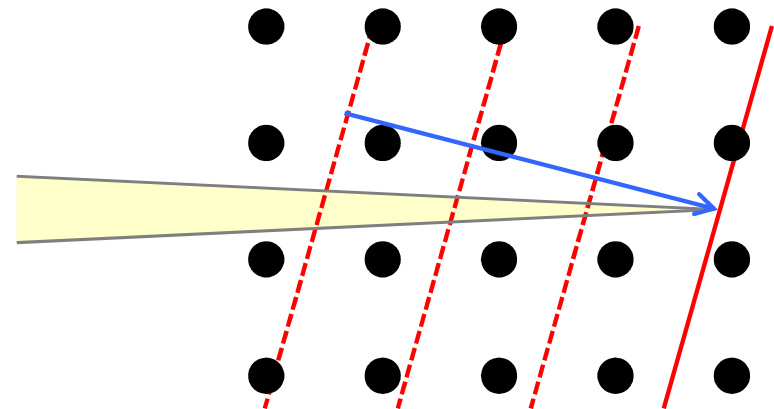
□ Canonical form

$$\max \quad \mathbf{c}^T \mathbf{x}$$

$$\text{subject to} \quad \mathbf{Ax} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0}, \mathbf{x} \in \mathbb{Z}$$

□ NP-complete

➤ Why is it hard?



Mixed Integer Linear Programming (MILP)

□ Mixed Integer Linear Programming (MILP)

➤ Maximize

- $2x_1 - x_2$

➤ Subject to

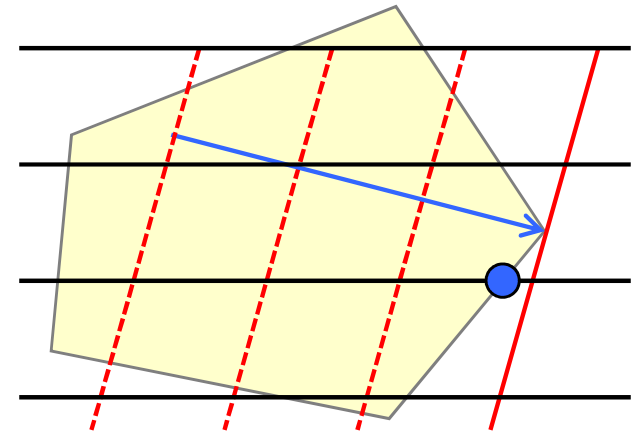
- $x_1 - x_2 \leq 1$

- $2x_2 \leq 3$

- $x_1, x_2 \geq 0$

- x_1 is real number

- x_2 is integer



□ Canonical form

$$\max \quad \mathbf{c}^T \mathbf{x}$$

subject to $\mathbf{Ax} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0}$, some x are integers

□ Many tools are available for solving LP, ILP, and MILP

Quadratic Programming (QP)

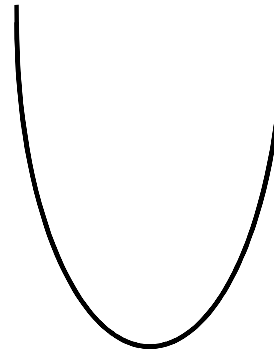
□ Canonical form

$$\min \quad \frac{1}{2} \mathbf{x}^T \mathbf{Q} \mathbf{x} + \mathbf{c}^T \mathbf{x}$$

$$\text{subject to} \quad \mathbf{A} \mathbf{x} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0}$$

- \mathbf{Q} is symmetric

➤ A special case of convex optimization



□ Convex optimization

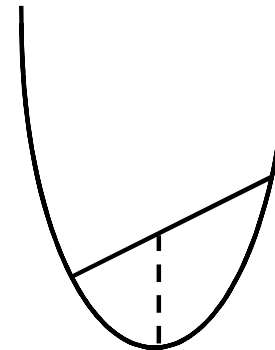
$$\max \quad f(\mathbf{x})$$

$$\text{subject to} \quad g_i(\mathbf{x}) \leq 0 \text{ for all } i$$

- f, g_i are convex

➤ A function f is convex if

- For all x_1 and x_2 in the domain of f , t in $[0,1]$
 $f(t \cdot x_1 + (1-t) \cdot x_2) \leq t \cdot f(x_1) + (1-t) \cdot f(x_2)$



Practice

□ Bin packing problem [Wikipedia]

