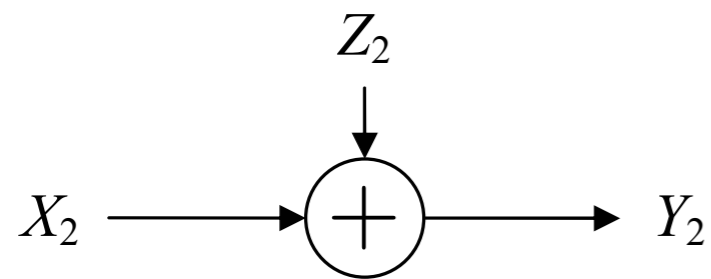
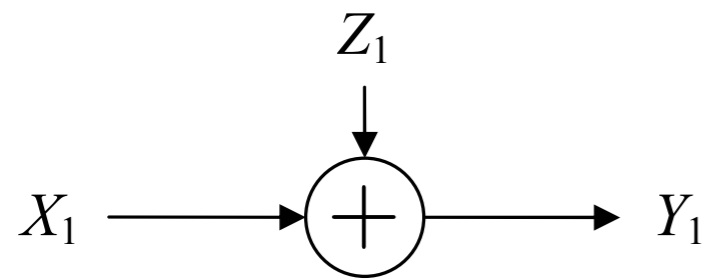


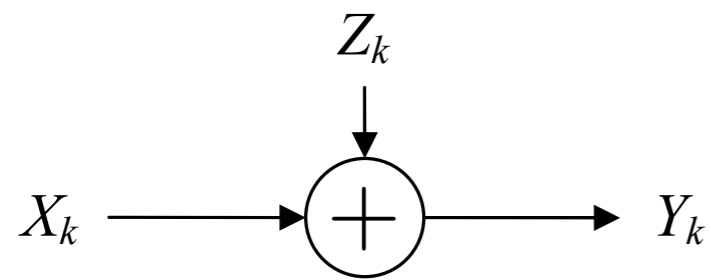


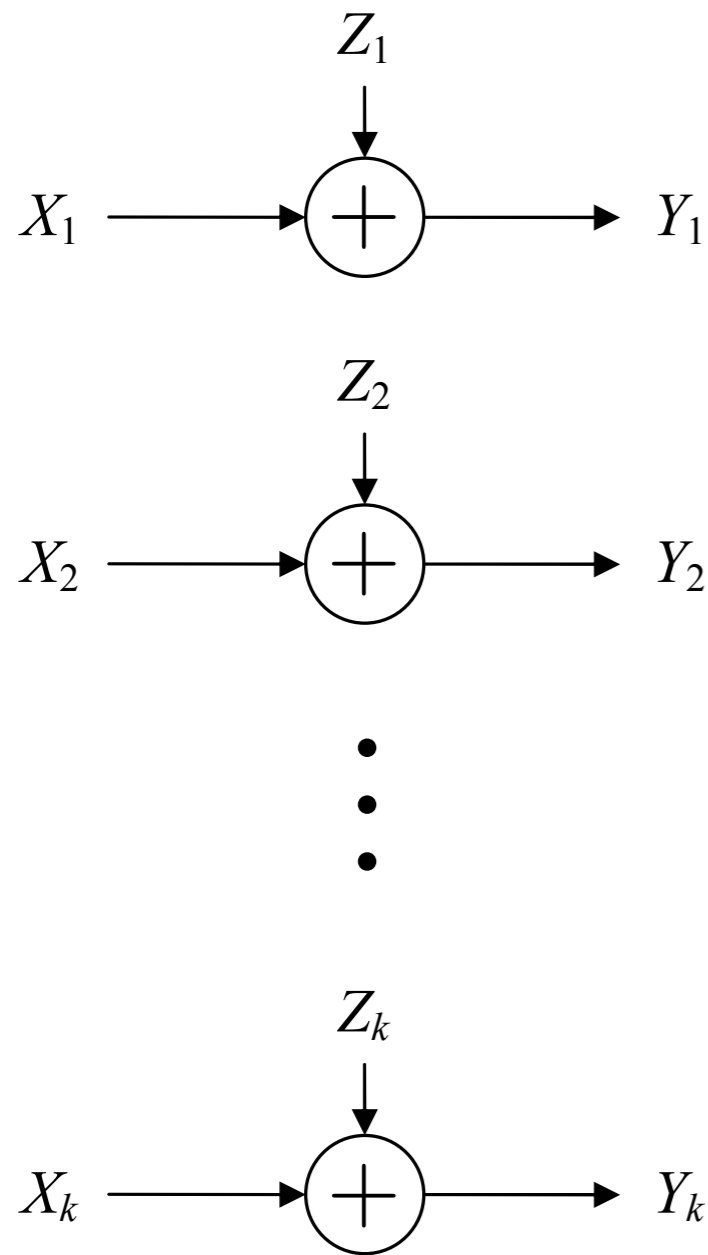
香港中文大學
The Chinese University of Hong Kong

11.5 Parallel Gaussian Channels

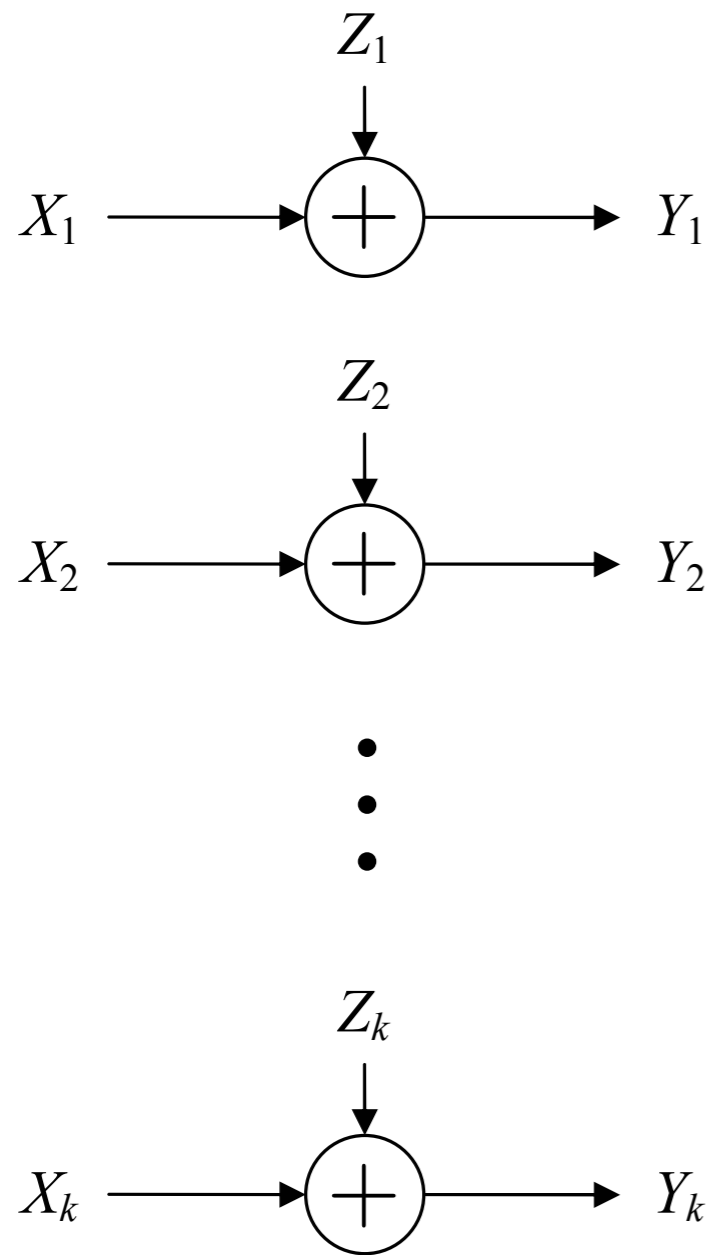


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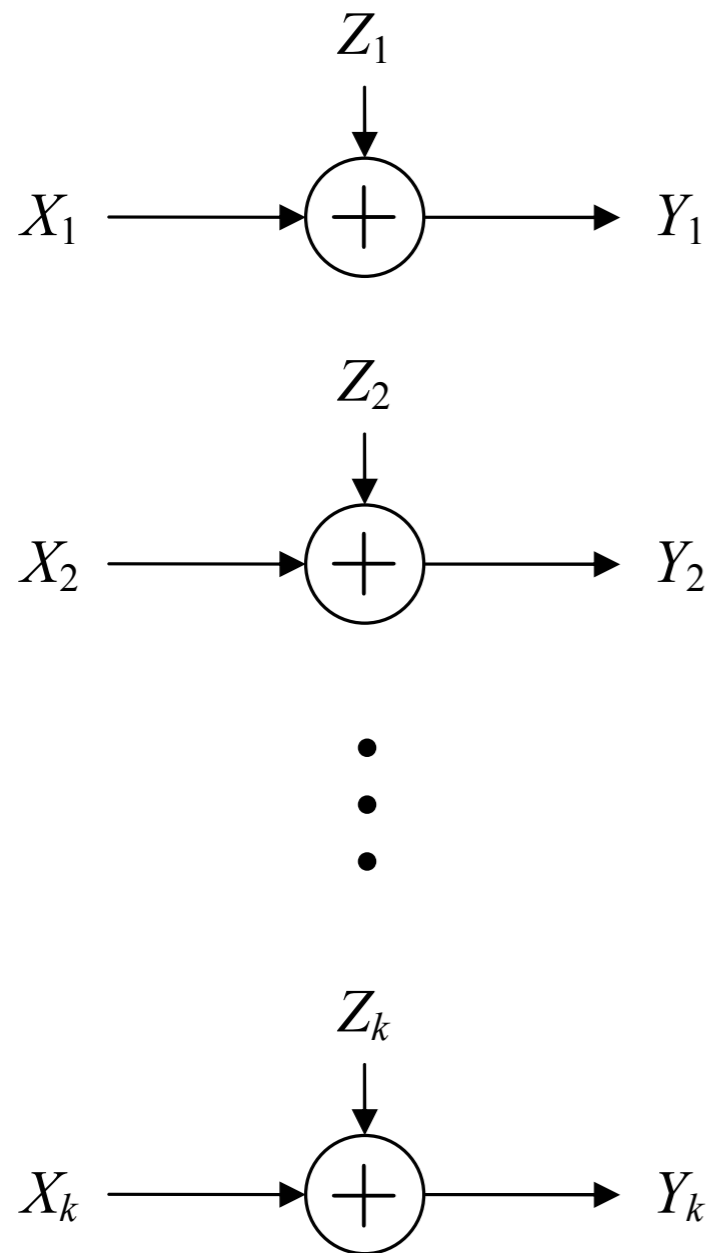




- $Z_i \sim \mathcal{N}(0, N_i)$ and $Z_i, 1 \leq i \leq k$ are independent.



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- Total input power constraint: $E \sum_{i=1}^k X_i^2 \leq P$.

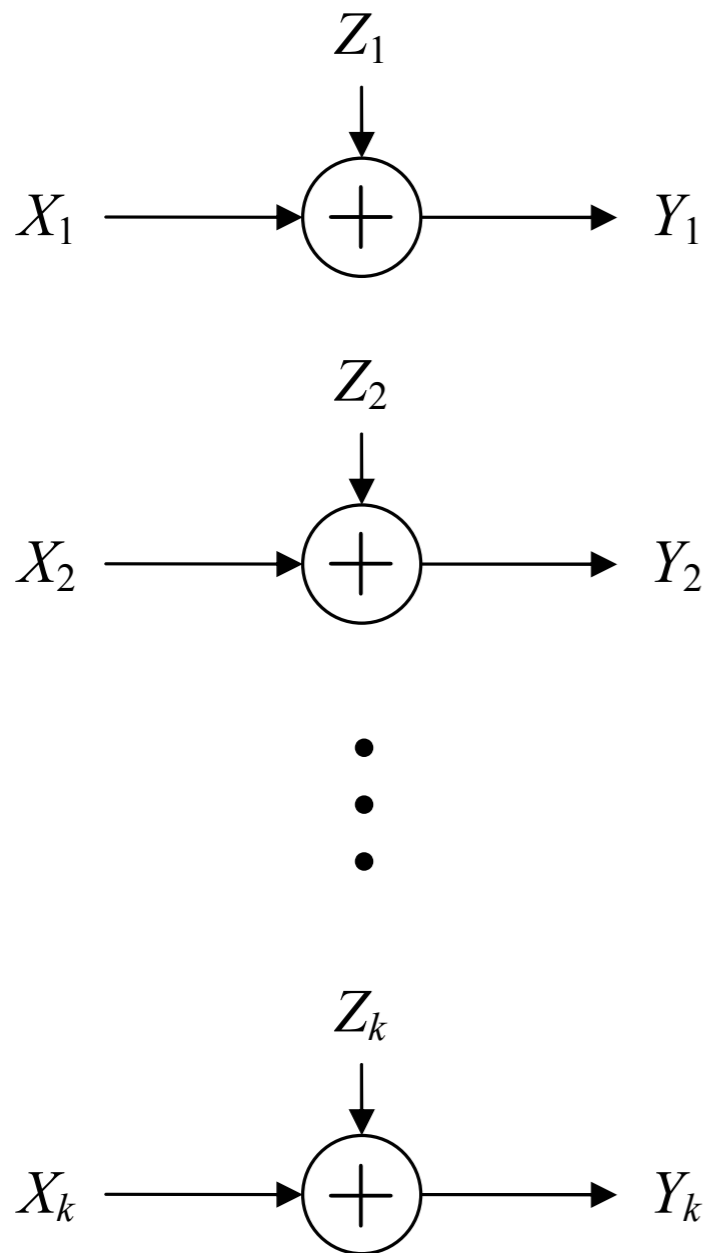


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-

$$C(P) = \sup_{F(\mathbf{x}): E \sum_i X_i^2 \leq P} I(\mathbf{X}; \mathbf{Y})$$



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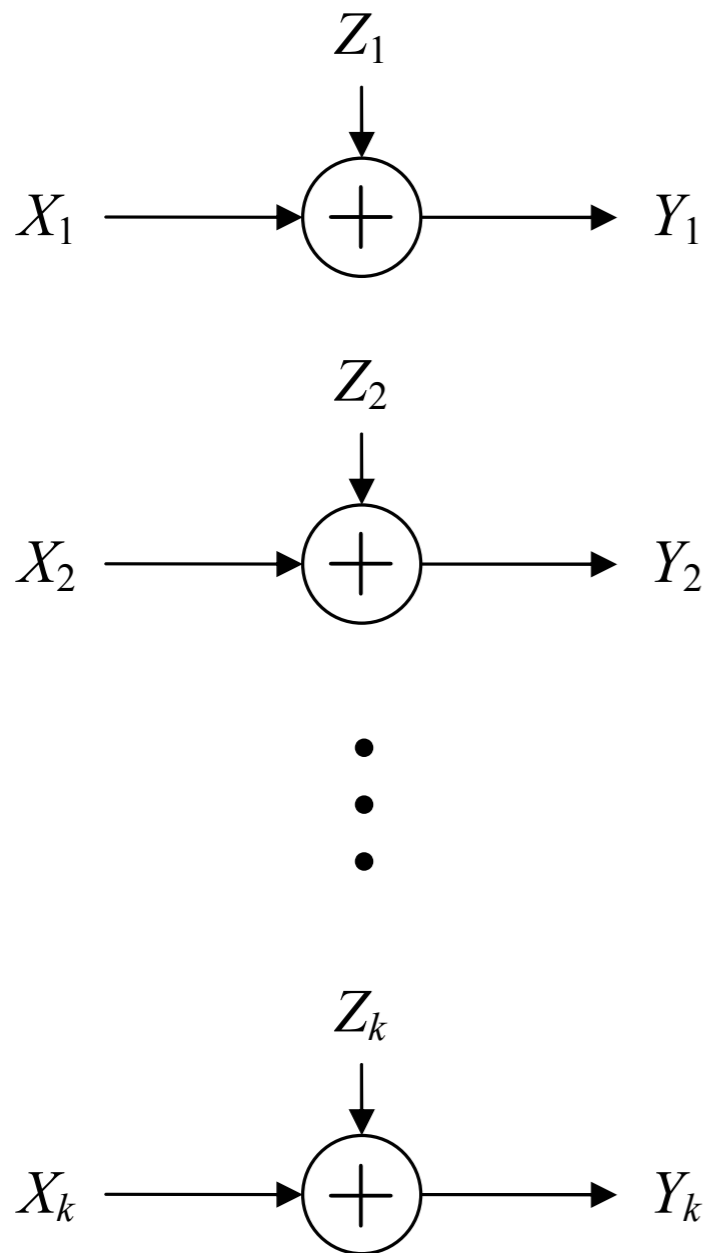
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$$C(P) = \sup_{F(\mathbf{x}): E \sum_i X_i^2 \leq P} I(\mathbf{X}; \mathbf{Y})$$

