

# **Traffic Analysis and System Identification with Legendre Transform**

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# 1. Introduction — Objective and Tools

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## Objective

- Traffic analysis in communication network
- Identification of communication network system

**Network system is nonlinear.**

⇒ The analysis is very complicated.

## Tools

- Max-plus algebra
- Legendre transform

# 1. Introduction — Linear description of nonlinear system

**max-plus algebra** is defined by

**addition** :  $\oplus \equiv \max$       and      **multiplication** :  $\otimes \equiv +$  .

We can obtain **linear description** of nonlinear systems;

- Discrete event systems
  - Dynamic programming
  - Markov decision processes (Viterbi algorithm)
  - Morphological operations
- etc

**TCP network is also max-plus linear.**

# 1. Introduction — Our project

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For the system described with max-plus algebra, Legendre transform is a powerful tool.

**Legendre transform in max-plus algebra**

**$\Rightarrow$  Fourier transform and  $z$ -transform in linear system**

**Legendre transform gives**

- **Time series analysis**  
 **$\Rightarrow$  Traffic analysis**
- **Deconvolution**  
 **$\Rightarrow$  System identification**

Our objective is **to clarify complicated traffic and to identify complicated network system** with **Legendre transform**.

## 2. Max-plus algebra — Definition

- **Addition  $\oplus$**

$$\oplus \equiv \max$$

- **Multiplication  $\otimes$**

$$\otimes \equiv +$$

### Example

$$1 \oplus 2 = \max(1, 2) = 2$$

$$2 \otimes 3 = 2 + 3 = 5$$

## 2. Max-plus algebra — **Linear system**

### Difference equations

$$\mathbf{x}(t+1) = \begin{pmatrix} 3 & 7 \\ 2 & 4 \end{pmatrix} \mathbf{x}(t), \quad \mathbf{x}(0) = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

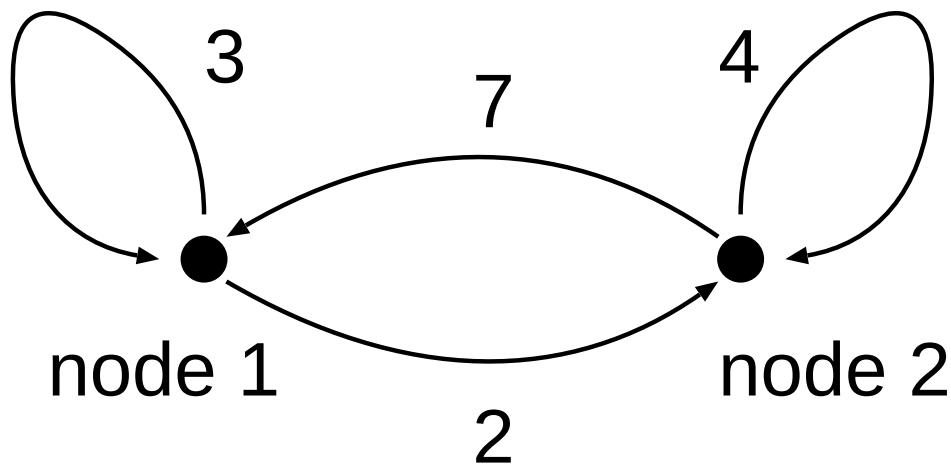
### **Conventional** difference equations

$$\mathbf{x}(1) = \begin{pmatrix} 3 \times 1 + 7 \times 0 \\ 2 \times 1 + 4 \times 0 \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \end{pmatrix}, \quad \mathbf{x}(2) = \begin{pmatrix} 23 \\ 14 \end{pmatrix}, \dots$$

### **Max-plus** difference equations

$$\mathbf{x}(1) = \begin{pmatrix} \max(3+1, 7+0) \\ \max(2+1, 4+0) \end{pmatrix} = \begin{pmatrix} 7 \\ 4 \end{pmatrix}, \quad \mathbf{x}(2) = \begin{pmatrix} 11 \\ 9 \end{pmatrix}, \dots$$

## 2. Max-plus algebra — Simple example

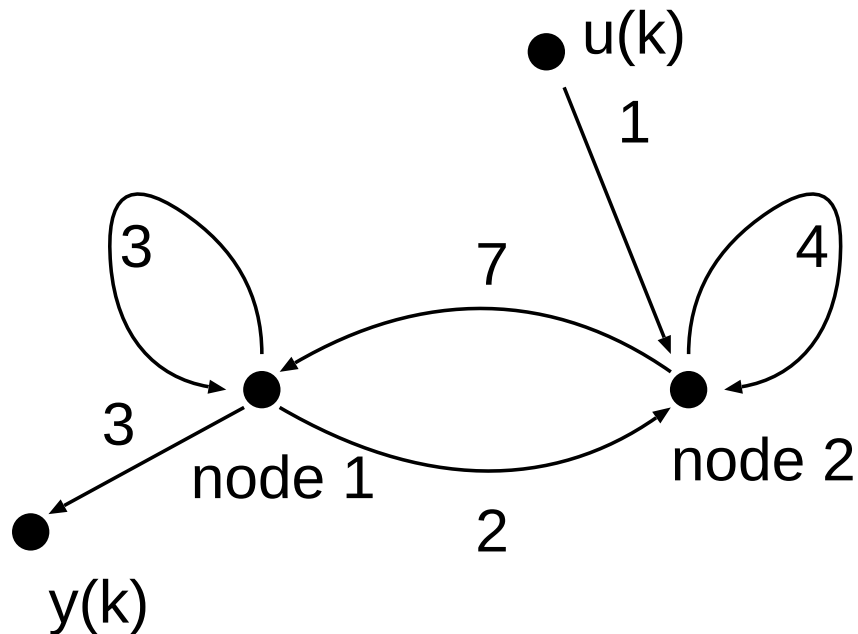


$k$ ) is the earliest epoch when node  $i$  becomes active for the  $k$ -th time.  
is the traveling time from node  $j$  to node  $i$ .

$$\begin{pmatrix} x_1(k+1) \\ x_2(k+1) \end{pmatrix} = \begin{pmatrix} \max(3 + x_1(k), 7 + x_2(k)) \\ \max(2 + x_1(k), 4 + x_2(k)) \end{pmatrix} = A\mathbf{x}(k)$$

$$\mathbf{x}(0) = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \quad \mathbf{x}(1) = \begin{pmatrix} 7 \\ 4 \end{pmatrix}, \quad \mathbf{x}(2) = \begin{pmatrix} 11 \\ 9 \end{pmatrix}, \dots$$