➤ Given

- A set of bins S_1, S_2, \dots with the same size V
- A set of n items with sizes a_1, a_2, \dots, a_n

➤ "Pack" all items into bins and minimize the number of used bins

□ LP? ILP? MILP? QP?

max $\mathbf{c}^T \mathbf{x}$

subject to $\mathbf{Ax} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0}, \mathbf{x} \in \mathbb{Z}$

min $y_1 + y_2 + \dots + y_n$

subject to $x_{i1} + x_{i2} + \dots + x_{in} = 1$ for all i

$a_1 x_{1j} + a_2 x_{2j} + \dots + a_n x_{nj} \leq V y_j$ for all j

$x_{ij} = 0$ or $1 \rightarrow 1$: if and only if item i is packed in to bin j

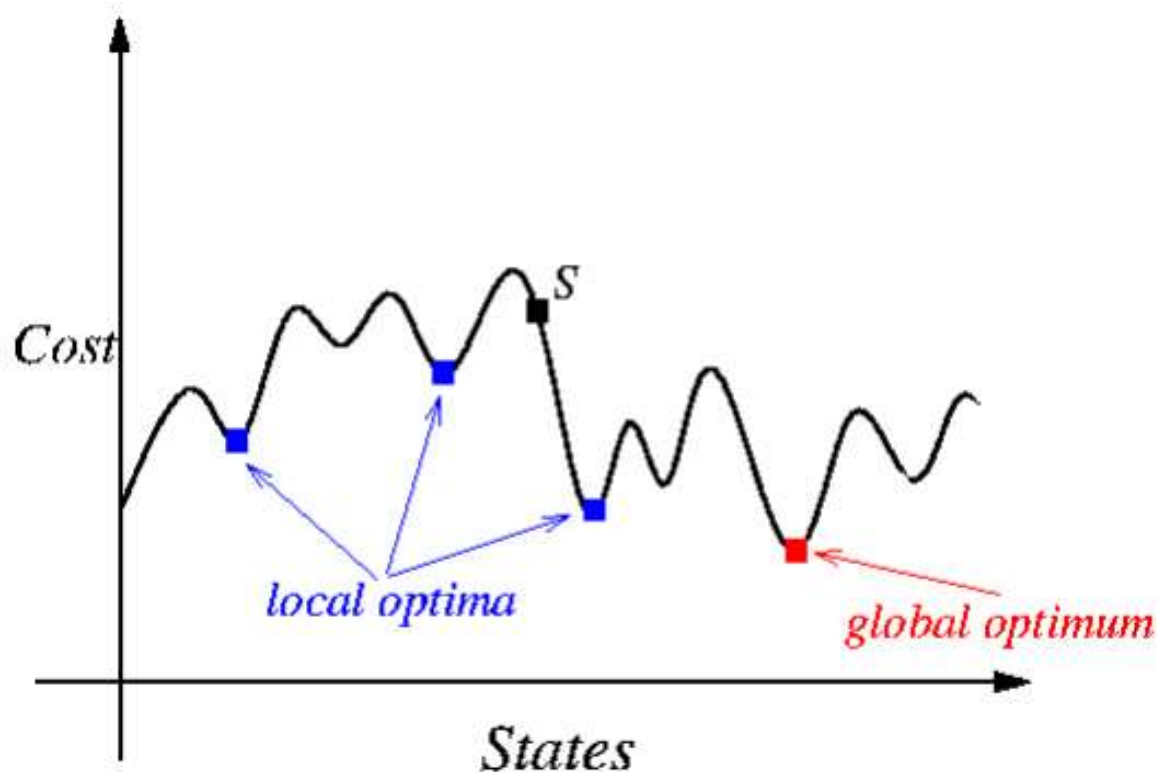
$y_j = 0$ or $1 \rightarrow 1$: if and only if bin j is used

Outline

- ❑ Mixed Integer Linear Programming (MILP)
- ❑ **Simulated Annealing**
- ❑ Mapping Problem
 - It is just a piece of system design

Simulated Annealing: Background

- Kirkpatrick, Gelatt, and Vecchi, "Optimization by simulated annealing," Science, May 1983.