- Intuitively,

$$C(P) = \max_{P_1, P_2, \dots, P_k: \sum_i P_i = P} \frac{1}{2} \sum_{i=1}^k \log \left(1 + \frac{P_i}{N_i} \right)$$



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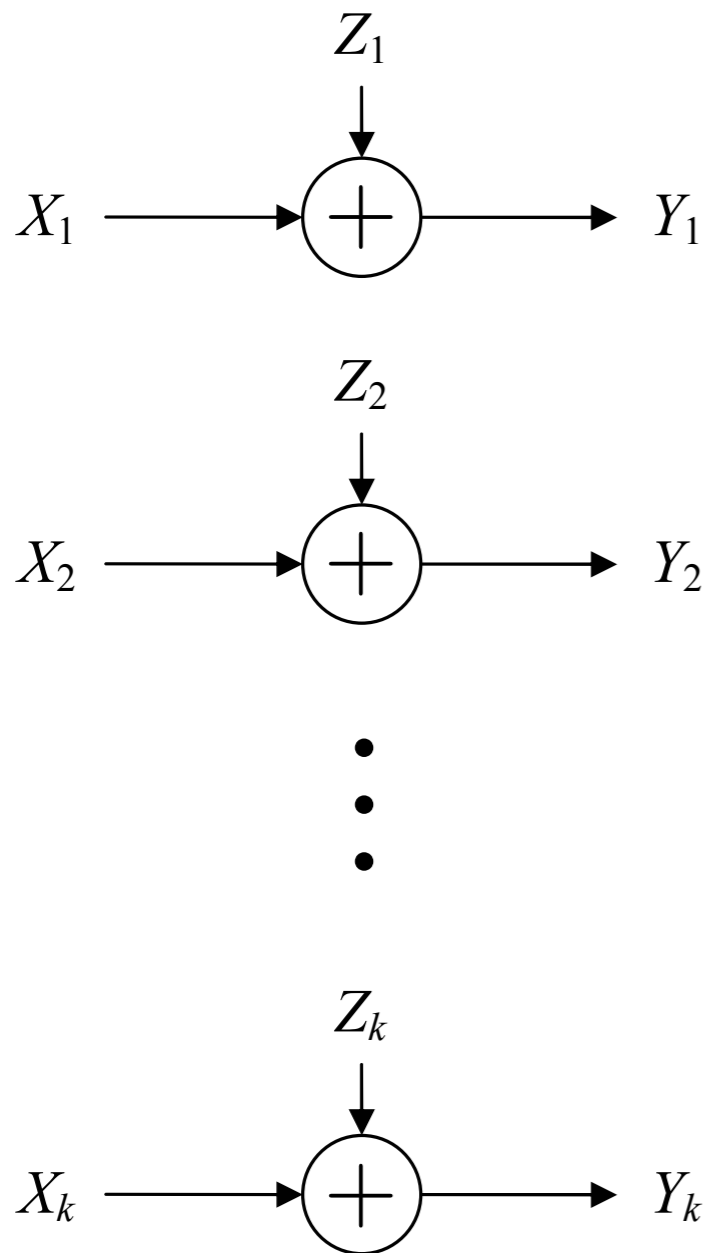
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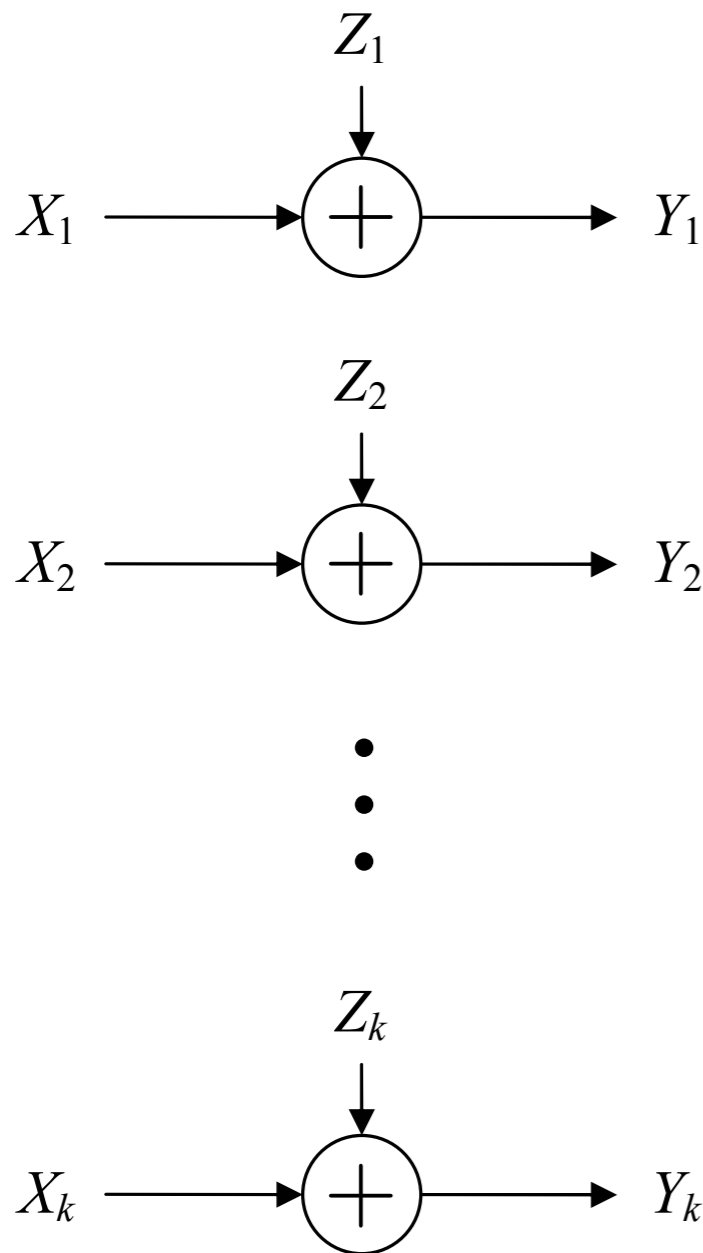
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$$\frac{1}{2} \log \left(1 + \frac{P_i}{N_i} \right)$$

is the capacity of the i th channel when the input power is P_i .



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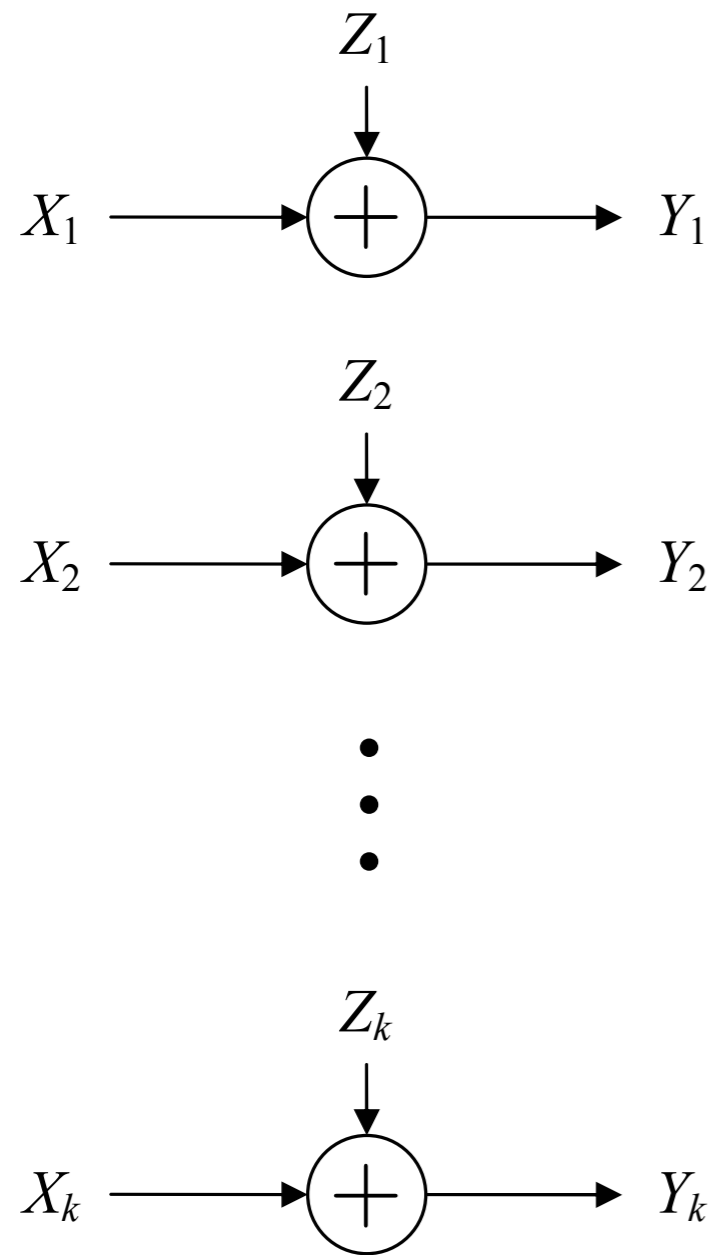
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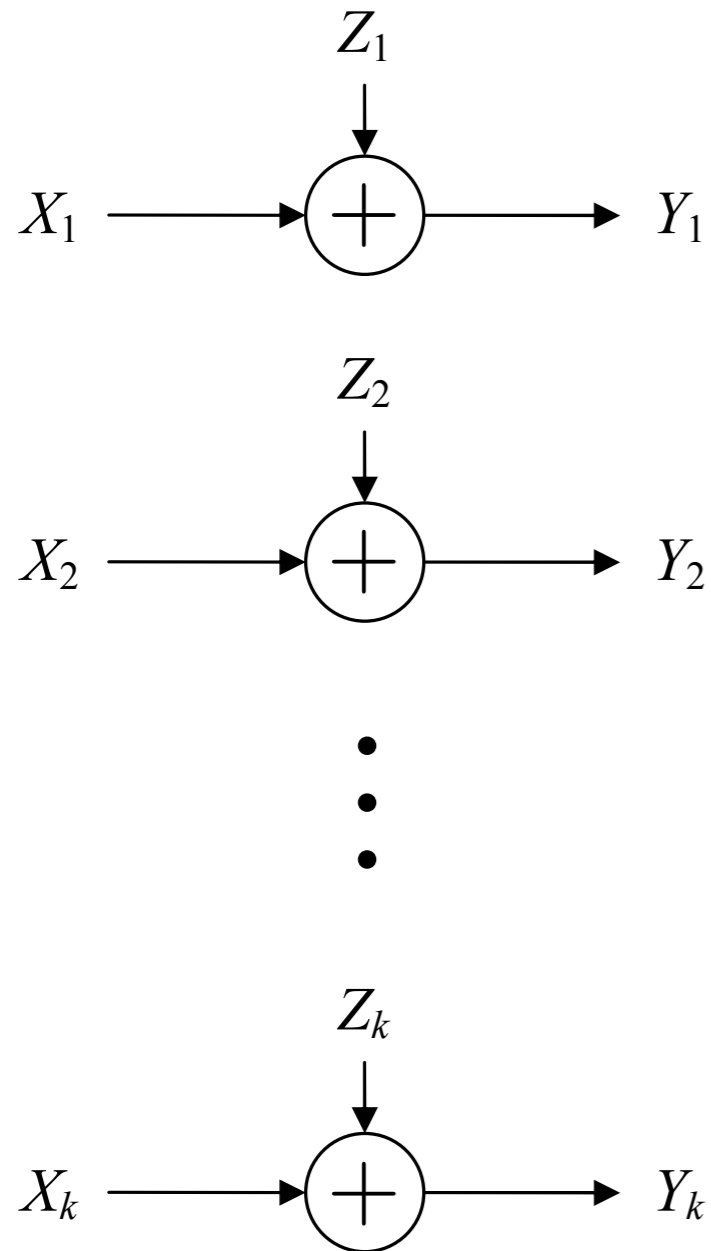
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Formal Justification:



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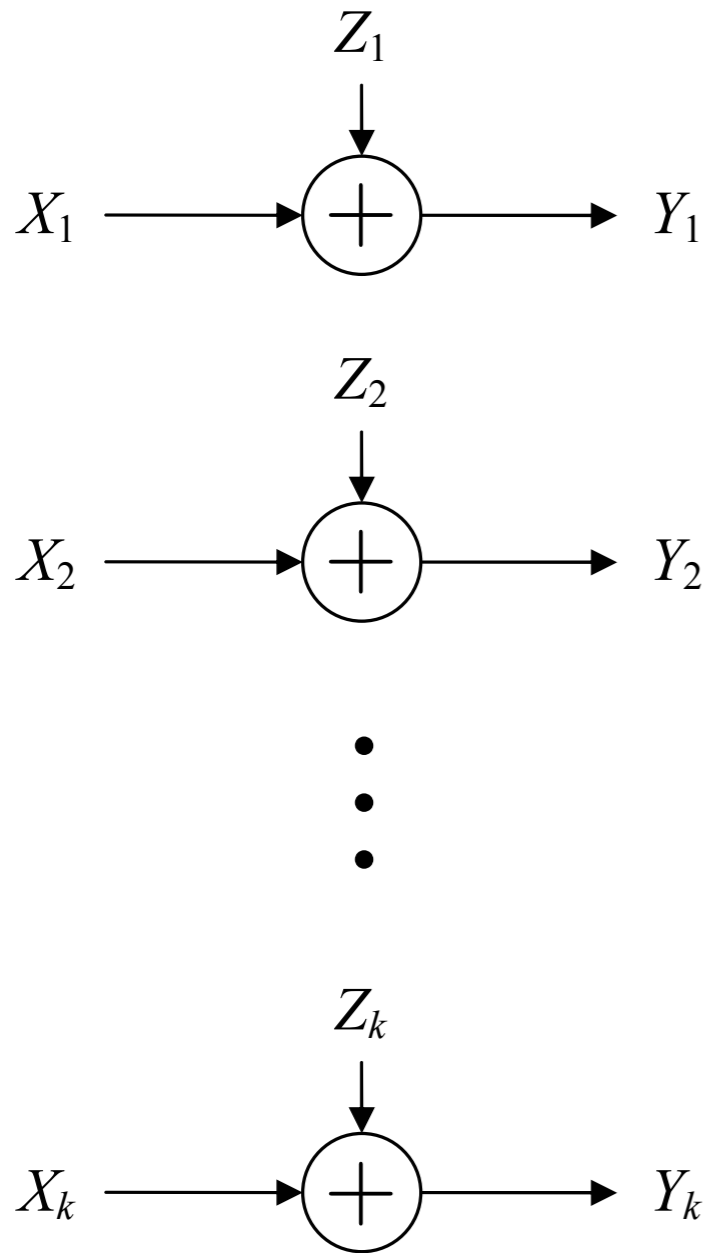
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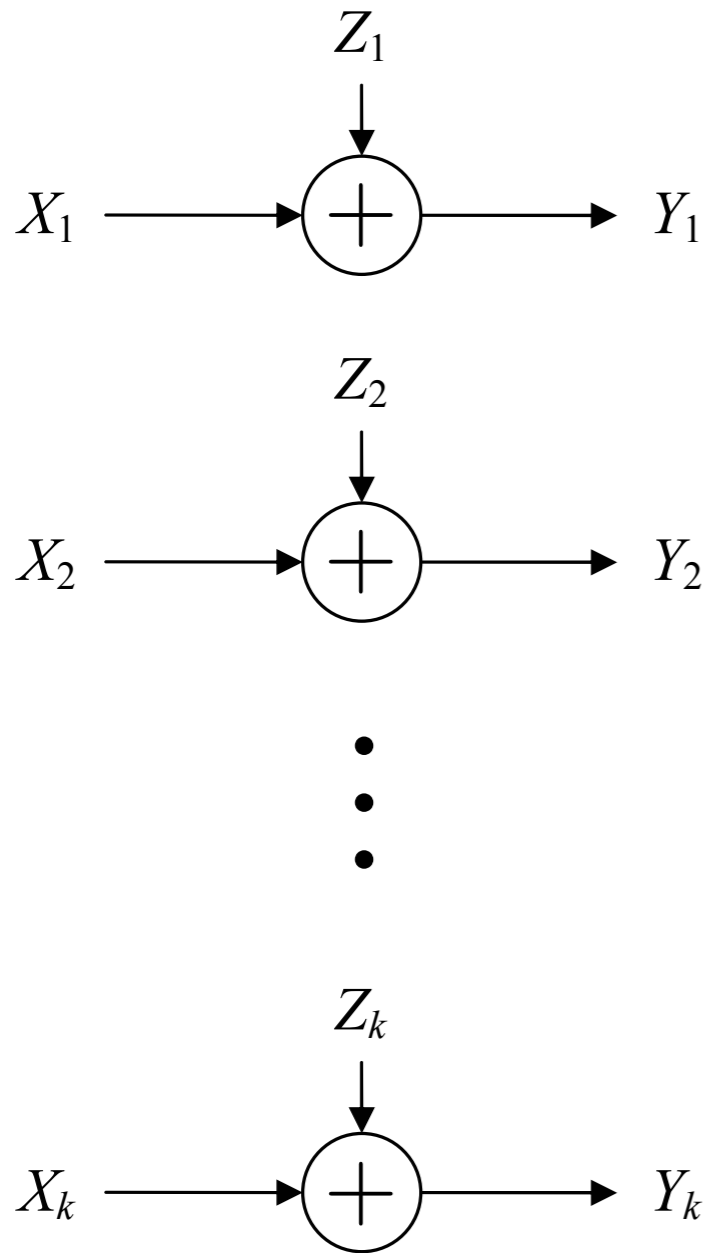
$$I(\mathbf{X}; \mathbf{Y}) = h(\mathbf{Y}) - h(\mathbf{Z})$$



Formal Justification:

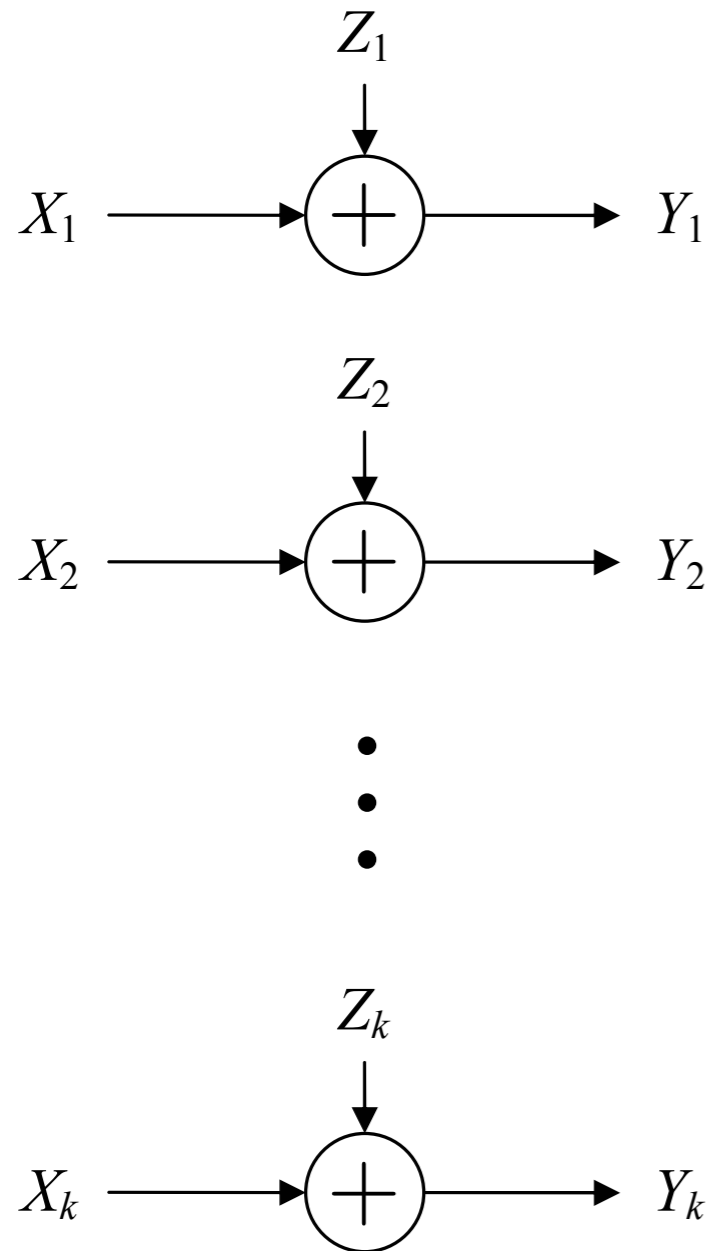
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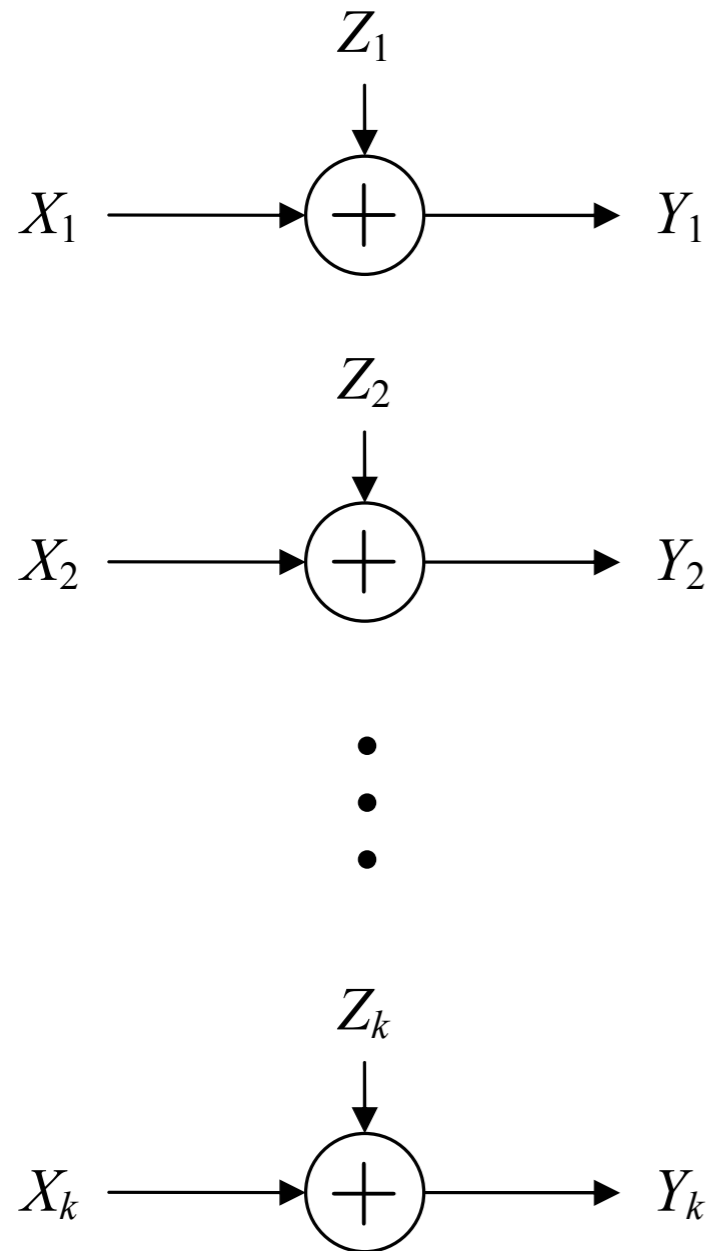
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$$\begin{aligned} I(\mathbf{X}; \mathbf{Y}) &= \underline{h(\mathbf{Y})} - h(\mathbf{Z}) \\ &\leq \underline{\sum_{i=1}^k h(Y_i)} - \sum_{i=1}^k h(Z_i) \end{aligned} \tag{1}$$

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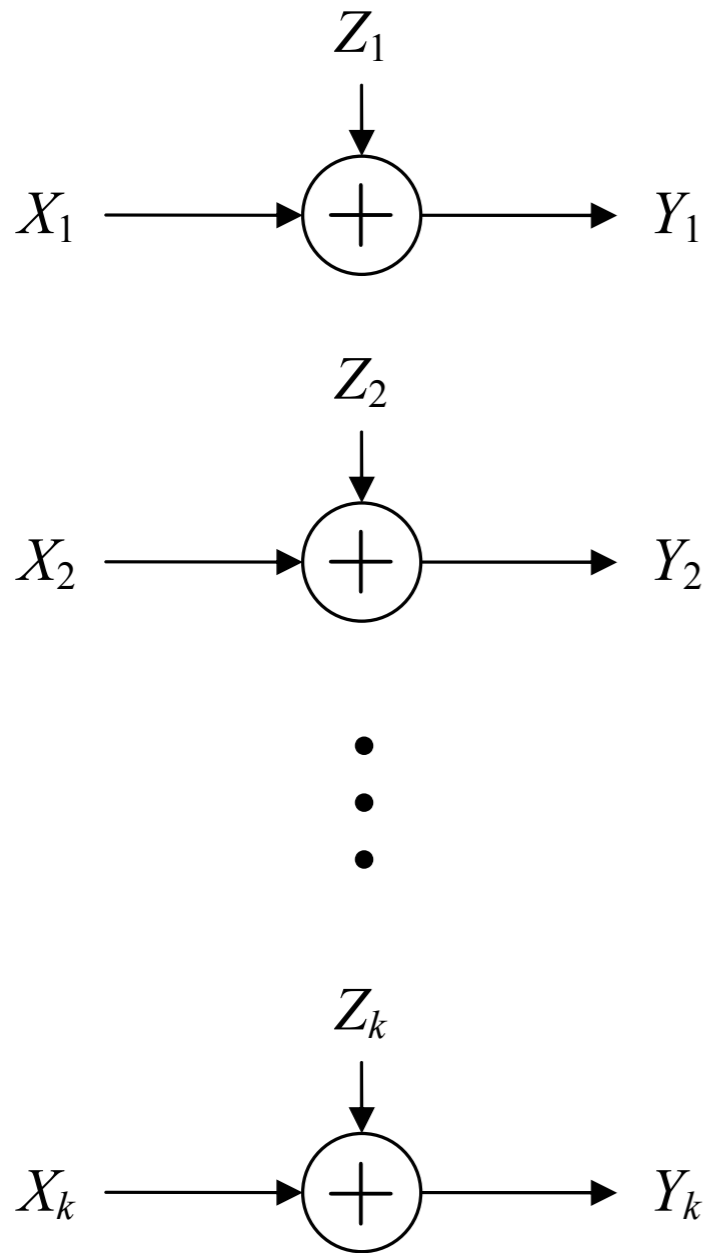
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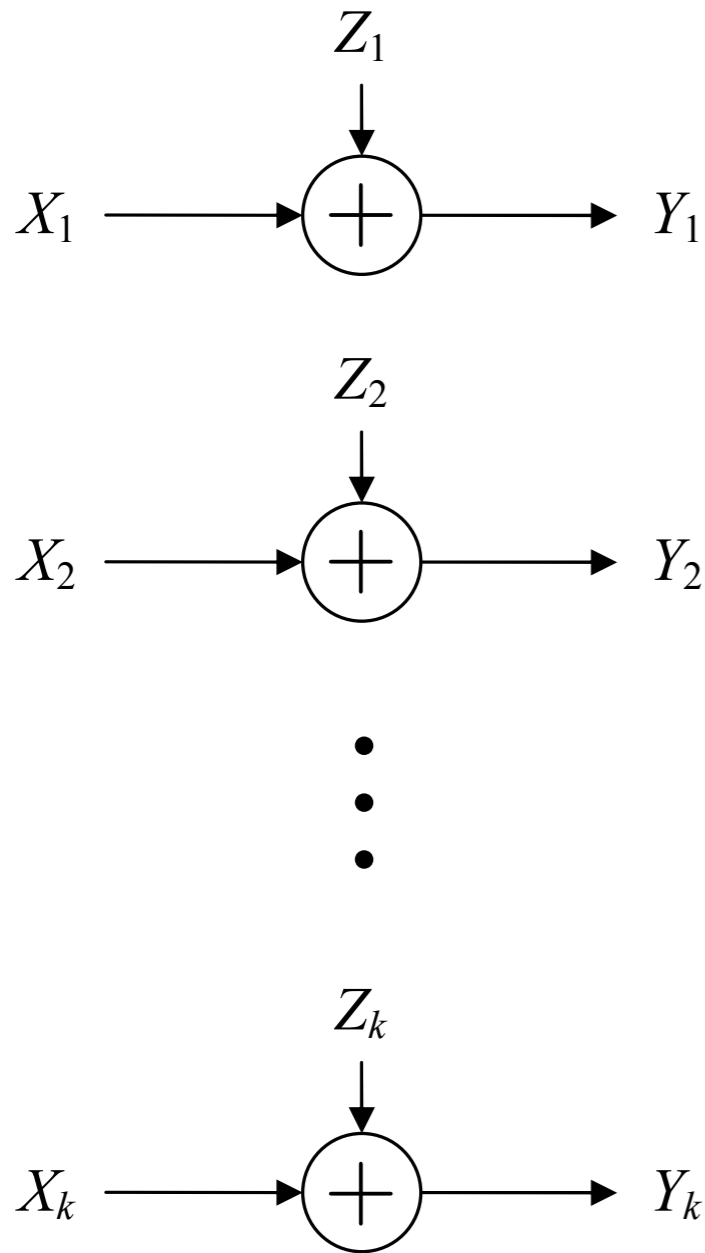
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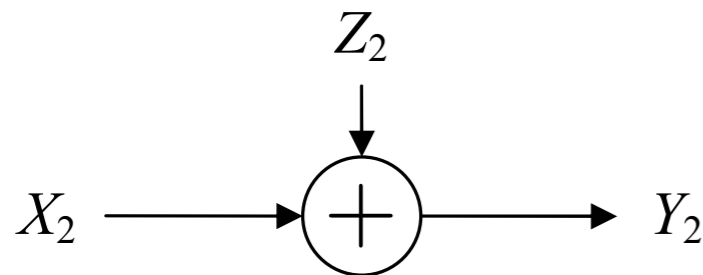
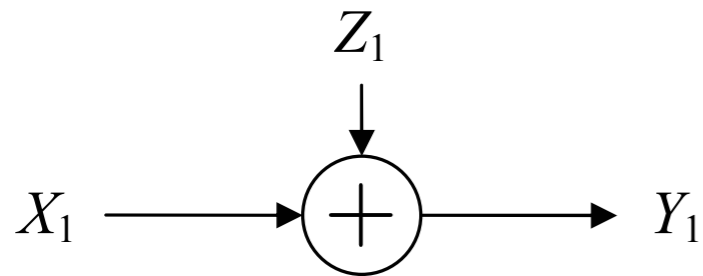
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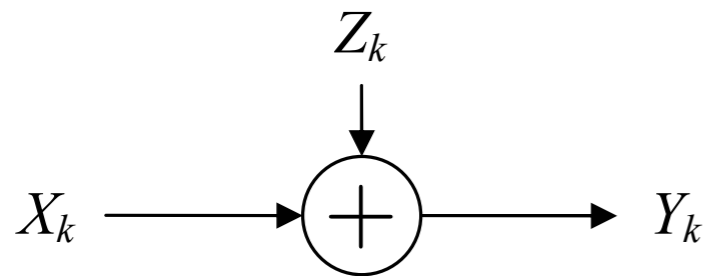
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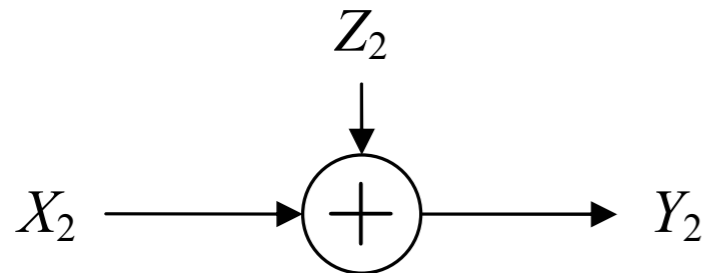
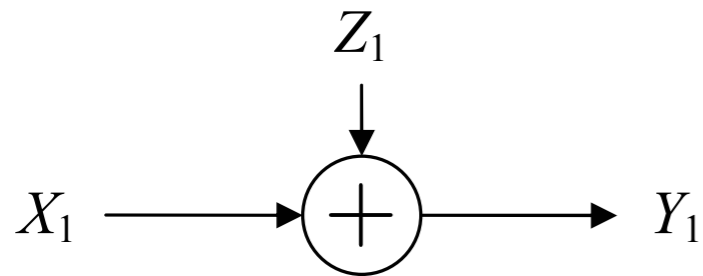