## 2. Max-plus algebra — Example with input and output



$$\begin{aligned} \mathbf{x}(k+1) &= \begin{pmatrix} 3 & 7 \\ 2 & 4 \end{pmatrix} \mathbf{x}(k) \oplus \begin{pmatrix} \epsilon \\ 1 \end{pmatrix} u(k) \\ y(k) &= (3 \quad \epsilon) \mathbf{x}(k) \end{aligned}$$

$$\epsilon \equiv -\infty$$

## 2. Max-plus algebra — Definition of neutral element

Neutral element with respect to **maximization**

$$\epsilon \equiv -\infty, \quad a \oplus \epsilon = a$$

Neutral element with respect to **addition**

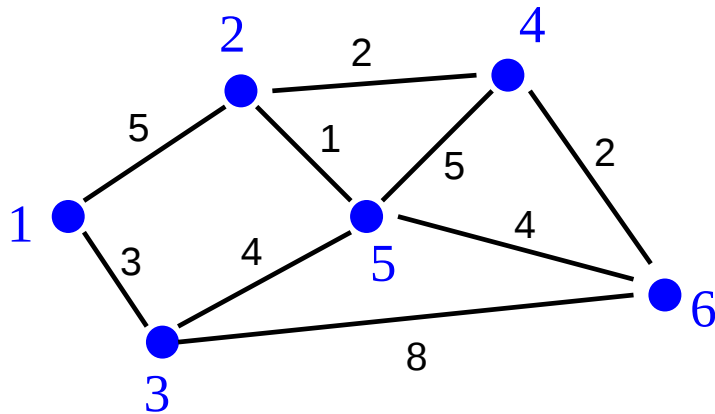
$$e \equiv 0, \quad a \otimes e = a$$

Relation between conventional and max-plus algebra

$$-\infty = \log 0$$

$$0 = \log 1$$

### 3. Modeling — Shortest path problem



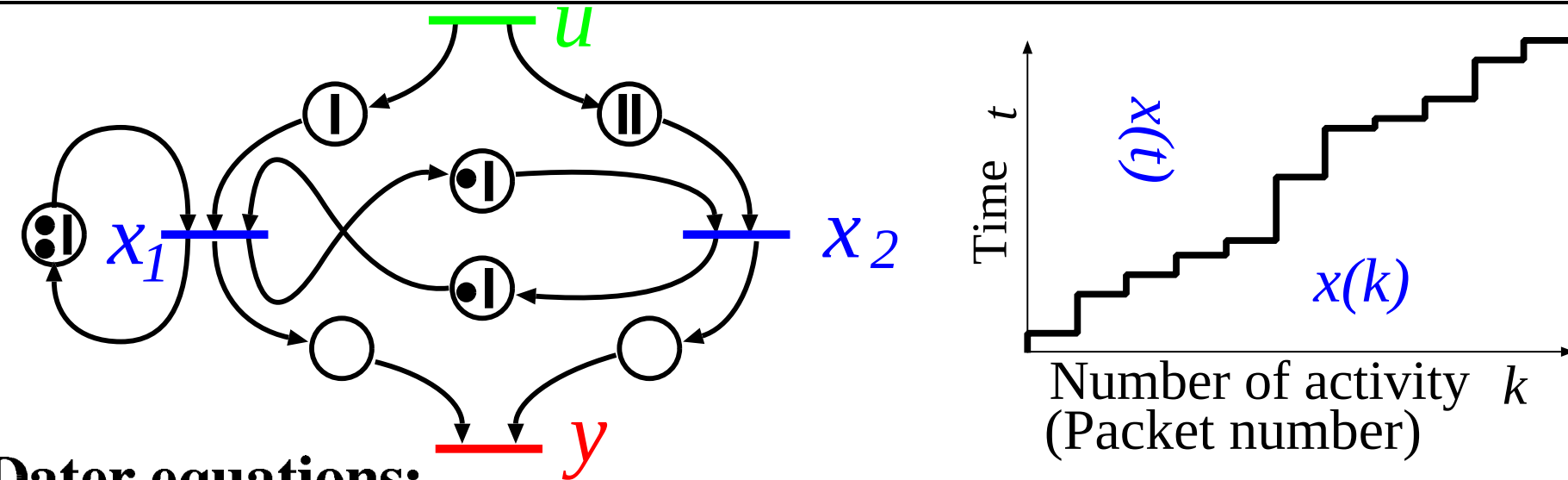
$X_{ij}(k)$  represents the shortest path from  $j$  to  $i$  with  $k$ -path

$$X_{ij}(k+1) = \min_{l=1,\dots,6} (A_{il} + X_{lj}(k)), \quad A = \begin{pmatrix} \epsilon & 5 & 3 & \epsilon & \epsilon & \epsilon \\ 5 & \epsilon & \epsilon & 2 & 1 & \epsilon \\ 3 & \epsilon & \epsilon & \epsilon & 4 & 8 \\ \epsilon & 2 & \epsilon & \epsilon & 5 & 2 \\ \epsilon & 1 & 4 & 5 & \epsilon & 4 \\ \epsilon & \epsilon & 8 & 2 & 4 & \epsilon \end{pmatrix}$$

$$X(k+1) = A \otimes X(k) \quad \text{where} \quad \oplus \equiv \min$$

Shortest path problem is max-plus linear!

### 3. Modeling — Discrete event system and its dual description



**Dater equations:**

$$x_1(k) = 1x_1(k-2) \oplus 1x_2(k-1) \oplus 1u(k)$$

$$x_2(k) = 1x_1(k-1) \oplus 2u(k)$$

$$y(k) = x_1(k) \oplus x_2(k)$$

**Counter equations:**

$$x_1(t) = 2x_1(t-1) \oplus 1x_2(t-1) \oplus u(t-1)$$

$$x_2(t) = 1x_1(t-1) \oplus u(t-2)$$

$$y(t) = x_1(t) \oplus x_2(t)$$

**Petri net is also max-plus linear!**

### 3. Modeling — TCP network

F. Baccelli, D. Hong: **TCP is Max-Plus Linear**, ACM-SIGCOMM, 2000.

$$x(k+1) = A \otimes x(k) \quad x(k) : \text{data vector}$$

- TCP Tahoe and Reno are max-plus linear.
- Analytical evaluation of throughput.
- Analytical detection of losses and timeouts.
- Fractal scaling of TCP is also reported in later paper.

Thus, **Max-plus algebra is efficient for TCP traffic analysis.**

## 4. System Theory — Conventional linear system theory

### Conventional linear system

$$\mathbf{x}(k + 1) = \mathbf{A}\mathbf{x}(k) + \mathbf{B}\mathbf{u}(k), \quad \mathbf{y}(k) = \mathbf{C}\mathbf{x}(k)$$

The output  $\mathbf{y}(k)$  is the **convolution** of impulse response and the input  $\mathbf{u}(k)$ .

$$\mathbf{y}(k) = \sum_{i=-\infty}^k \mathbf{h}(k - i)\mathbf{u}(i), \quad \mathbf{h}(k) : \text{impulse response}$$

The **z-transform**  $X(z) \equiv \sum_{i=0}^{\infty} x(i)z^{-i}$  is convenient because we obtain the following representation:

$$Y(z) = H(z)U(z)$$

The convolution becomes **multiplication**.