Simulated Annealing: Basics

- ❑ Non-zero probability for "up-hill" moves
- ❑ Probability depends on
 - Magnitude of the "up-hill" movement
 - Total search time
- ❑ $\text{Prob}(S \rightarrow S') = \min(1, e^{-\Delta C/T})$
 - Equivalence
 - $\text{Prob}(S \rightarrow S') = 1$ if $\Delta C \leq 0$ ("down-hill" move)
 - $\text{Prob}(S \rightarrow S') = e^{-\Delta C/T}$ if $\Delta C > 0$ ("up-hill" move)
 - $\Delta C = \text{cost}(S') - \text{cost}(S)$
 - "Smaller" is better
 - T = temperature
 - T will cool down gradually

Simulated Annealing: Algorithm

- ❑ Get an initial solution S
- ❑ Get an initial temperature $T > 0$
- ❑ $S^* = S$
- ❑ While "not yet frozen" (i.e., T is large enough)
 - (Iterate many times)
 - Pick a random neighbor S' of S
 - $\Delta C = \text{cost}(S') - \text{cost}(S)$
 - If $\text{cost}(S') < \text{cost}(S^*)$, then $S^* = S'$
 - If $\Delta C \leq 0$, then $S = S'$
 - If $\Delta C > 0$, then $S = S'$ with probability $e^{-\Delta C/T}$
 - $T = rT$ (where $r < 1$)
- ❑ Return S^*

Simulated Annealing

□ Basic ingredients

- Solution space
 - What are the feasible solutions?
- Neighborhood structure
 - How to find a neighboring solution from the current one?
- Cost function
 - How to evaluate the quality of a solution?
- Annealing schedule
 - How to conduct the search process to find a desired solution?

□ Philosophy

Practice

□ Bin packing problem [Wikipedia]

- Given
 - A set of bins S_1, S_2, \dots with the same size V
 - A set of n items with sizes a_1, a_2, \dots, a_n
- "Pack" all items into bins and minimize the number of used bins