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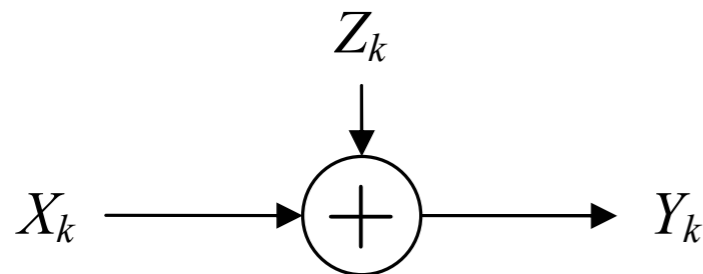
$$\leq \frac{1}{2} \sum_{i=1}^k \underline{\log[2\pi e(EY_i^2)]} - \frac{1}{2} \sum_{i=1}^k \log(2\pi eN_i) \quad (2)$$

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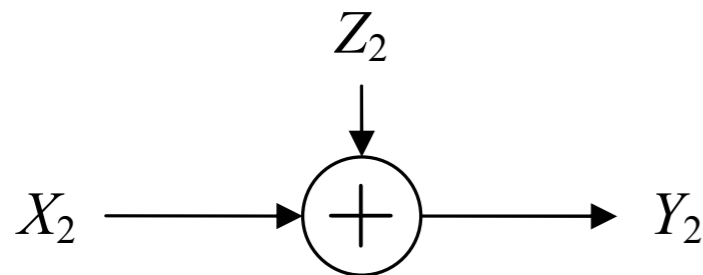
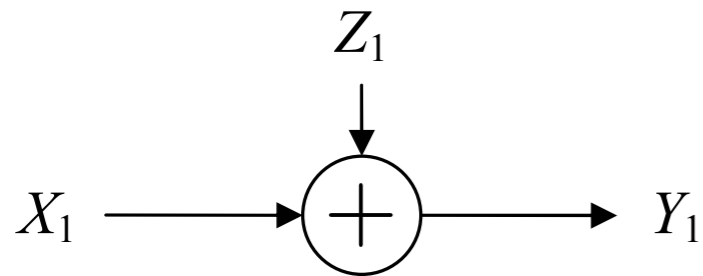


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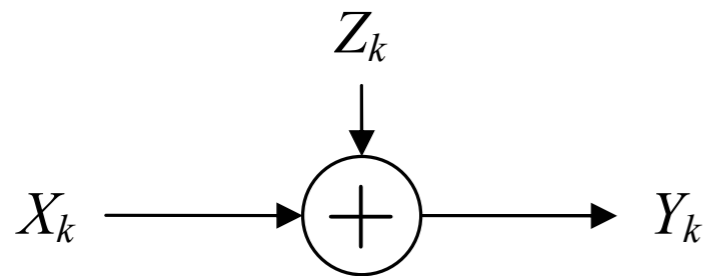
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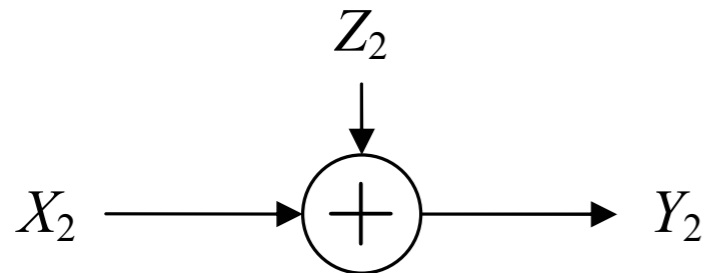
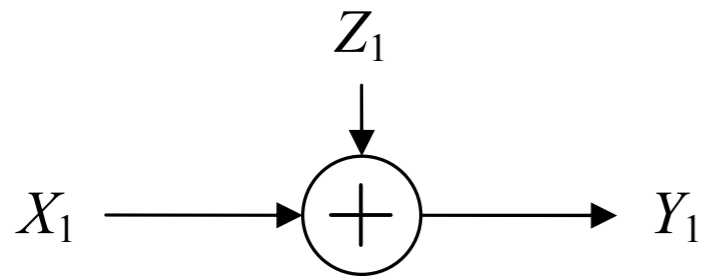


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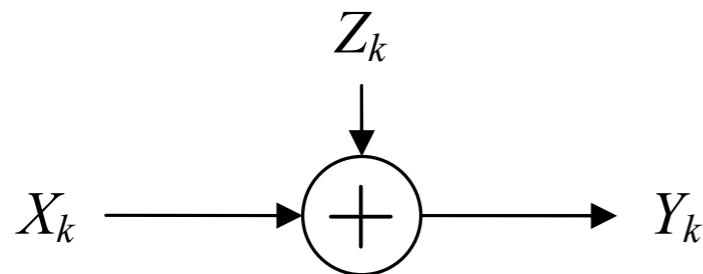
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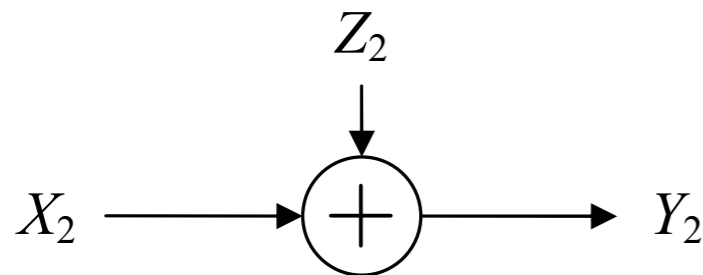
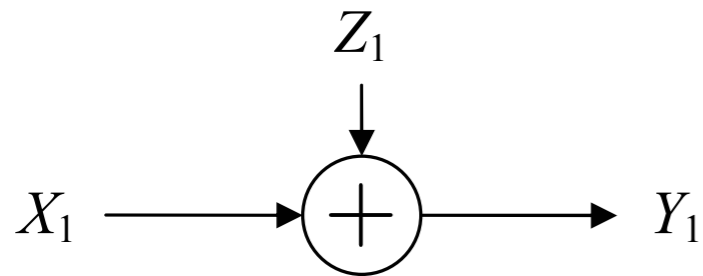


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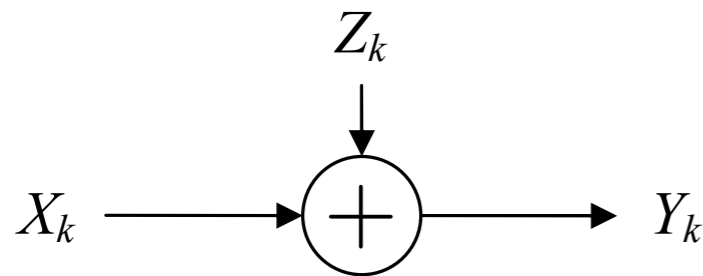
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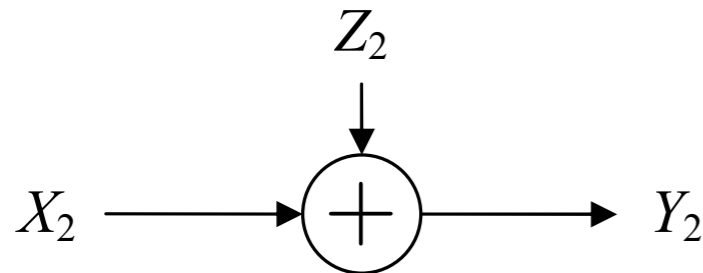
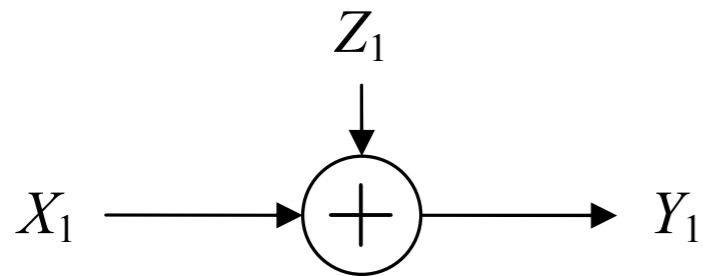


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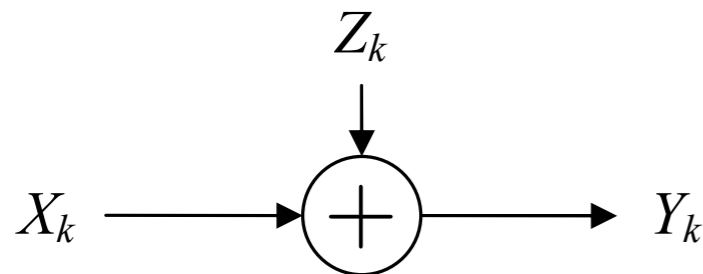
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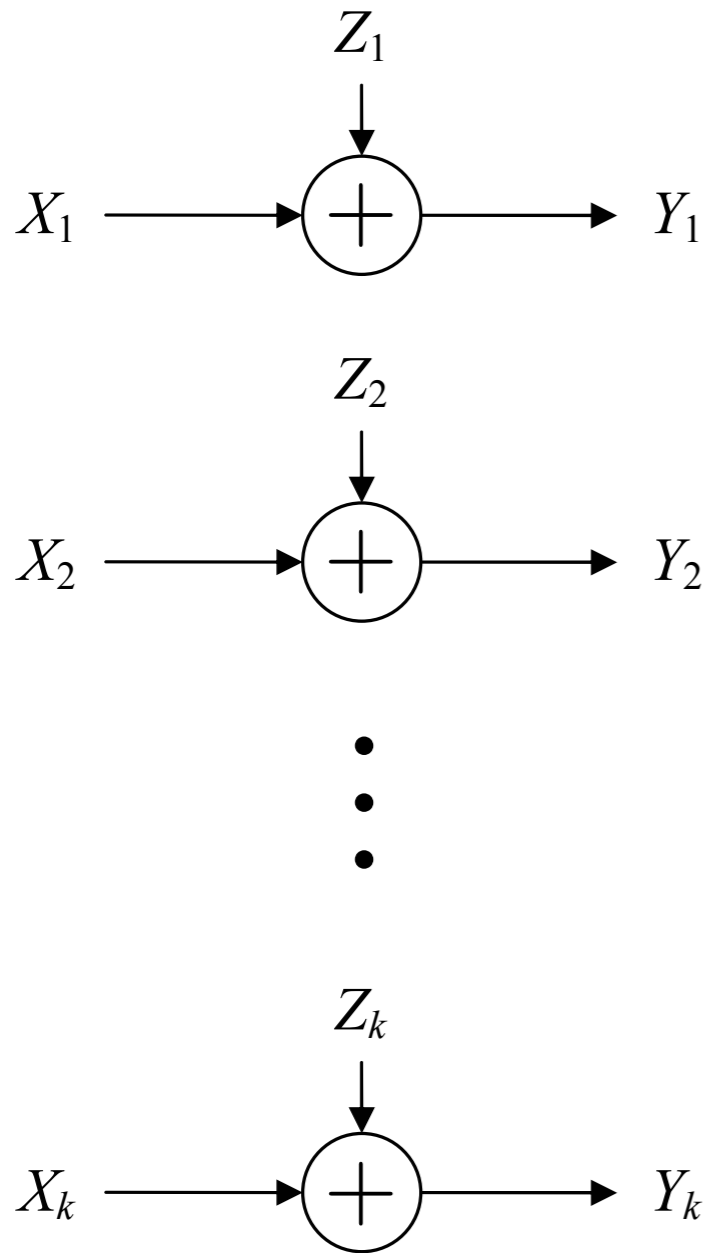
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$$= \frac{1}{2} \sum_{i=1}^k \log(EY_i^2) - \frac{1}{2} \sum_{i=1}^k \log N_i$$