## 4. System Theory — Linear system theory in max-plus algebra

Linear system in max-plus algebra

$$\mathbf{x}(k+1) = \mathbf{A}\mathbf{x}(k) \oplus \mathbf{B}\mathbf{u}(k), \quad \mathbf{y}(k) = \mathbf{C}\mathbf{x}(k)$$

The output  $\mathbf{y}(k)$  is the **convolution** of impulse response and the input  $\mathbf{u}(k)$ .

$$\mathbf{x}(k) = \bigoplus_{i=-\infty}^k \mathbf{h}(k-i) \otimes \mathbf{u}(i), \quad \mathbf{h}(k) : \text{impulse response}$$

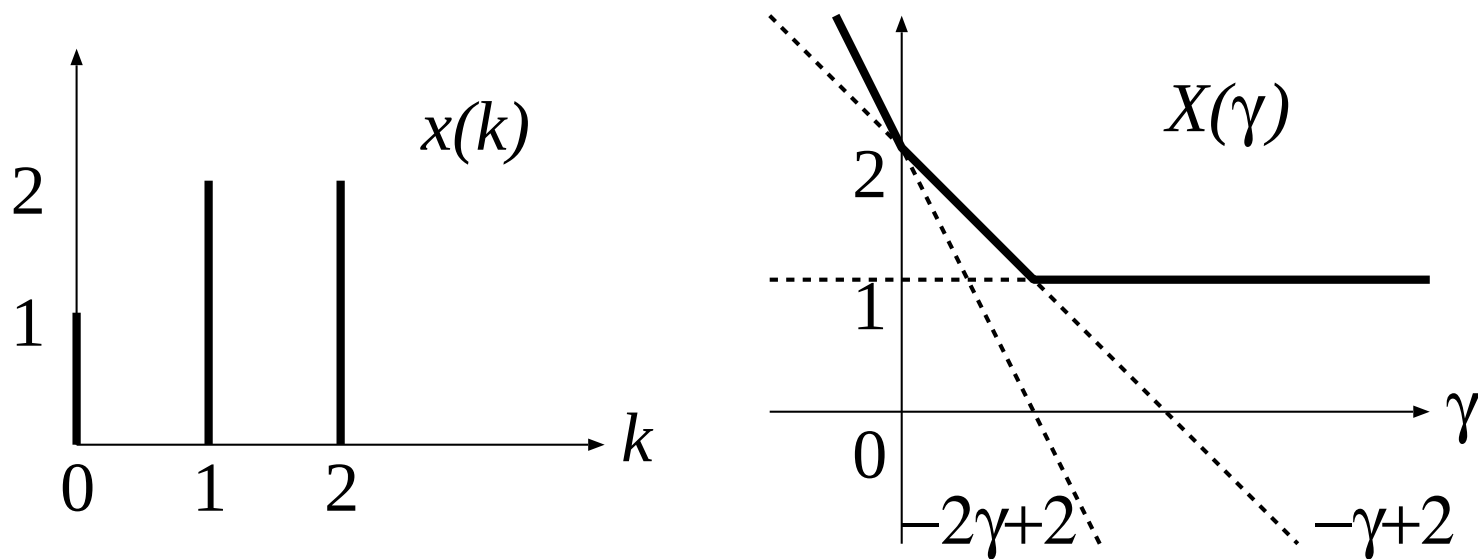
The  **$\gamma$ -transform**  $X(\gamma) \equiv \bigoplus_{i=0}^{\infty} x(i)\gamma^{-i}$  is convenient because we obtain the following representation:

$$Y(\gamma) = H(\gamma) \otimes U(\gamma) = H(\gamma) + U(\gamma)$$

The convolution becomes **multiplication**  $\otimes$ .

## 4. System Theory — Example of $\gamma$ -transform

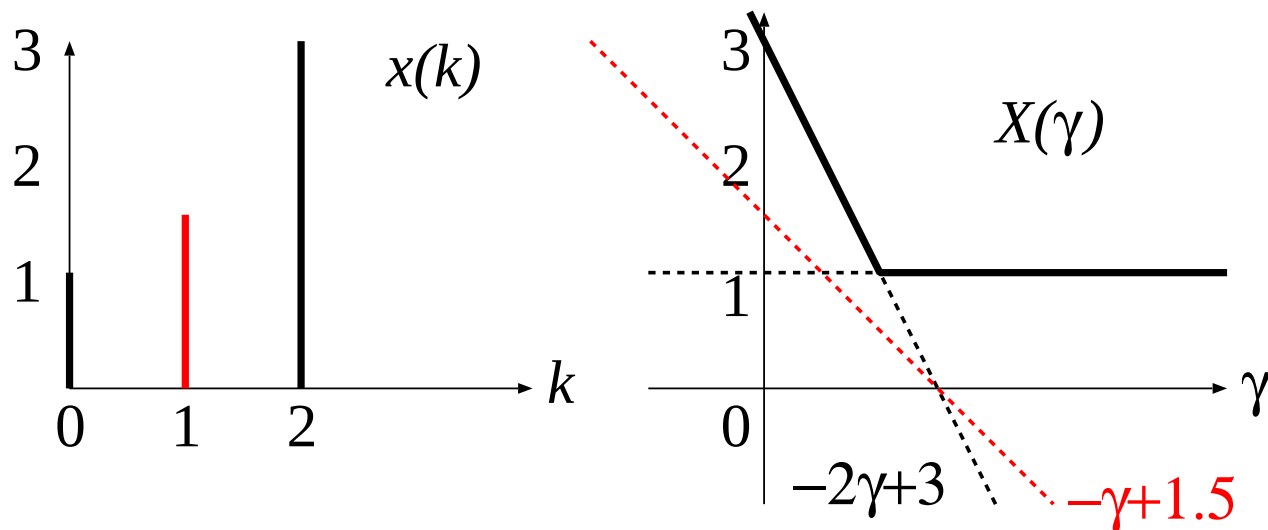
$$\begin{aligned} X(\gamma) &\equiv \bigoplus_k x(k) \gamma^{-k} = 1 \oplus 2\gamma^{-1} \oplus 2\gamma^{-2} \\ &= \max(1, -\gamma + 2, -2\gamma + 2) \\ &= \max_k (-k\gamma + x(k)) \quad : \text{Legendre transform!} \end{aligned}$$



The  $\gamma$ -transform is equal to Legendre transform

#### 4. System Theory — Example of $\gamma$ -transform in nonconcave case

$$\begin{aligned} X(\gamma) &= 1 \oplus 1.5\gamma^{-1} \oplus 3\gamma^{-2} \\ &= \max(1, -\gamma + 1.5, -2\gamma + 3) \end{aligned}$$



The  $x(1)$  is neglected because the function is non-concave and non-convex. In order to remove this drawback, we extend the gendre ( $\gamma$ ) transform.