□ Basic ingredients

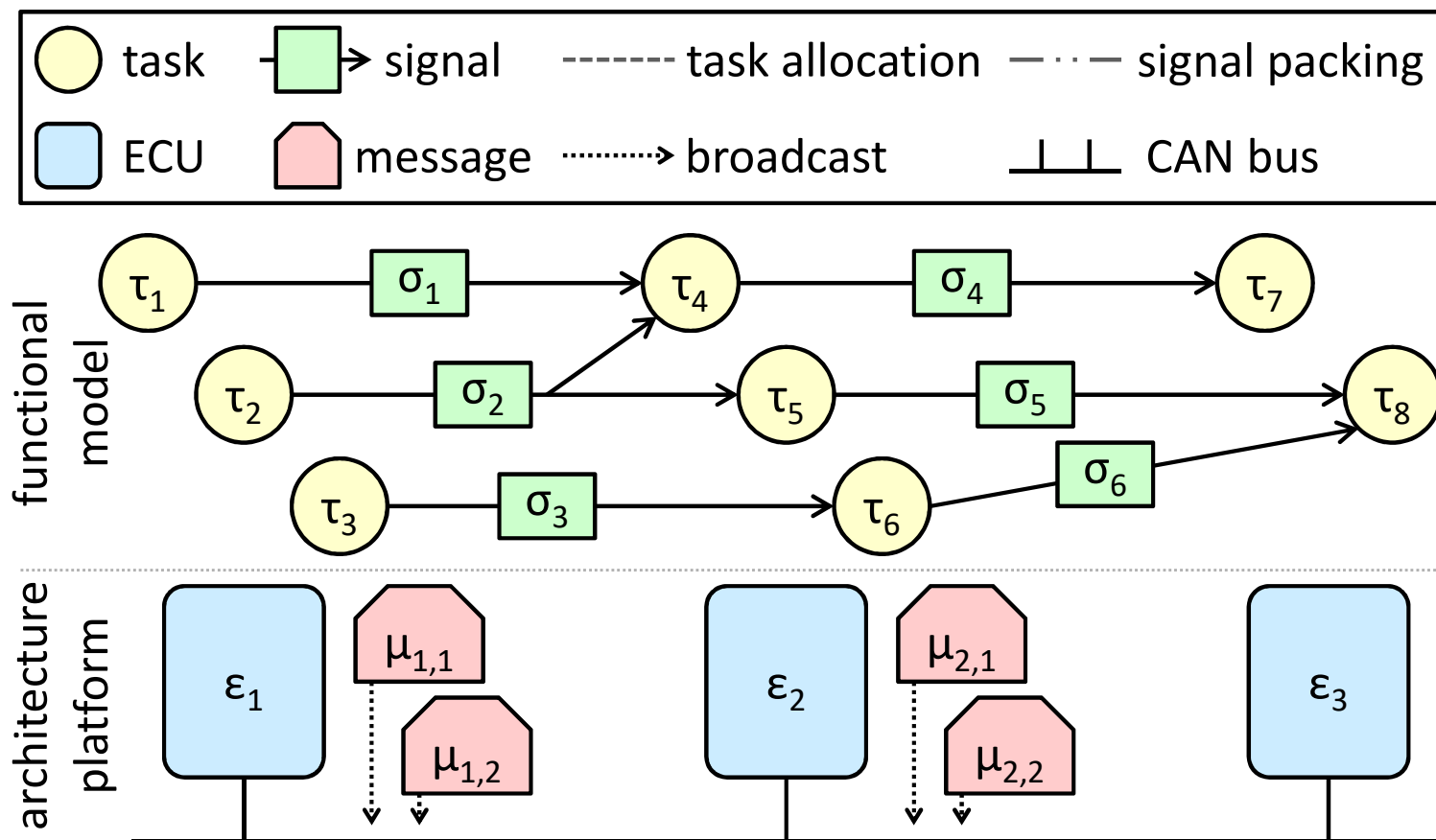
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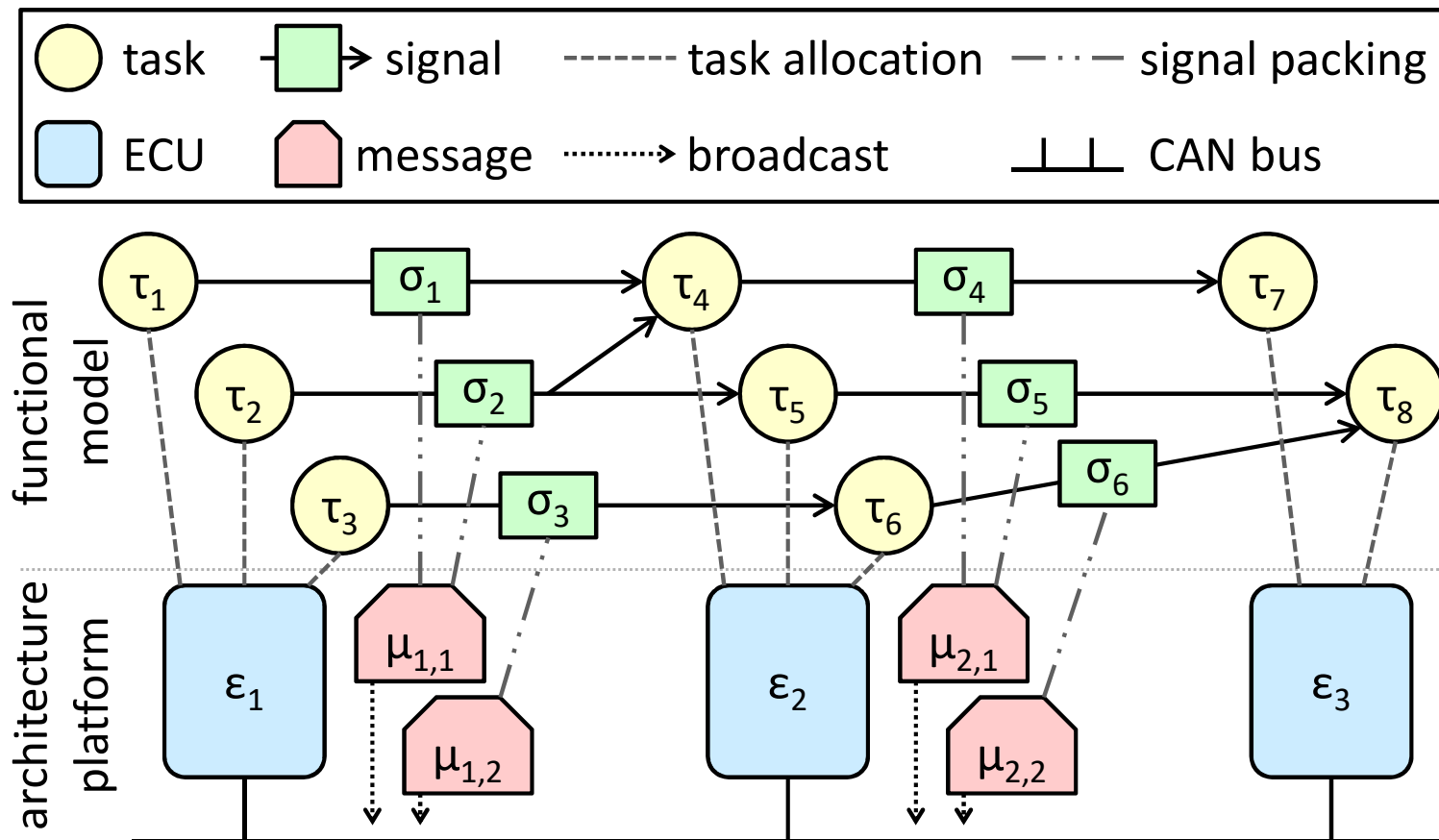
Mapping from Software to Hardware