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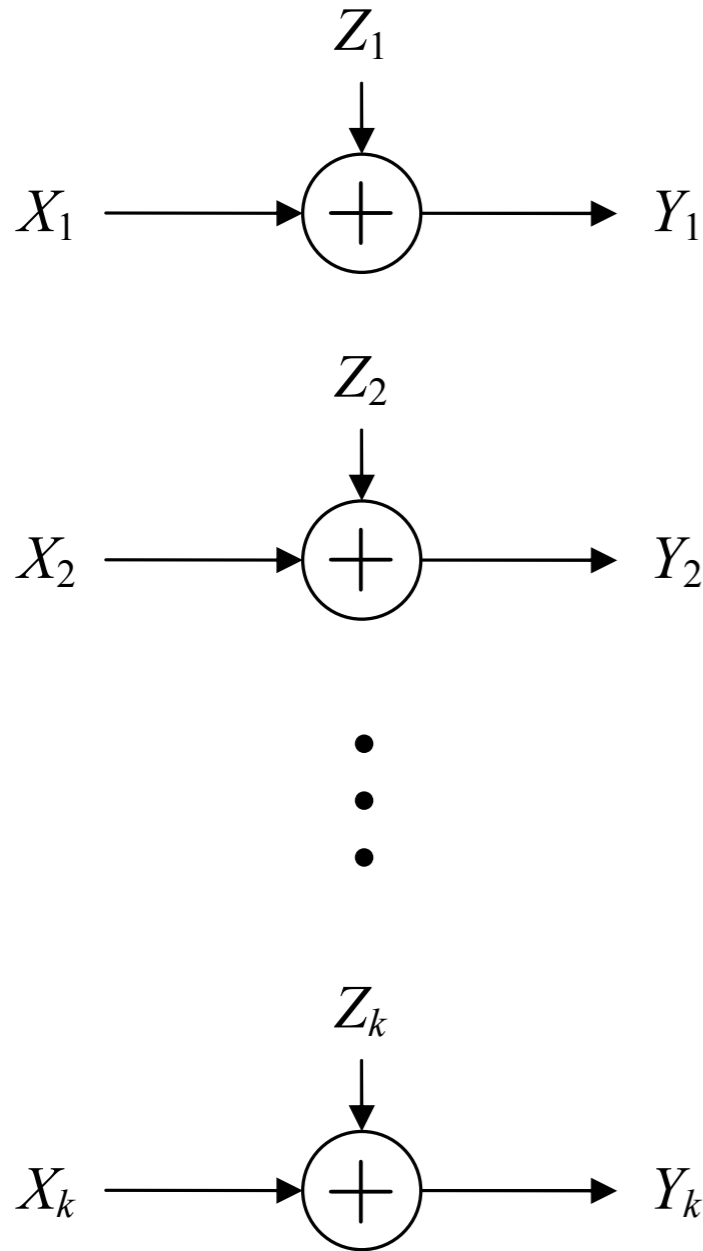
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Formal Justification:

1. Let $P_i = EX_i^2$ be the input power of the i th channel. Consider



$$I(\mathbf{X}; \mathbf{Y}) = h(\mathbf{Y}) - h(\mathbf{Z}) \leq \sum_{i=1}^k h(Y_i) - \sum_{i=1}^k h(Z_i) \quad (1)$$

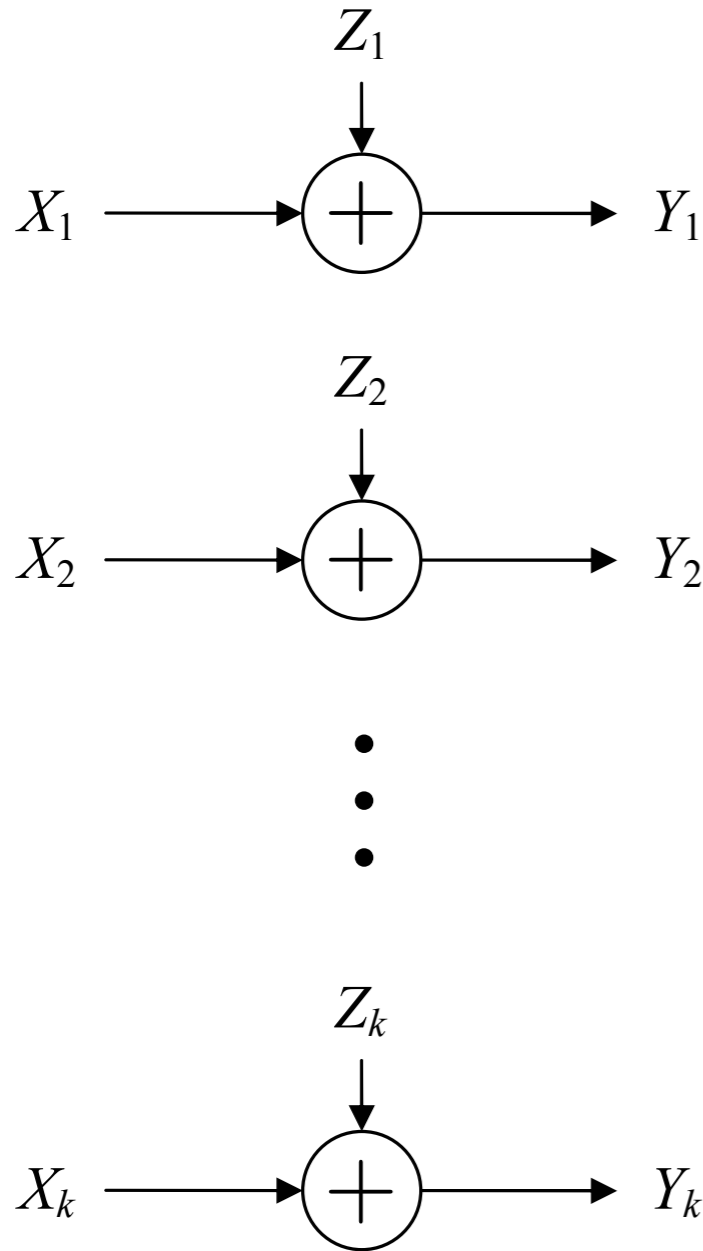
$$\leq \frac{1}{2} \sum_{i=1}^k \log[2\pi e(EY_i^2)] - \frac{1}{2} \sum_{i=1}^k \log(2\pi e N_i) \quad (2)$$

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$$= \frac{1}{2} \sum_{i=1}^k \log(\underline{EX_i^2 + EZ_i^2}) - \frac{1}{2} \sum_{i=1}^k \log N_i$$

Formal Justification:

1. Let $P_i = EX_i^2$ be the input power of the i th channel. Consider



$$I(\mathbf{X}; \mathbf{Y}) = h(\mathbf{Y}) - h(\mathbf{Z}) \leq \sum_{i=1}^k h(Y_i) - \sum_{i=1}^k h(Z_i) \quad (1)$$

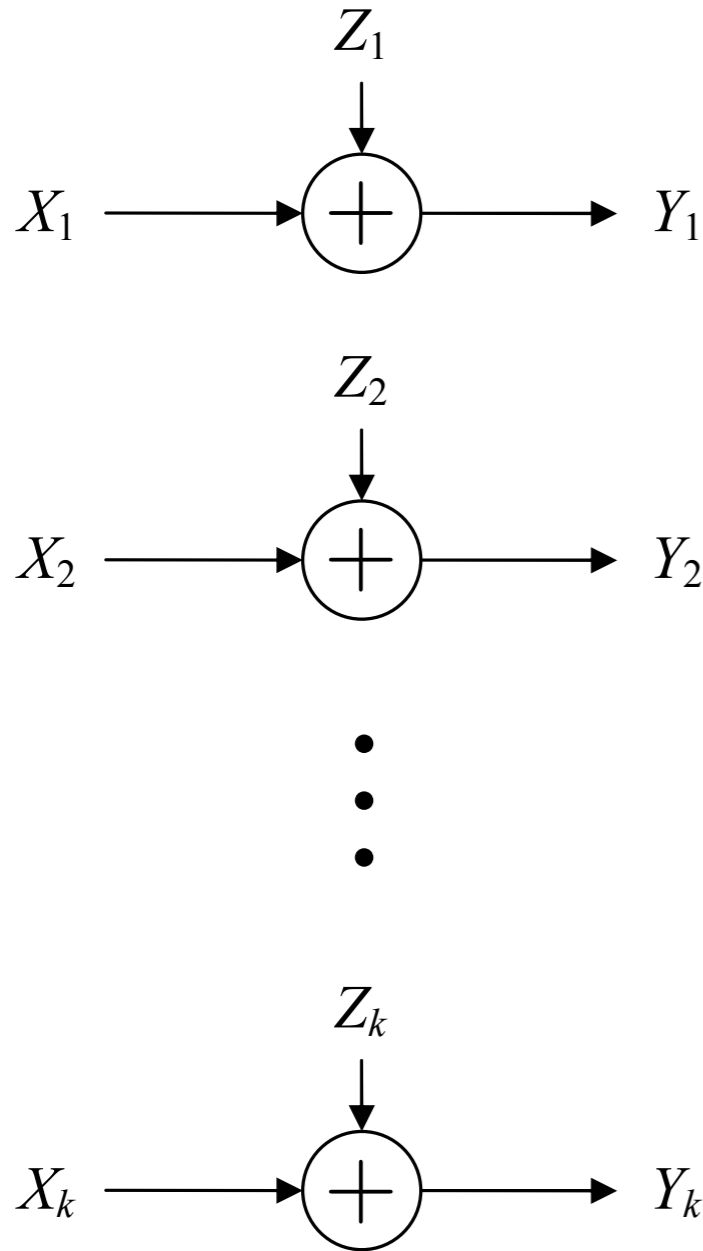
$$\leq \frac{1}{2} \sum_{i=1}^k \log[2\pi e(EY_i^2)] - \frac{1}{2} \sum_{i=1}^k \log(2\pi eN_i) \quad (2)$$

$$= \frac{1}{2} \sum_{i=1}^k \log(EY_i^2) - \frac{1}{2} \sum_{i=1}^k \log N_i$$

$$= \frac{1}{2} \sum_{i=1}^k \log(\underline{EX_i^2} + EZ_i^2) - \frac{1}{2} \sum_{i=1}^k \log N_i$$

Formal Justification:

1. Let $P_i = EX_i^2$ be the input power of the i th channel. Consider



$$I(\mathbf{X}; \mathbf{Y}) = h(\mathbf{Y}) - h(\mathbf{Z}) \leq \sum_{i=1}^k h(Y_i) - \sum_{i=1}^k h(Z_i) \quad (1)$$

$$\leq \frac{1}{2} \sum_{i=1}^k \log[2\pi e(EY_i^2)] - \frac{1}{2} \sum_{i=1}^k \log(2\pi eN_i) \quad (2)$$

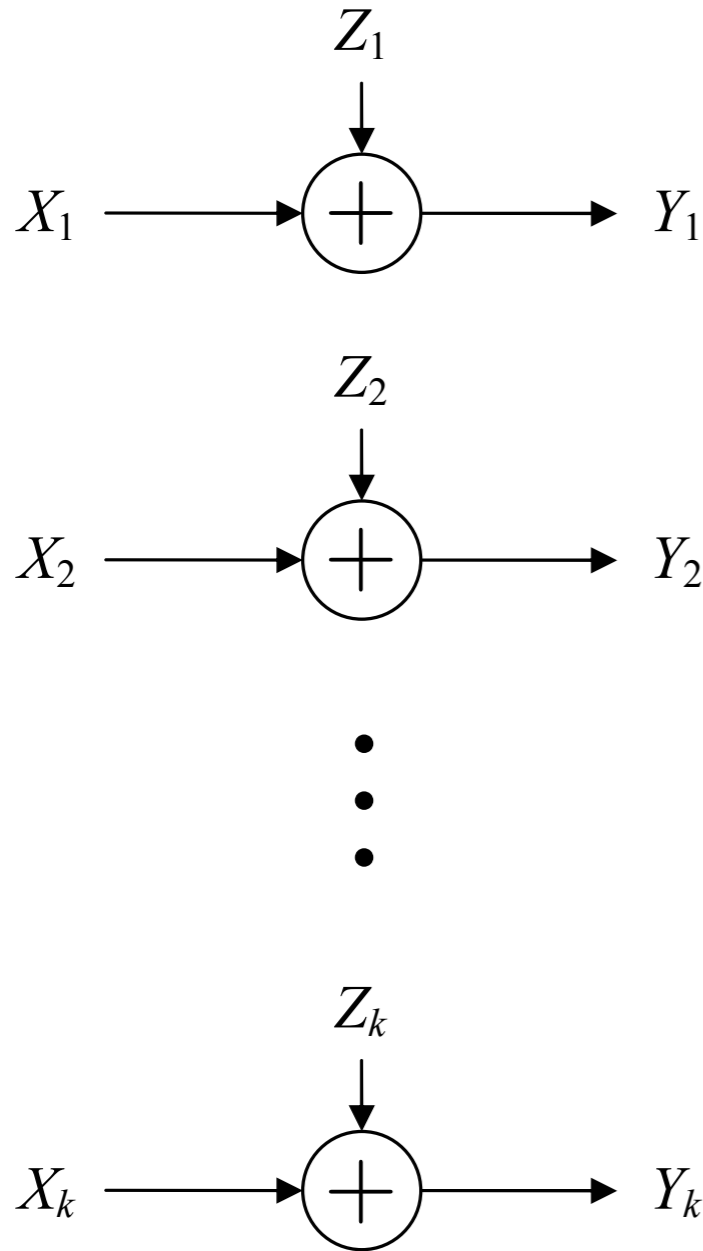
$$= \frac{1}{2} \sum_{i=1}^k \log(EY_i^2) - \frac{1}{2} \sum_{i=1}^k \log N_i$$

$$= \frac{1}{2} \sum_{i=1}^k \log(\underline{EX_i^2} + EZ_i^2) - \frac{1}{2} \sum_{i=1}^k \log N_i$$

$$= \frac{1}{2} \sum_{i=1}^k \log(\underline{P_i} + N_i) - \frac{1}{2} \sum_{i=1}^k \log N_i \quad (3)$$

Formal Justification:

1. Let $P_i = EX_i^2$ be the input power of the i th channel. Consider



$$I(\mathbf{X}; \mathbf{Y}) = h(\mathbf{Y}) - h(\mathbf{Z}) \leq \sum_{i=1}^k h(Y_i) - \sum_{i=1}^k h(Z_i) \quad (1)$$

$$\leq \frac{1}{2} \sum_{i=1}^k \log[2\pi e(EY_i^2)] - \frac{1}{2} \sum_{i=1}^k \log(2\pi eN_i) \quad (2)$$

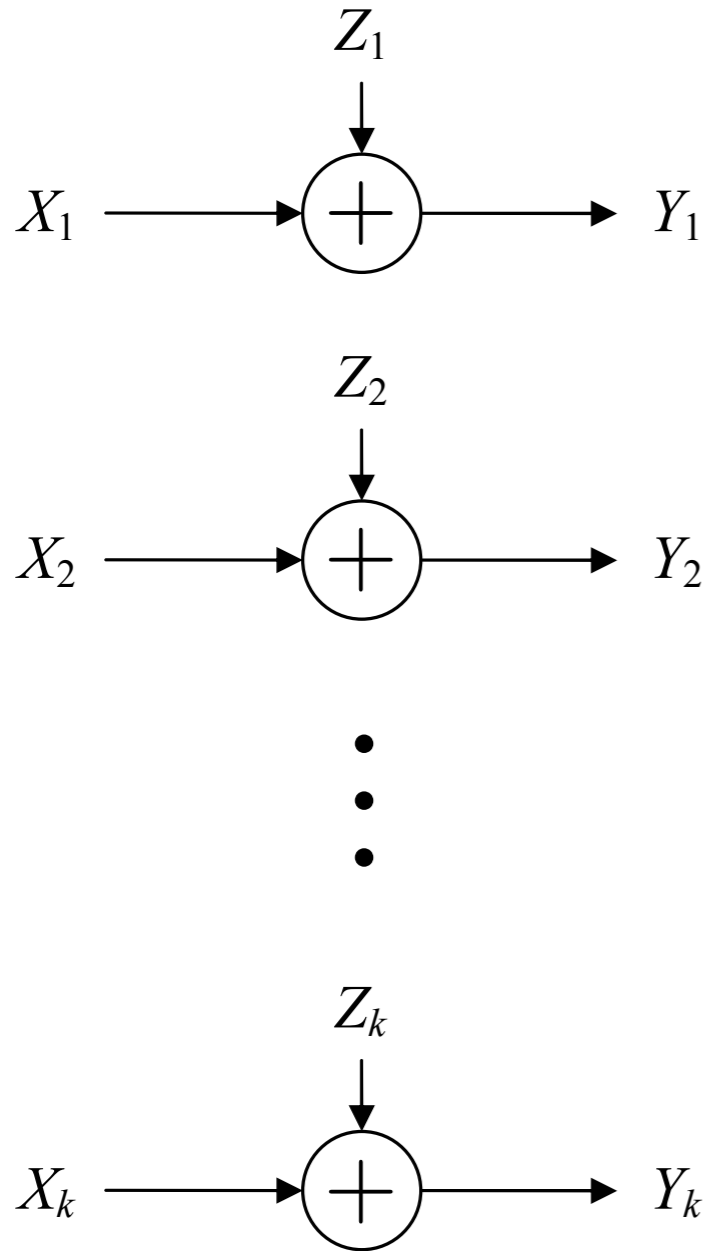
$$= \frac{1}{2} \sum_{i=1}^k \log(EY_i^2) - \frac{1}{2} \sum_{i=1}^k \log N_i$$

$$= \frac{1}{2} \sum_{i=1}^k \log(EX_i^2 + \underline{EZ_i^2}) - \frac{1}{2} \sum_{i=1}^k \log N_i$$

$$= \frac{1}{2} \sum_{i=1}^k \log(P_i + N_i) - \frac{1}{2} \sum_{i=1}^k \log N_i \quad (3)$$

Formal Justification:

1. Let $P_i = EX_i^2$ be the input power of the i th channel. Consider



$$I(\mathbf{X}; \mathbf{Y}) = h(\mathbf{Y}) - h(\mathbf{Z}) \leq \sum_{i=1}^k h(Y_i) - \sum_{i=1}^k h(Z_i) \quad (1)$$

$$\leq \frac{1}{2} \sum_{i=1}^k \log[2\pi e(EX_i^2)] - \frac{1}{2} \sum_{i=1}^k \log(2\pi eN_i) \quad (2)$$