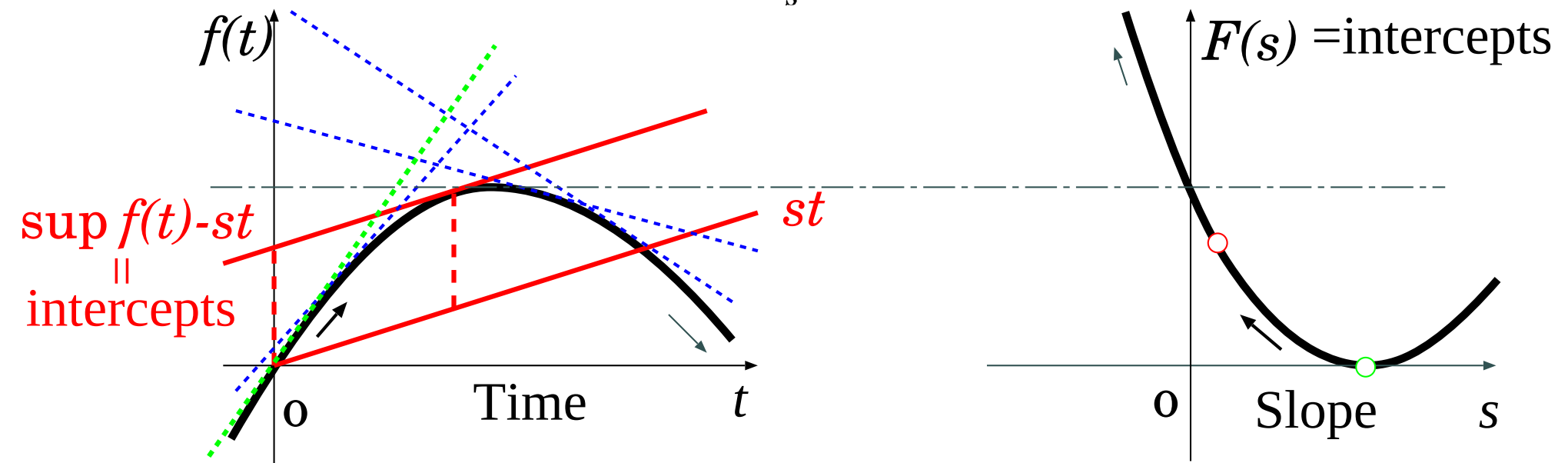
## 5. Legendre transform — Definition

### Legendre transform

$$\mathcal{L}[f(t)](s) = \sup_t (f(t) - st)$$

### Inverse Legendre transform

$$\mathcal{L}^{-1}[F(s)](t) = \inf_s (F(s) + st)$$



**Legendre transform is the description with set of tangent.**

## 5. Legendre transform — Property of the Legendre transform

$$a + f(t) \Leftrightarrow a + \mathcal{L}[f(t)](s)$$

$$f(t - a) \Leftrightarrow \mathcal{L}[f(t)](s) - sa$$

$$f(t) + at \Leftrightarrow \mathcal{L}[f(t)](s - a)$$

$$f(at) \Leftrightarrow \mathcal{L}[f(t)](s/a)$$

$$af(t) \Leftrightarrow a\mathcal{L}[f(t)](s/a)$$

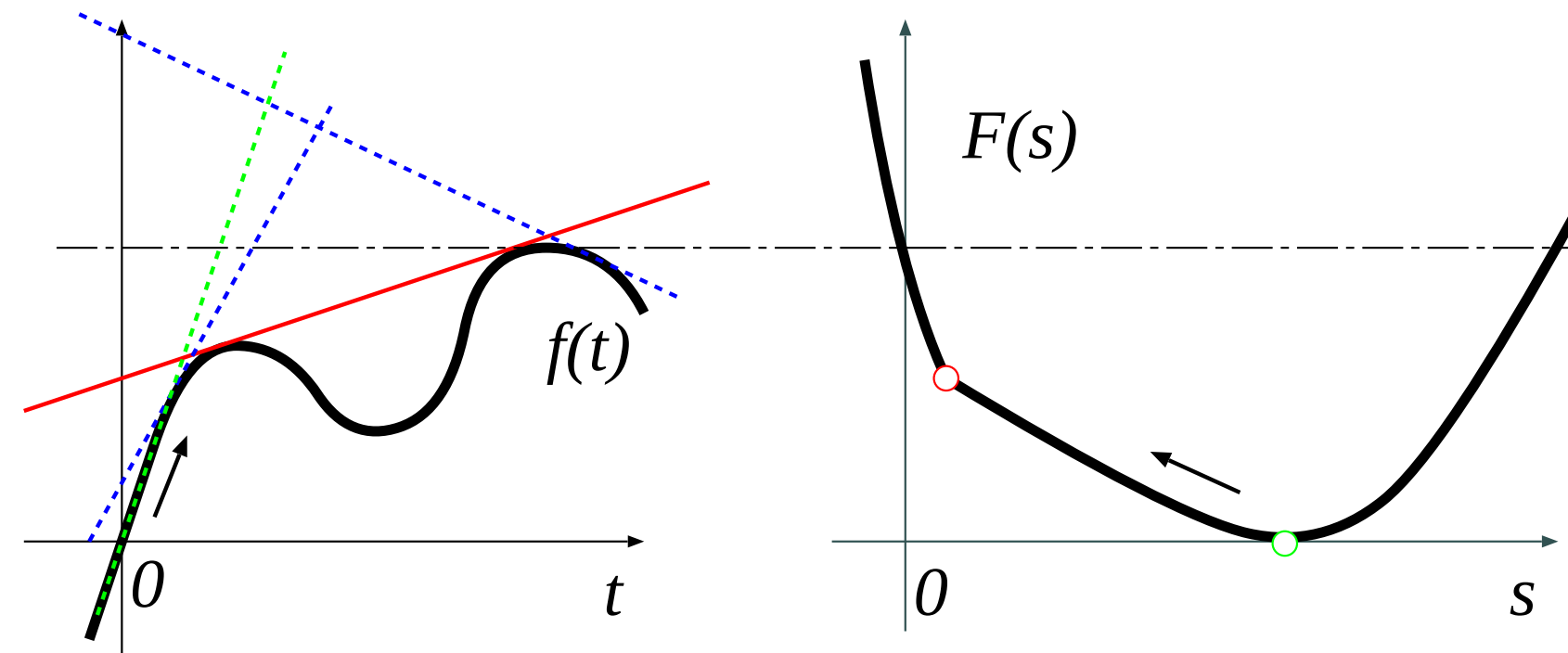
$$\sup_{\tau} (f(\tau) + g(t - \tau)) \Leftrightarrow \mathcal{L}[f(t)](s) + \mathcal{L}[g(t)](s)$$

$$f(t) + g(t) \Leftrightarrow \sup_{\sigma} (\mathcal{L}[f(t)](\sigma) + \mathcal{L}[g(t)](s - \sigma))$$

The above relations are satisfied for convex or concave functions.