- ❑ Software (functional model): task graph
- ❑ Hardware (architectural platform): distributed Electronic Control Units (ECUs) connected by a network



Problem Formulation

- ❑ Decide task allocation, signal packing, and priority assignments (tasks on ECUs and messages on CAN bus)
- ❑ Satisfy timing constraints for tasks, signals, and paths



Task Allocation

□ Indices and variables

- i, j : index of a task
- k : index of an ECU
- $a_{i,k}$: $[0,1]$ task i is allocated onto ECU k
- $s_{i,j}$: $[0,1]$ tasks i and j are allocated onto the same ECU

□ Constraints

- $\sum_k a_{i,k} = 1$ for all i
- $a_{i,k} + a_{j,k} + s_{i,j} \neq 2$ for all i, j, k

□ Questions

- What are the meanings of the constraints?
- Why do we need $s_{i,j}$?
- Is it possible that all tasks are allocated onto one ECU?

Task Priority Assignment

□ Indices and variables

- i, j, j' : index of a task
- $p_{i,j}$: $[0,1]$ task i has a higher priority than another task j

□ Constraints

- $p_{i,j} + p_{j,i} = 1$ for all $i, j, i \neq j$
- $p_{i,j} + p_{j,j'} - 1 \leq p_{i,j'}$ for all $i, j, j', i \neq j, i \neq j', j \neq j'$

□ Questions

- What are the meanings of the constraints?
- Why do we use binary variables $p_{i,j}$, not integer $p_i = 1, 2, 3, \dots$?
 - $p_i \neq p_j$
 - $p_i < p_j$ or $p_j < p_i$
 - $p_{i,j} + p_{j,i} = 1$ and $(p_i - p_j) < (1 - p_{i,j}) M$ and $(p_j - p_i) < p_{i,j} M$
 - M : a large constant

Signal Packing

□ Indices, constant parameters, and variables

- i, j : index of a task; k : index of an ECU; l : index of a message
- $T_{i,j}$: the period of the signal from task i to task j
- $T_{k,l}$: the period of ECU k 's message l
- $a_{i,k}$: $[0,1]$ task i is allocated onto ECU k
- $t_{i,j,k,l}$: $[0,1]$ the signal from task i to task j is packed into ECU k 's message l
- $v_{k,l}$: $[0,1]$ ECU k 's message l is used

□ Constraints

- $\sum_l t_{i,j,k,l} = a_{i,k} (1 - a_{j,k})$ for all i, j, k
- $t_{i,j,k,l} \leq v_{k,l}$ for all i, j, k, l
- $t_{i,j,k,l} T_{k,l} \leq T_{i,j}$; $t_{i,j,k,l} T_{i,j} \leq T_{k,l}$ for all i, j, k, l

□ Questions

- What are the meanings of the constraints?
- Why do we need $v_{k,l}$?

Message Priority Assignment

- Similar to task priority assignment

Timing Constraints

❑ Task: response time \leq period (given)