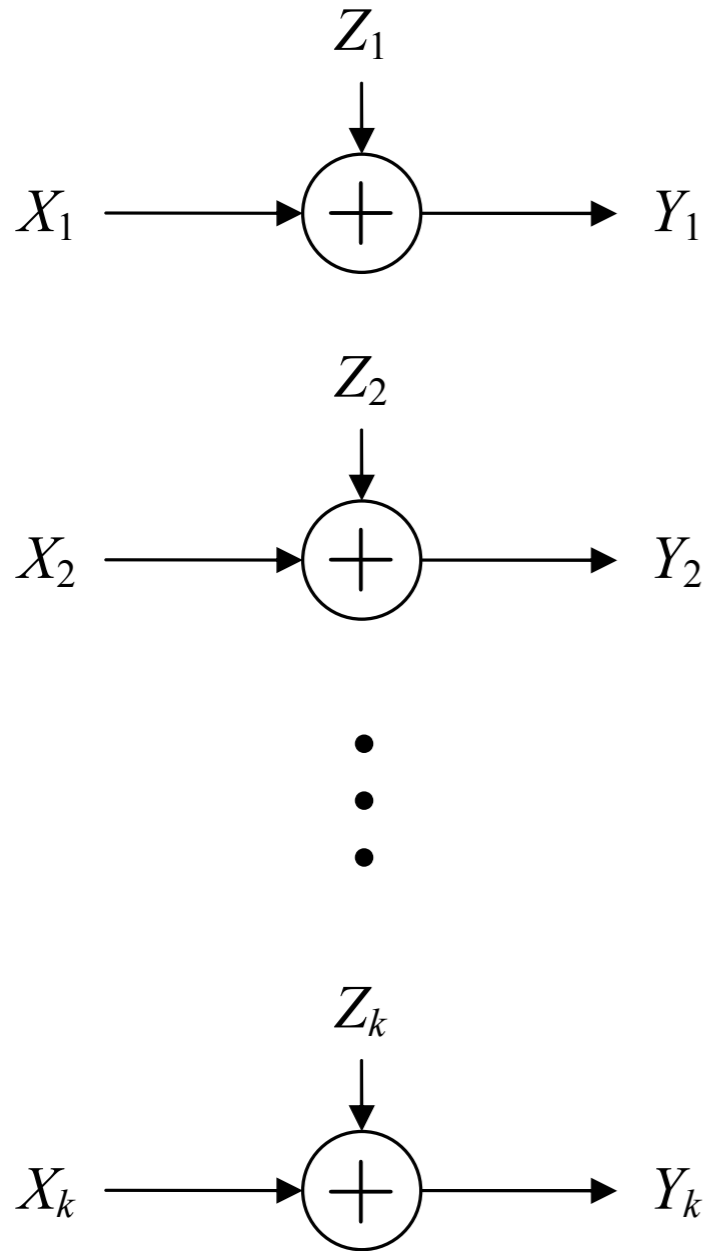
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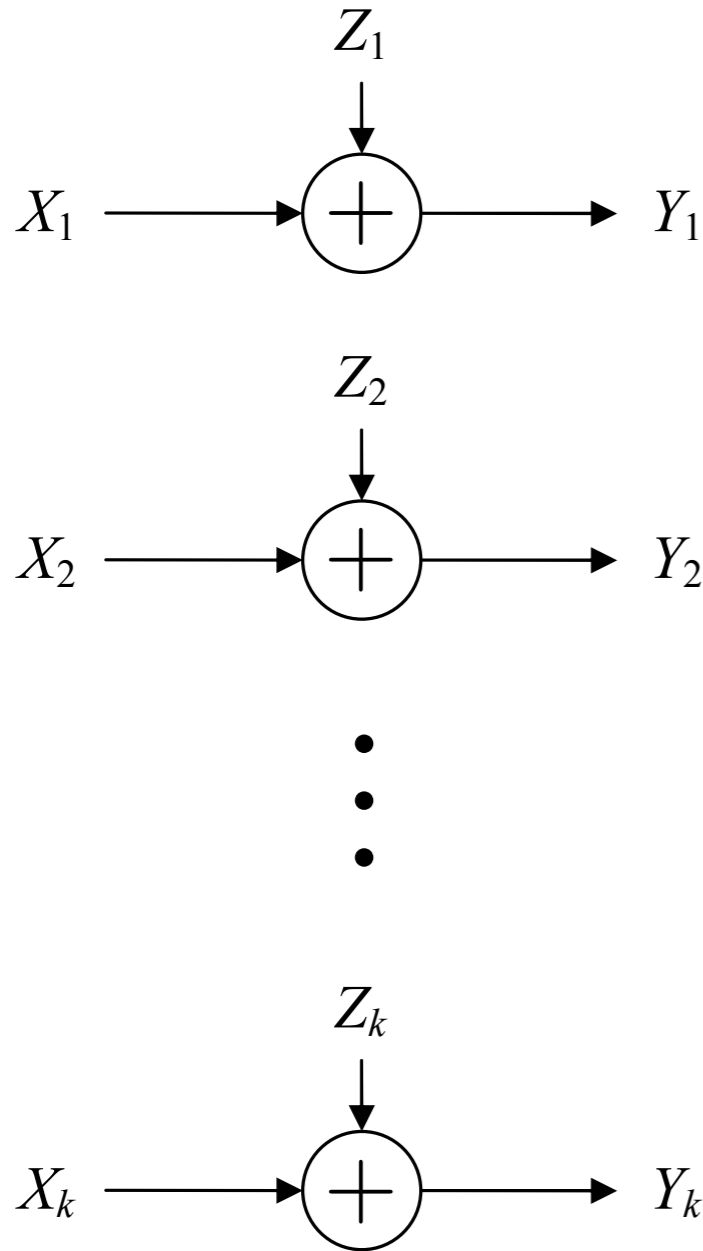
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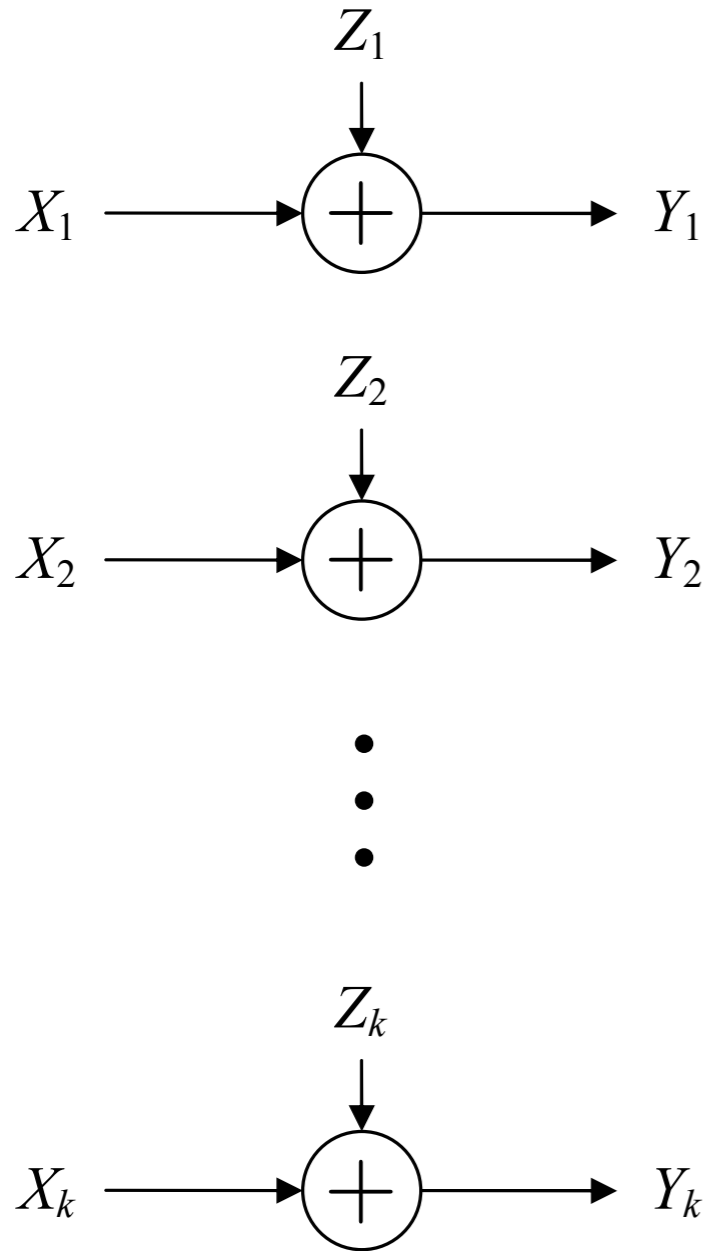
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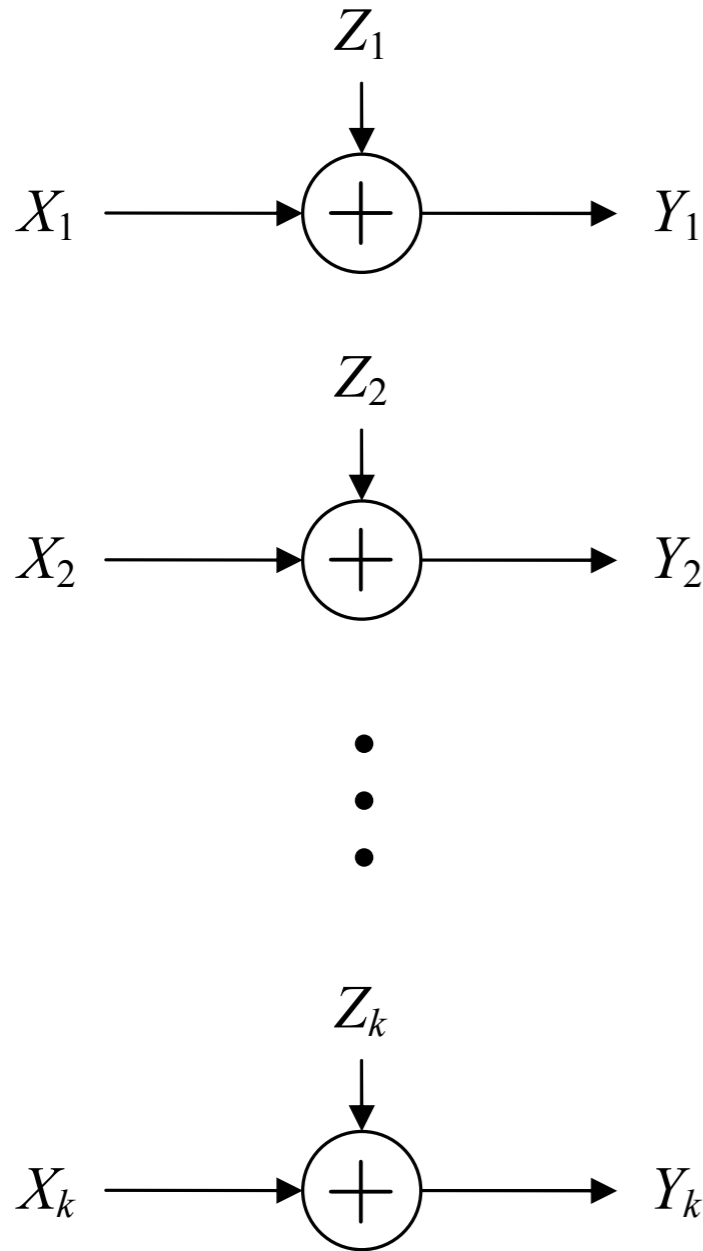
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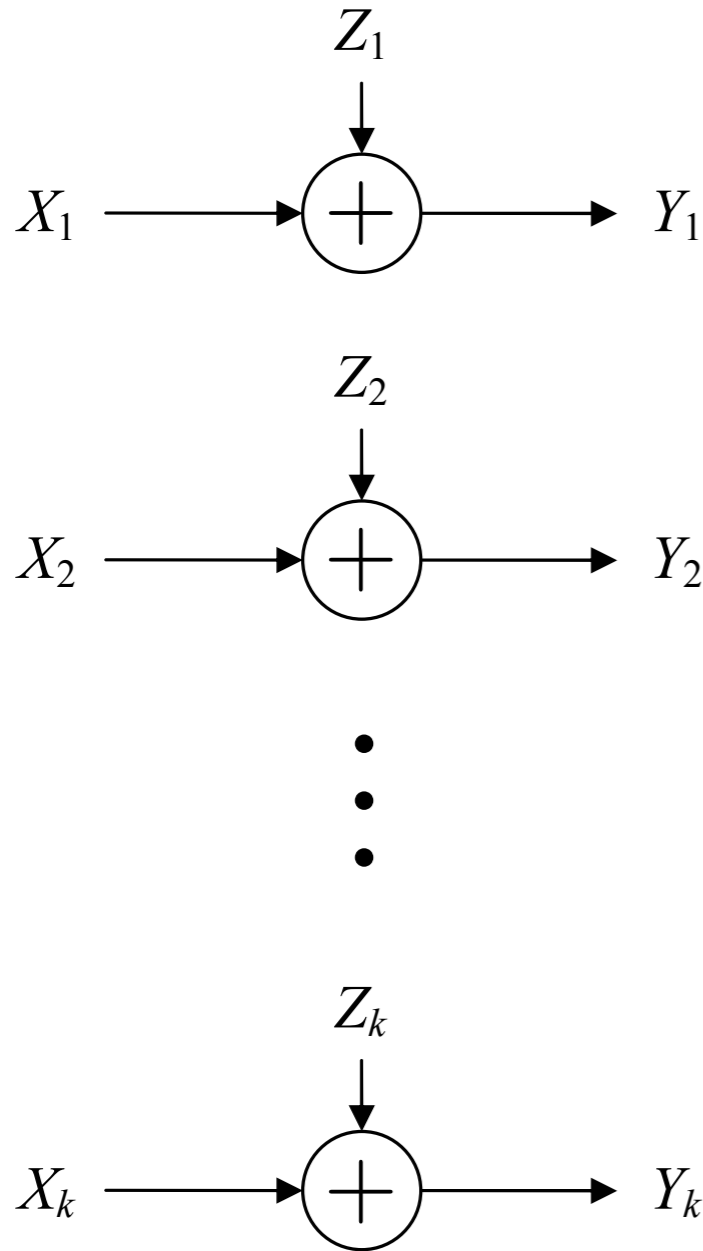
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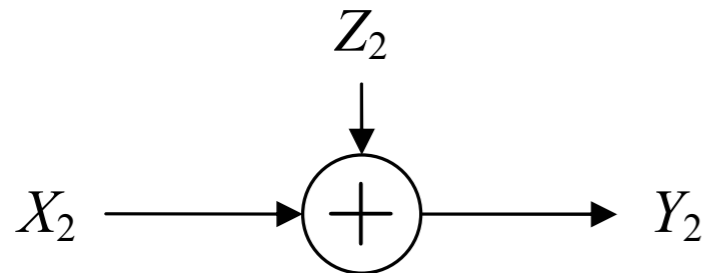
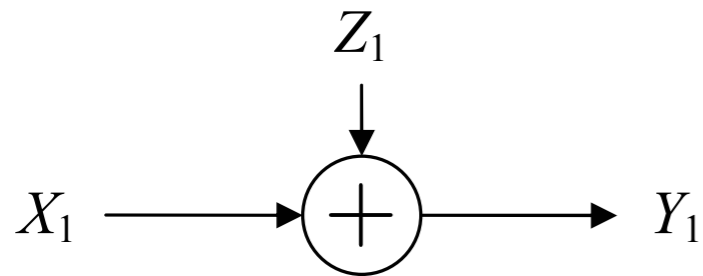
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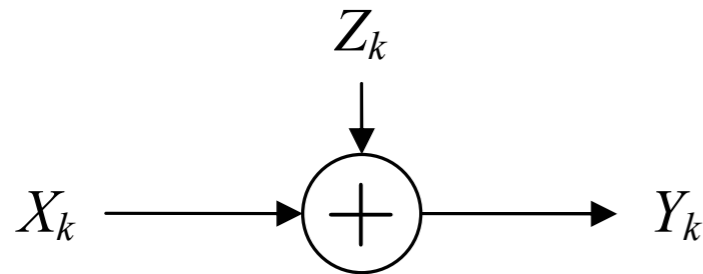
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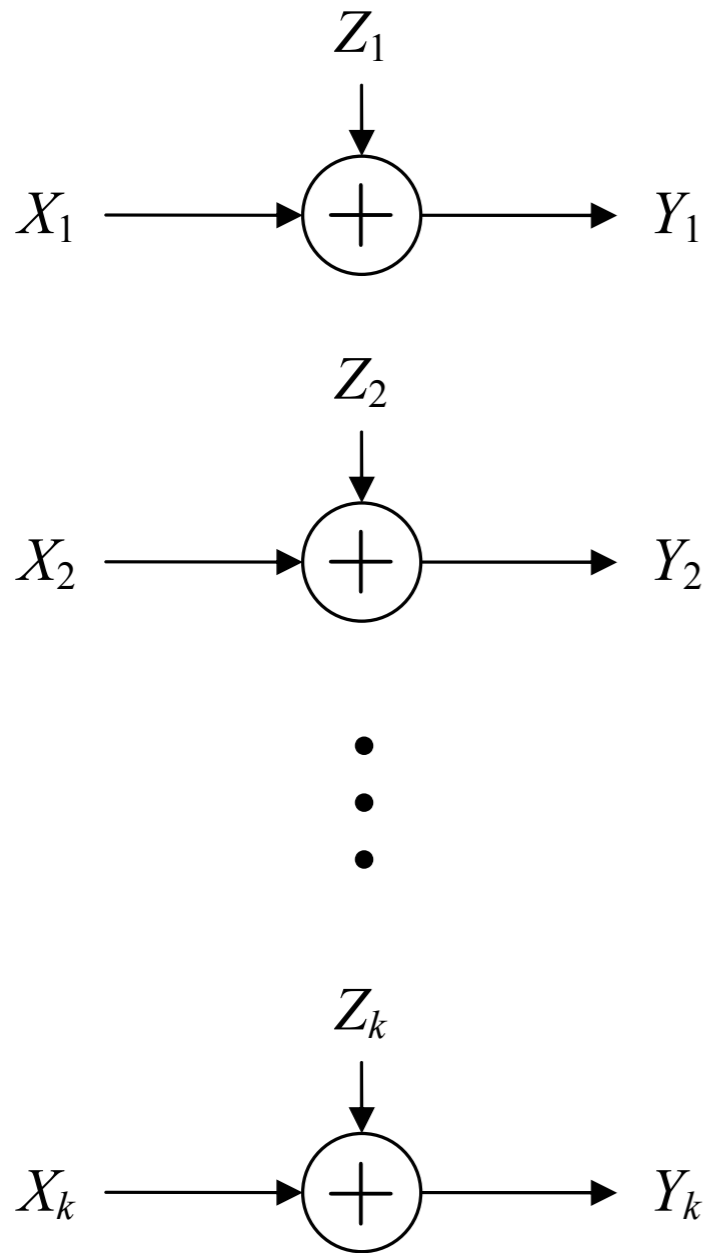
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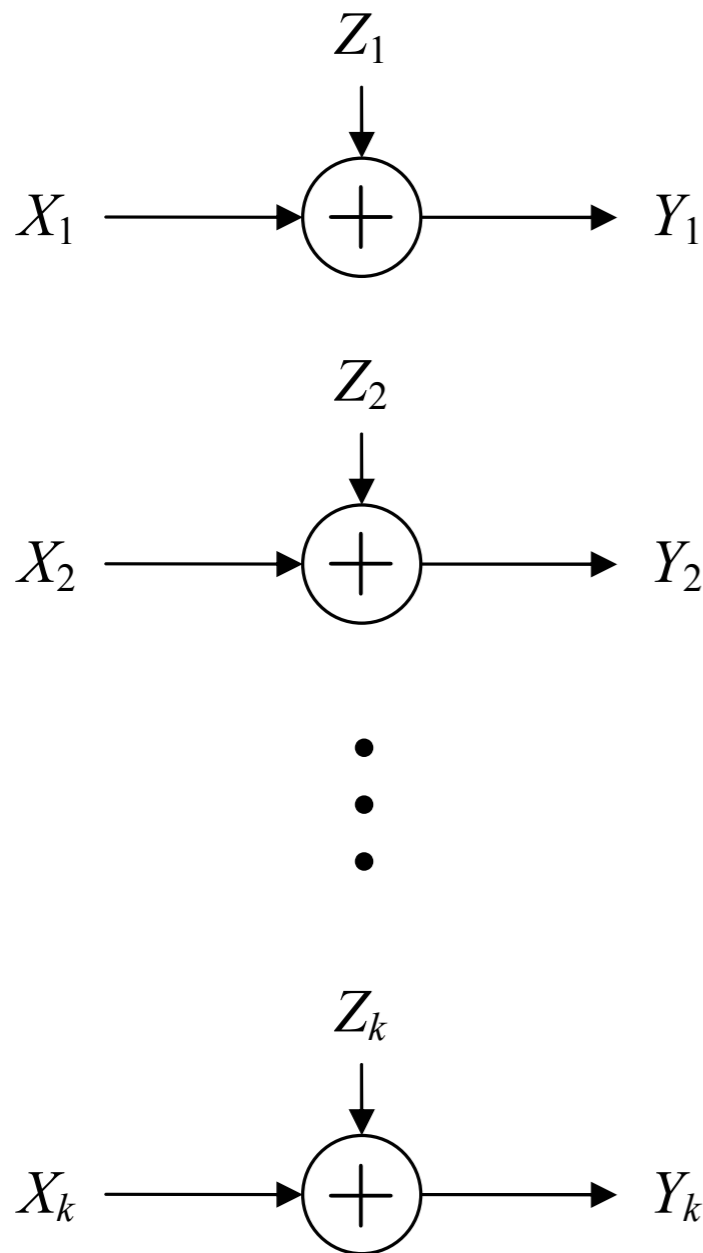
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4. The capacity of the system of parallel Gaussian channels is equal to the sum of the capacities of the individual Gaussian channels with the input power optimally allocated.

Maximize $\sum_i \log(P_i + N_i)$ subject to $\sum_i P_i \leq P$ and $P_i \geq 0$.

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This solution has a [water-filling](#) interpretation.

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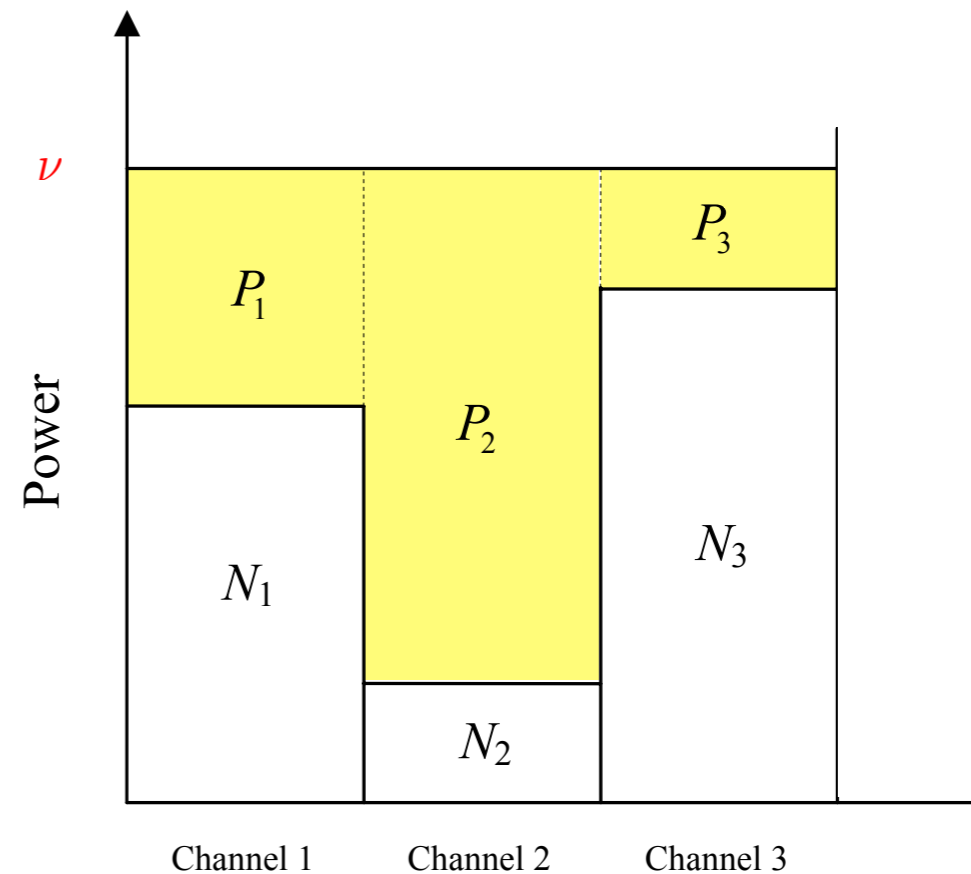
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3. Therefore, set $\sum_i P_i = P$.
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$$J = \sum_{i=1}^k \log(P_i + N_i) - \mu \sum_{i=1}^k P_i.$$