## 5. Legendre transform — Non-convex or non-concave function

In the case of non-convex or non-concave function, the Legendre transform is not suitable for the description of function, because **some part of function is neglected**.



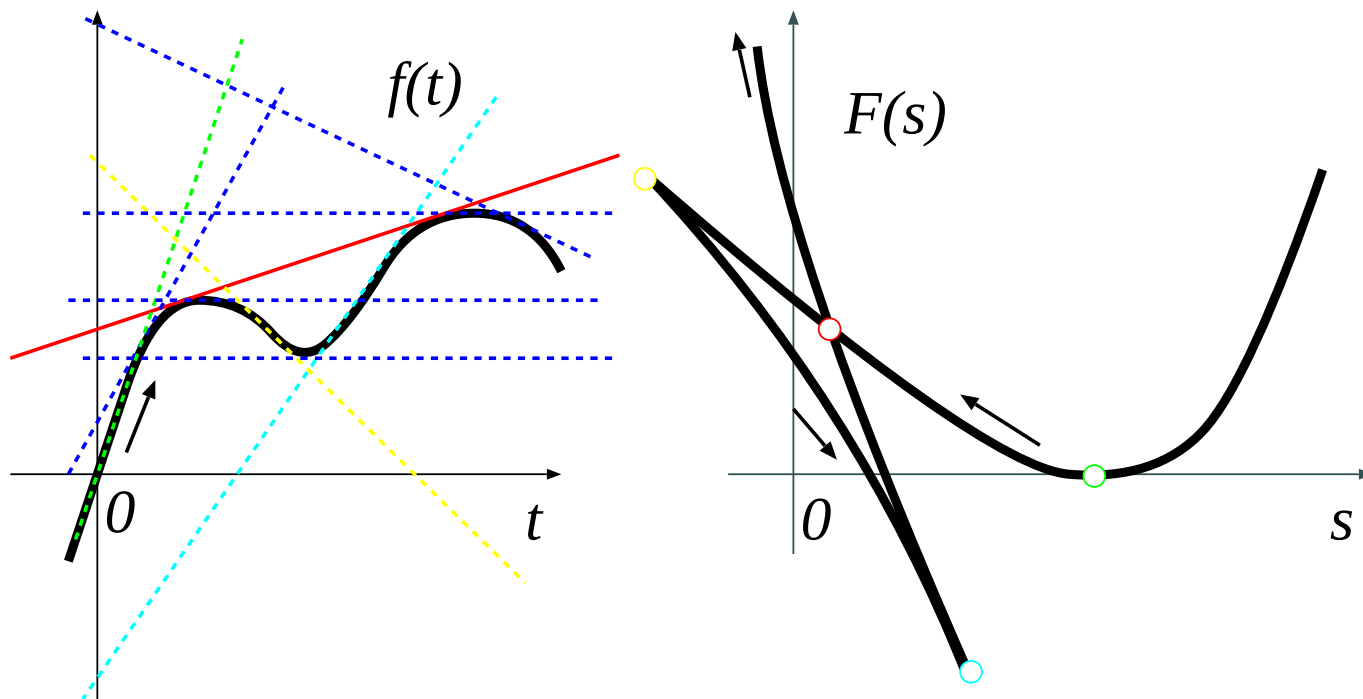
## 5. Legendre transform — Extension of Legendre transform