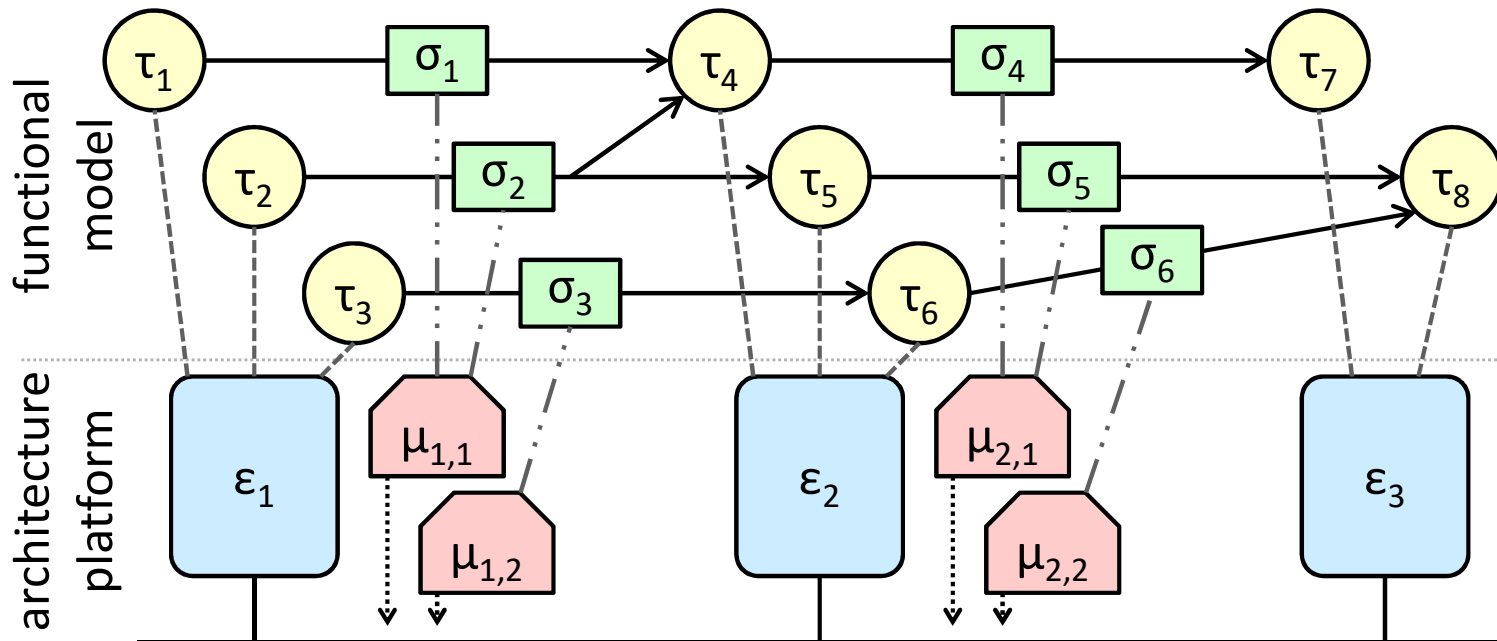
➤ Response time? $r_i = c_i + \sum_{p_j < p_i} \left\lceil \frac{r_i}{T_j} \right\rceil C_j$

❑ Message and signal: response time \leq period (given)

➤ Response time? $q_i = b_i + \sum_{p_j < p_i} \left\lceil \frac{q_i + \tau}{T_j} \right\rceil c_j$ and $r_i = q_i + c_i$

❑ Path: latency \leq deadline (given)

➤ Latency?



Remaining Questions

□ Variables

- $s_{i,j}$: $[0,1]$ tasks i and j are allocated onto the same ECU
- $v_{k,l}$: $[0,1]$ ECU k 's message l is used

□ Questions

- Task allocation
 - Why do we need $s_{i,j}$?
 - $s_{i,j} \cdot p_{i,j}$ in response time computation
 - Is it possible that all tasks are allocated onto one ECU?
 - We have timing constraints
- Signal packing
 - Why do we need $v_{k,l}$?
 - $v_{k,l} \cdot (\text{message size})$ in response time computation

Solved by MILP: Linearization

□ Inequality of three binary variables: $\alpha + \beta + \gamma \neq 2$

➤ $\alpha + \beta + \gamma \neq 2 \iff \alpha + \beta - \gamma \leq 1; \alpha - \beta + \gamma \leq 1; -\alpha + \beta + \gamma \leq 1$

□ Ceiling function: $\text{ceil}(f)$

➤ Replace $\text{ceil}(f)$ by an integer x

➤ $\text{ceil}(f) = x \iff 0 \leq x - f < 1$

□ Multiplication of two binary variables: $\alpha \cdot \beta$

➤ Replace $\alpha \cdot \beta$ by a binary variable γ

➤ $\alpha \cdot \beta = \gamma \iff \alpha + \beta - 1 \leq \gamma; \gamma \leq \alpha; \gamma \leq \beta$

□ Multiplication of a binary variable α and a real variable x : $\alpha \cdot x$

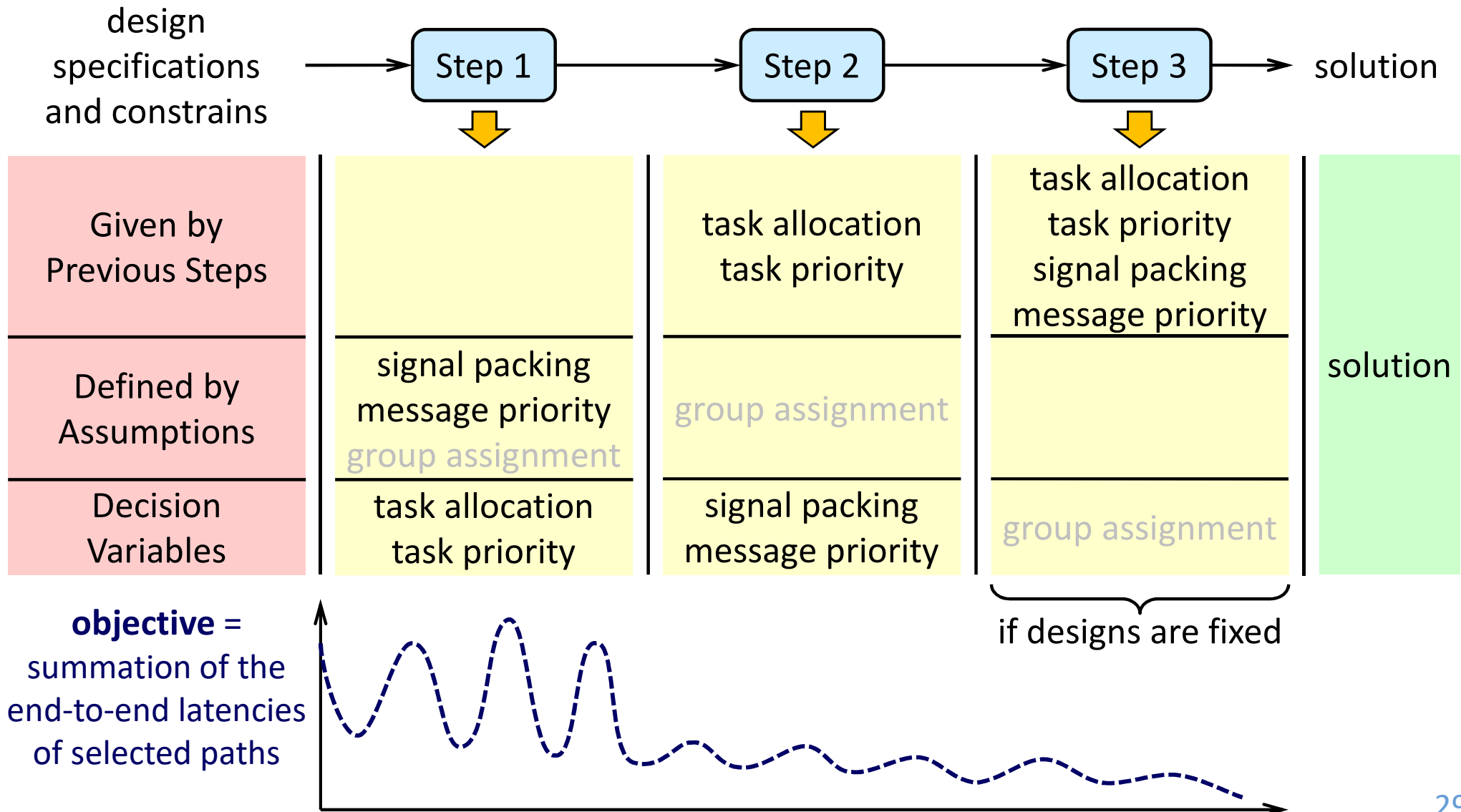
➤ Replace $\alpha \cdot x$ by a real variable y

➤ $\alpha \cdot x = y \iff 0 \leq y \leq x; x - M(1 - \alpha) \leq y \leq M\alpha$

- M : a large constant

Solved by MILP: Scalability Issue

□ Let us ignore "group assignment" first



Solved by Simulated Annealing

□ Basic ingredients

- Solution space
 - What are the feasible solutions?
- Neighborhood structure
 - How to find a neighboring solution from the current one?
- Cost function
 - How to evaluate the quality of a solution?
- Annealing schedule
 - How to conduct the search process to find a desired solution?