### extended Legendre transform

$$\mathcal{L}[f(t)](s) = \{f(t^*) - st^* \mid f'(t^*) = s\}$$

### extended inverse Legendre transform

$$\mathcal{L}^{-1}[F(s)](t) = \{F(s^*) + ts^* \mid F'(s^*) = t\}$$



**The complete set of tangent line is considered.**

## 5. Legendre transform — Property of the Legendre transform

$$a + f(t) \Leftrightarrow a + \mathcal{L}[f(t)](s)$$

$$f(t - a) \Leftrightarrow \mathcal{L}[f(t)](s) - sa$$

$$f(t) + at \Leftrightarrow \mathcal{L}[f(t)](s - a)$$

$$f(at) \Leftrightarrow \mathcal{L}[f(t)](s/a)$$

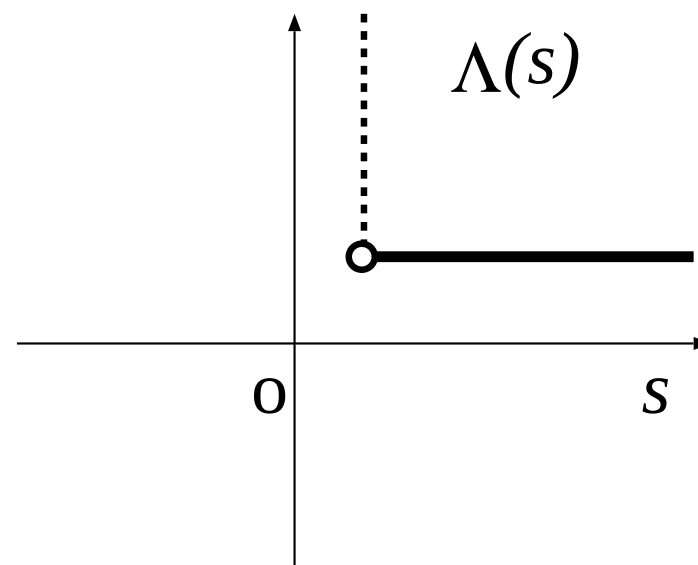
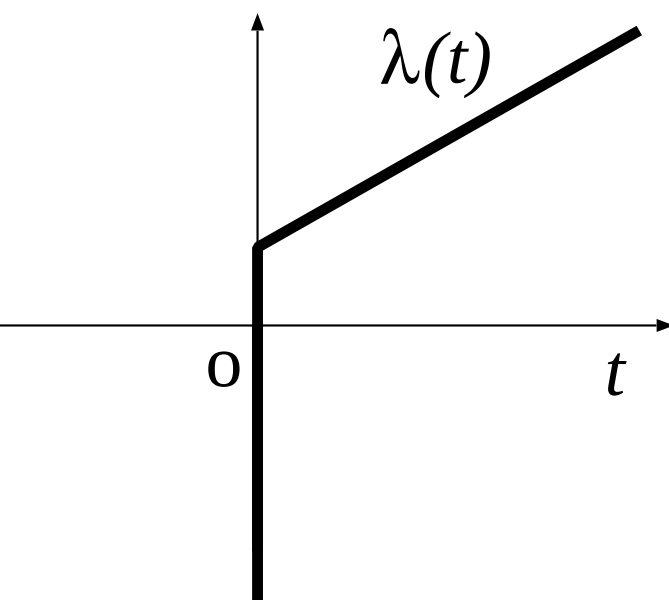
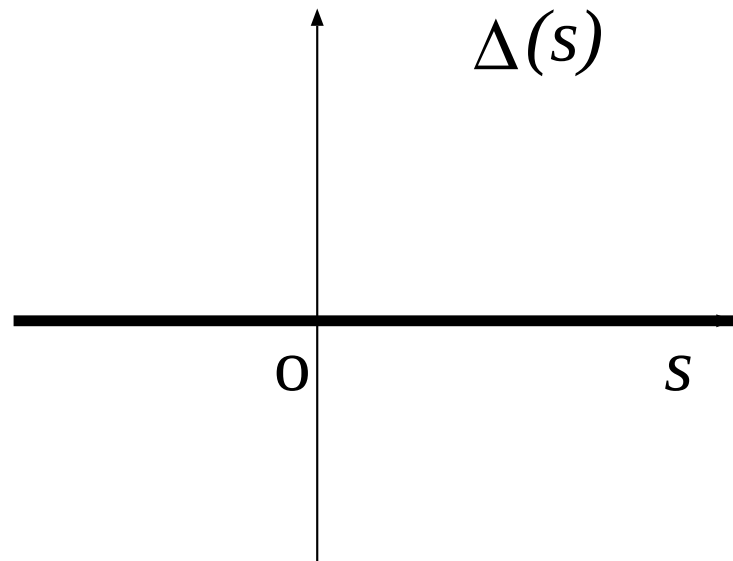
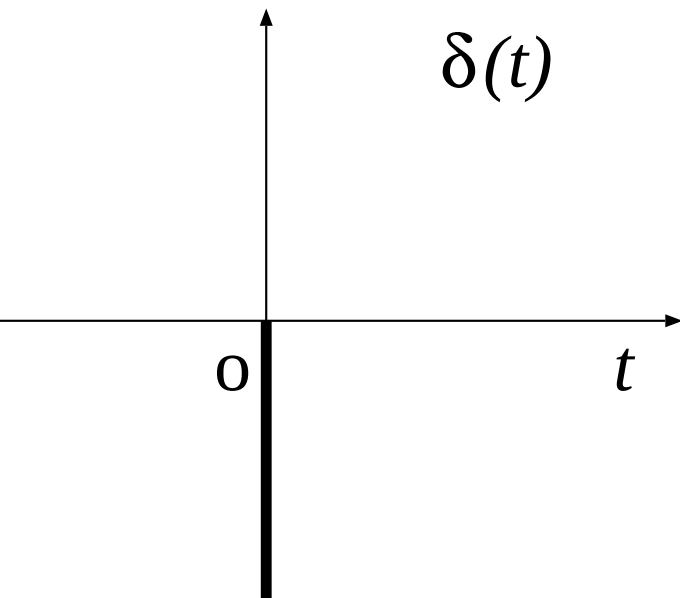
$$af(t) \Leftrightarrow a\mathcal{L}[f(t)](s/a)$$

$$\sup_{\tau} (f(\tau) + g(t - \tau)) \Leftrightarrow \mathcal{L}[f(t)](s) + \mathcal{L}[g(t)](s)$$

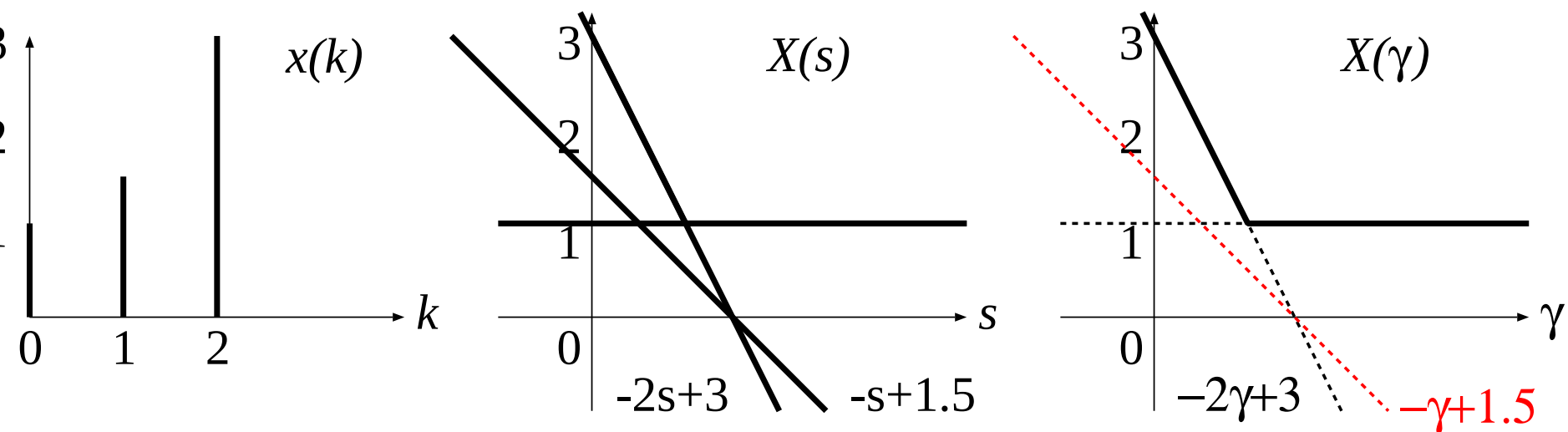
$$f(t) + g(t) \Leftrightarrow \sup_{\sigma} (\mathcal{L}[f(t)](\sigma) + \mathcal{L}[g(t)](s - \sigma))$$

The above relations are satisfied, even if functions are non-convex and non-concave.

## 5. Legendre transform — Examples of Legendre transform

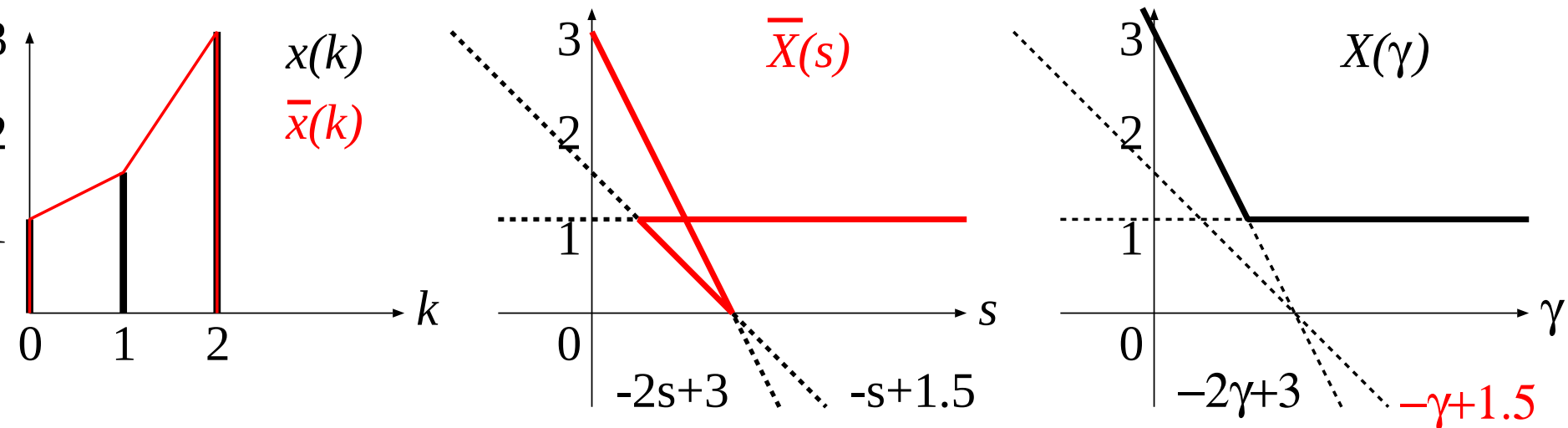


## 5. Legendre transform — Non-convex and non-concave case



The drawback is eliminated. However, the function is very complicated.

## 5. Legendre transform — Piecewise linear interpolation



Interpolation is effective for simplification.

For time series analysis, we apply **the extended Legendre transform with interpolation** instead of  $\gamma$  transform.

## 6. Application — Network analysis with Legendre transform

We consider the following max-plus linear network system.

$$y(k) = \bigoplus_{i=-\infty}^k h(k-i) \otimes u(i),$$

$u(k)$  : input trace  
 $h(k)$  : impulse response  
 $y(k)$  : output trace

1. Calculate **interpolation**  $\bar{h}(t)$  and  $\bar{u}(t)$ .
2. Apply **the extended Legendre transform** to  $\bar{h}(t)$  and  $\bar{u}(t)$  and obtain  $\bar{H}(s)$  and  $\bar{U}(s)$ .
3. Calculate  $\bar{Y}(s) = \bar{H}(s) + \bar{U}(s)$  with **addition**.
4. Obtain  $\bar{y}(k)$  using **the extended inverse Legendre transform**.

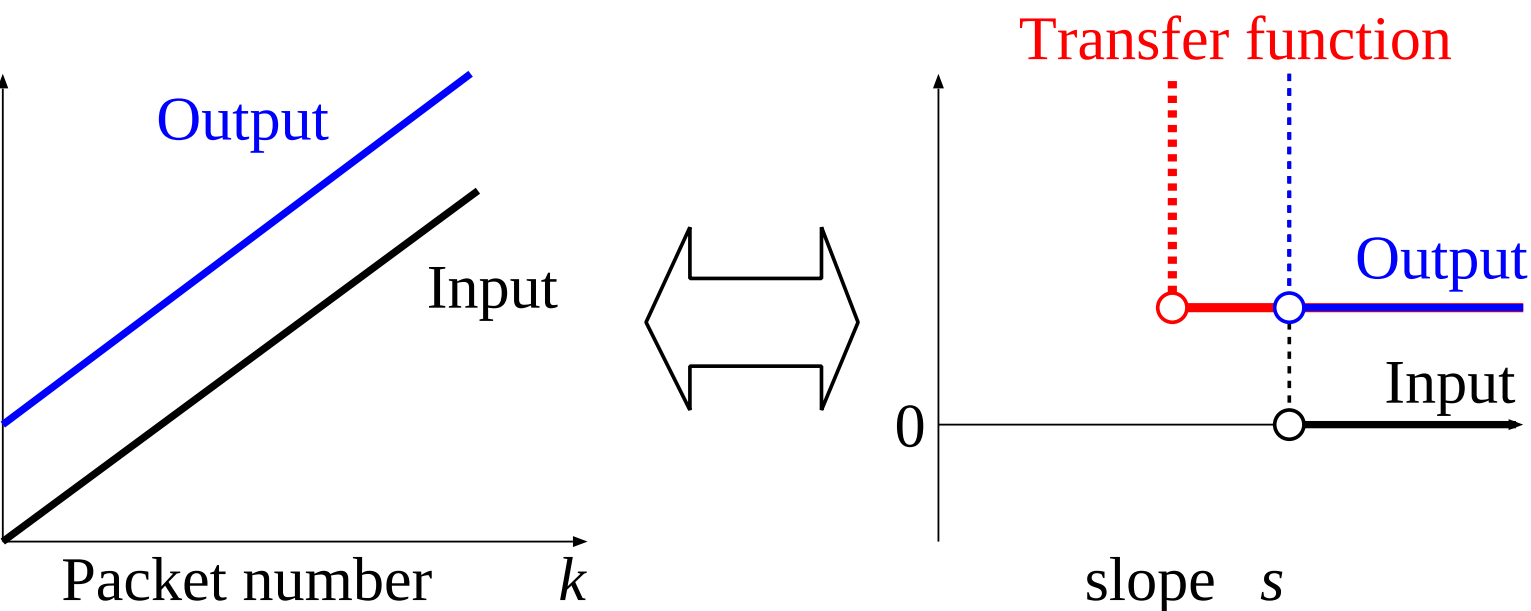
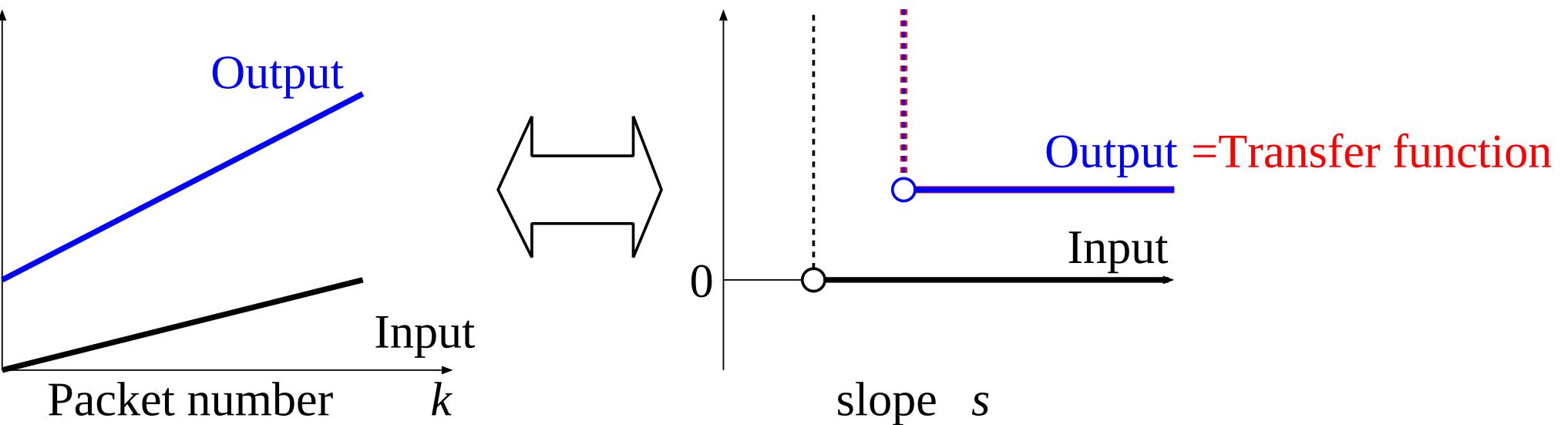
**The fast Legendre transform** is available.