Implications

❑ Allocate tasks onto the same ECU

- A signal does not need to be transmitted if its source and target tasks are allocated onto the same ECU
- Two signals can be packed into the same message if their source tasks are allocated onto the same ECU

❑ However, not allocate too many tasks onto the same ECU

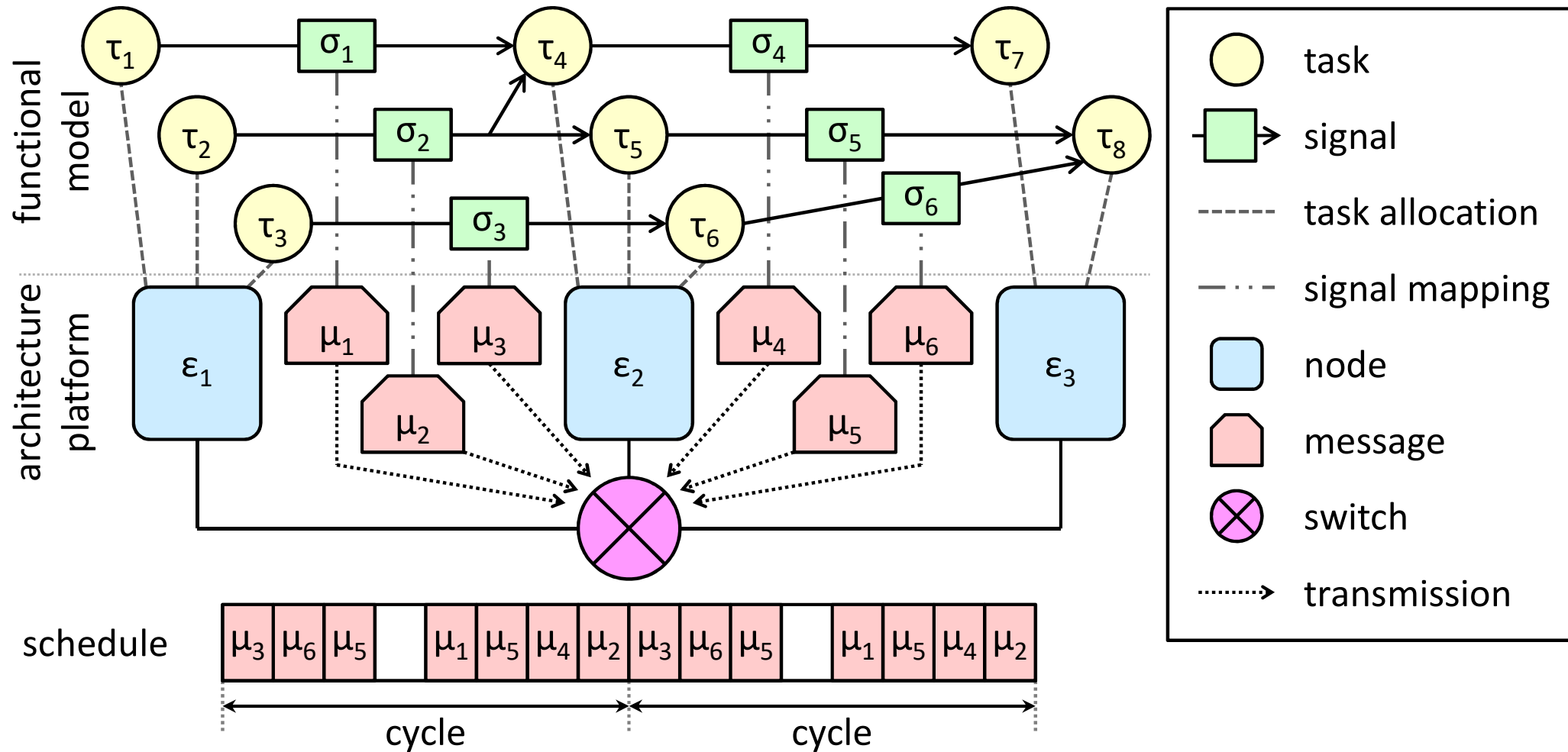
- Long task response time

❑ In most cases, pack signals into the same message

- Save headers!

Another Problem Formulation

Task allocation, task priority assignment, scheduling



Q&A