## 6. Application — Identification of network system

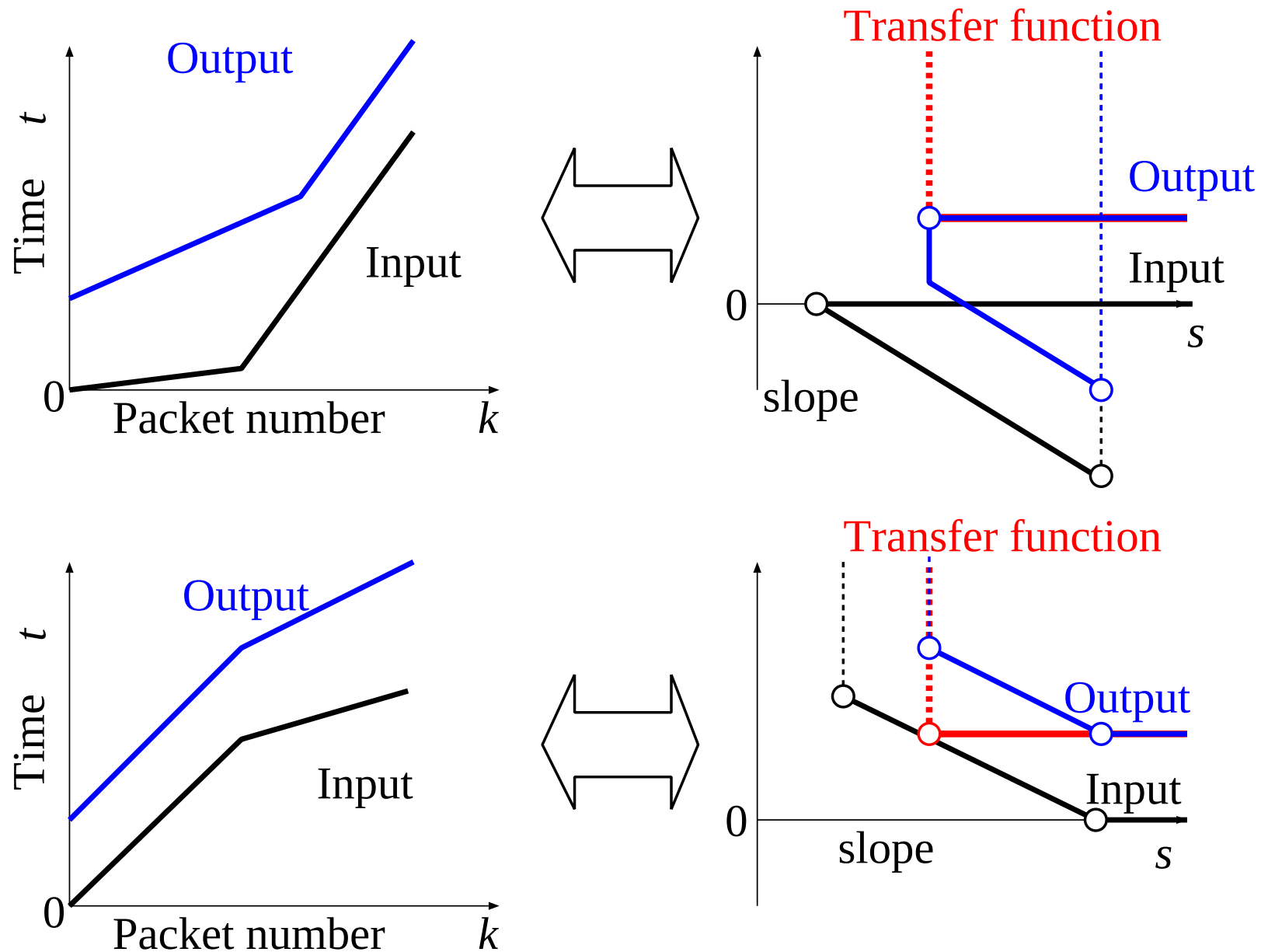
For **traffic analysis**, we need impulse response of **transfer function**. In order to obtain the transfer function, we use **the system identification with deconvolution**.

1. Obtain the input trace  $u(k)$  and output trace  $y(k)$
2. Calculate interpolation  $\bar{y}(t)$  and  $\bar{u}(t)$ .
3. Apply the extended Legendre transform to  $\bar{u}(t)$  and  $\bar{y}(t)$  and obtain  $\bar{U}(s)$  and  $\bar{Y}(s)$ .
4. Calculate the transfer function  $\bar{H}(s) = \bar{Y}(s) - \bar{U}(s)$ . This is the **deconvolution process**.
5. Obtain impulse response  $\bar{h}(t)$  using the extended inverse Legendre transform.

## 6. Application — Transfer function of Bottleneck router



## 6. Application — Representation of traffic change



## 7. Conclusion

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- TCP is **linear** in max-plus algebra.
- The output is given by the **convolution** of input and transfer function.
- **The extended Legendre transform** in max-plus algebra corresponds to Fourier transform and  $z$ -transform in linear system theory.
- The **deconvolution** is given by the difference of the transformed input and output.
- Thus, **we can identify the transfer function of network system using the extended Legendre transform.**

### **3. Future work**

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- **We have to confirm the effectiveness using simulation.**
- **We can try to apply this method to any protocols and any networks.**

## References

- [ ] **F.Baccelli, G.Cohen, G.J.Olsder, and J.P.Quadrant:** *Synchronization and Linearity*, Wiley, 1992.
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- [ ] **F.Baccelli and D. Hong :** “AIMD, Fairness and Fractal Scaling of TCP Traffic,” Proc. of IEEE INFOCOM, 2002.
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- [ ] **L.Dorst and R. van den Boomgaard:**”Morphological Signal Processing and the Slope Transform,” Signal Processing, Vol.38, pp.79–98, 1994.
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## 10. Appendix — Why Legendre transform?

### Conventional Fourier transform

$$F(s) = \int f(t)e^{-st} dt \quad (1)$$

$e^{-st}$  is based on characters of the group:

$$f(x + y) = f(x)f(y).$$

Max-plus algebra analog is

$$f(x + y) = f(x) + f(y).$$

Max-plus characters of that group are  $-s \cdot x = \log e^{-st}$

Fourier transform in max-plus algebra is

$$F(s) = \max(f(t) - s \cdot t) \quad (2)$$

**That is Legendre transform!**