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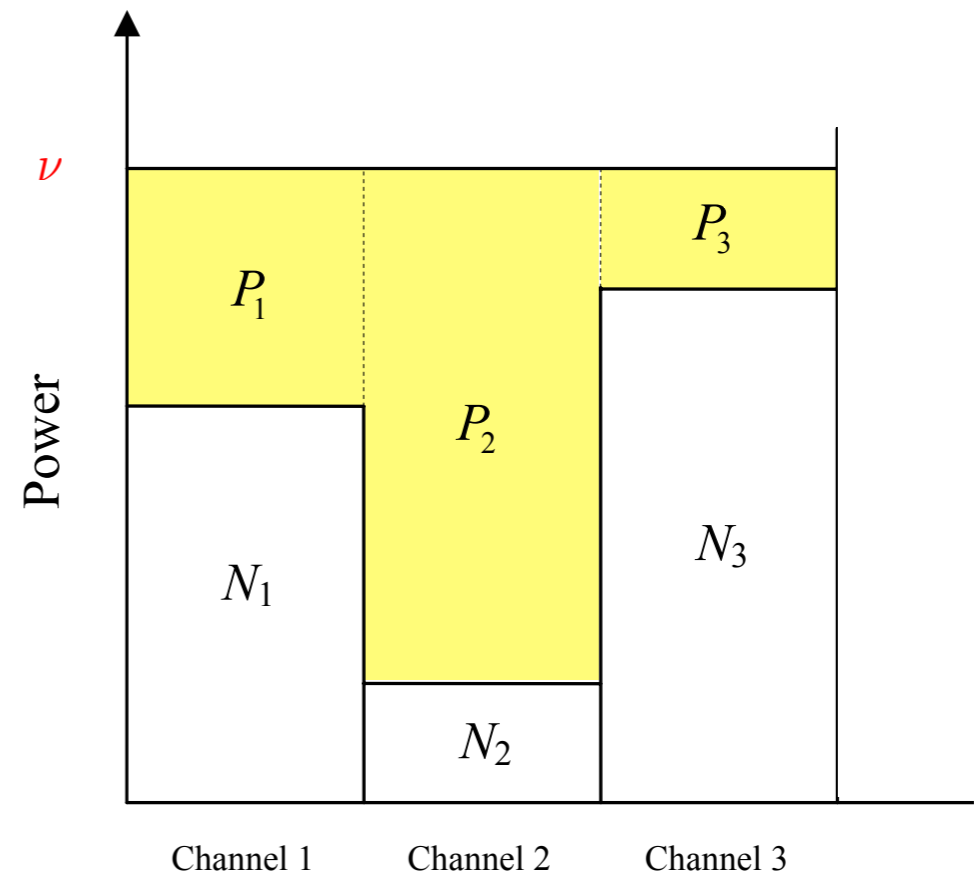
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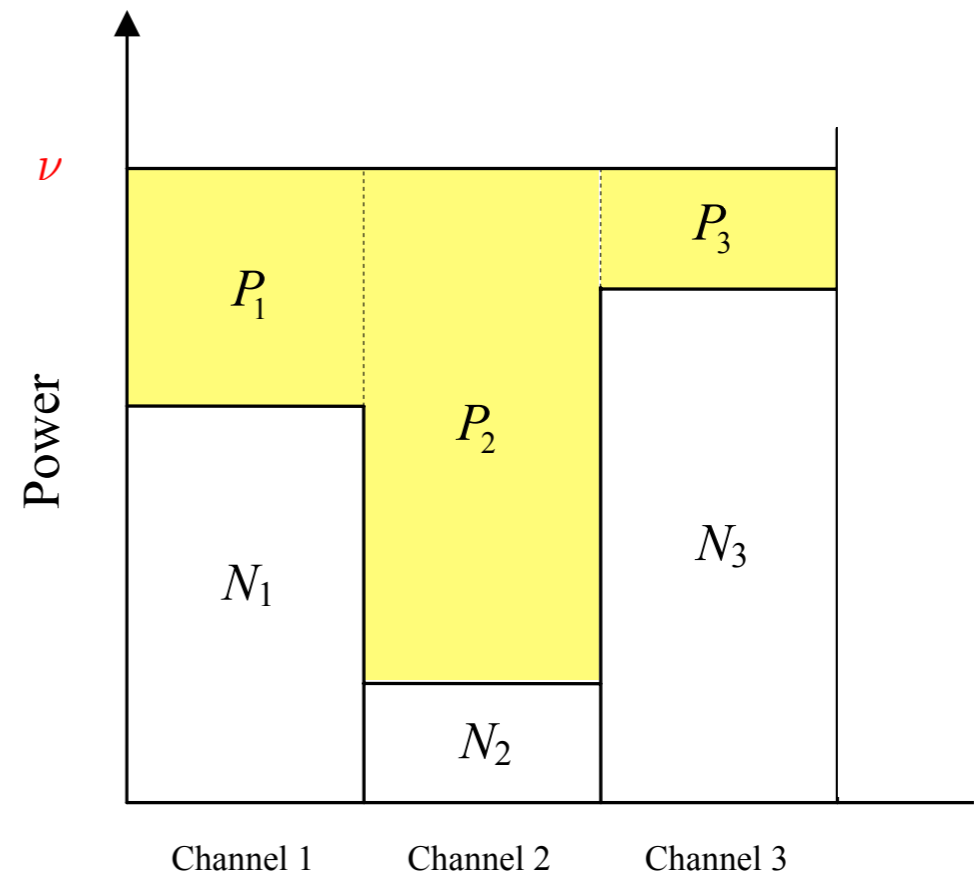
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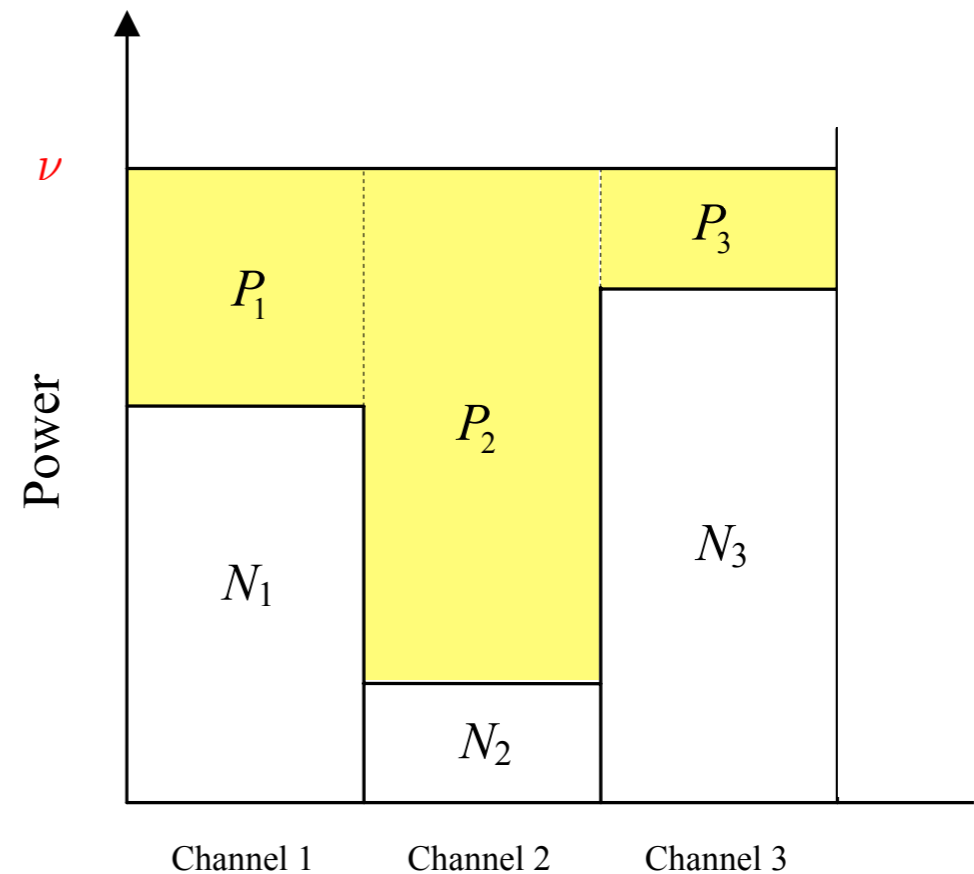
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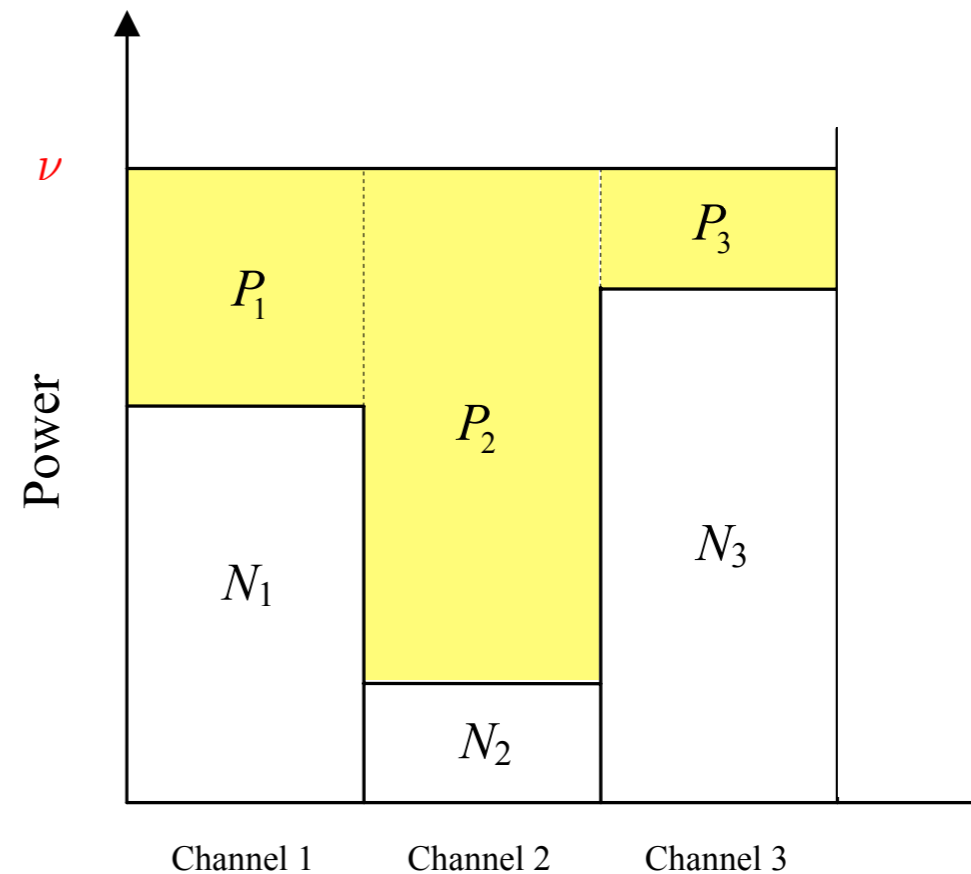
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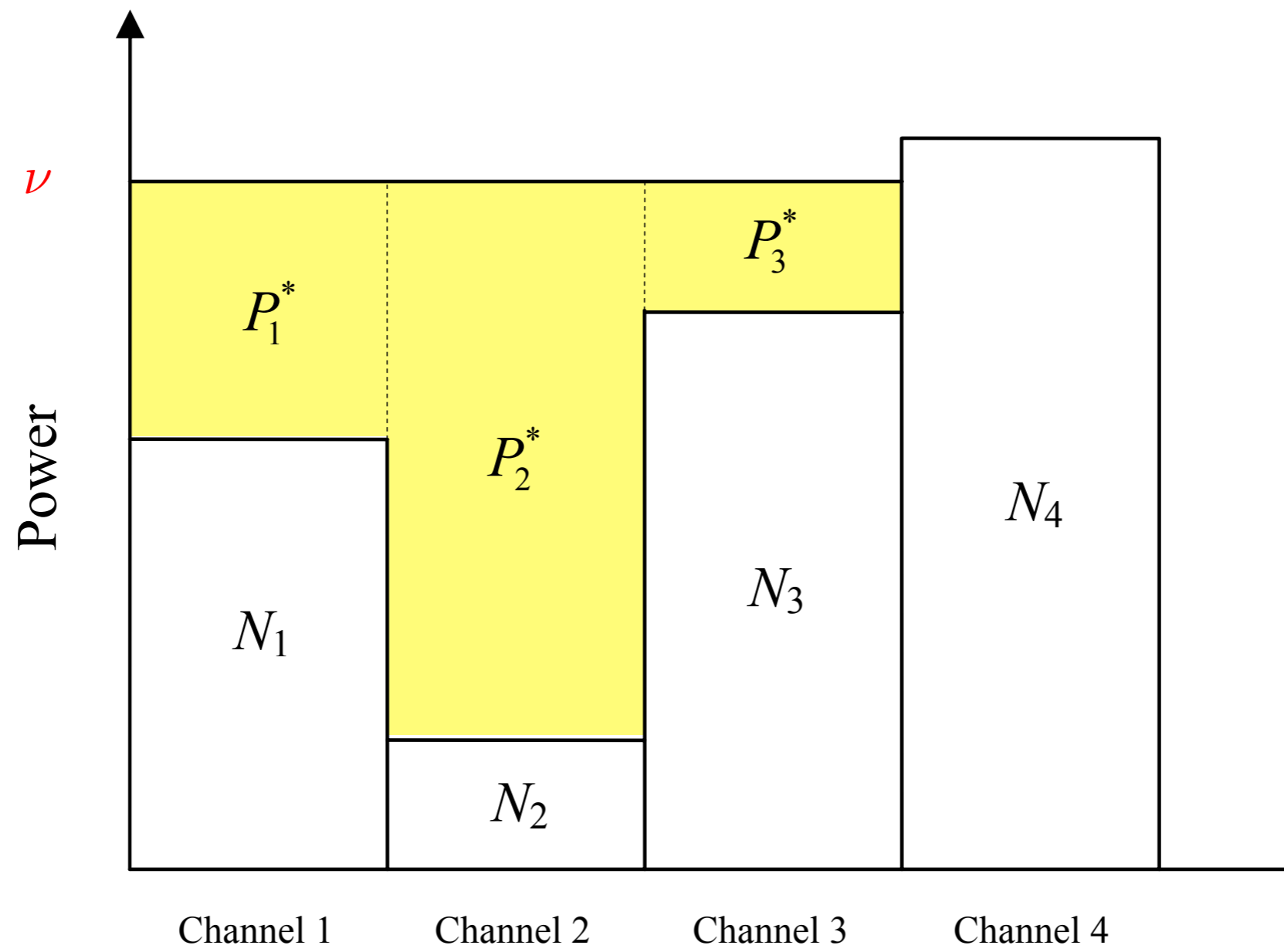
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$$\frac{\log e}{(a_i^* + \lambda_i)} - \mu + \mu_i = 0 \quad (3)$$

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$$\mu_i a_i^* = 0, \quad 1 \leq i \leq k, \quad (5)$$

where μ and μ_i are the multipliers associated with the constraints in (1) and (2), respectively.

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For given $\lambda_i \geq 0$, maximize $\sum_{i=1}^k \log(a_i + \lambda_i)$ subject to

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$$\begin{aligned} \frac{\log e}{(a_i^* + \lambda_i)} - \mu + \mu_i &= 0 \\ \frac{\log e}{\lambda_i} - \frac{\log e}{\nu} + \underline{\mu_i} &= 0 \end{aligned}$$

which implies

$$\mu_i = (\log e) \left(\frac{1}{\nu} - \frac{1}{\lambda_i} \right) \geq 0.$$

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For given $\lambda_i \geq 0$, maximize $\sum_{i=1}^k \log(a_i + \lambda_i)$ subject to

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4. For i such that $a_i^* = 0$,

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which implies

$$\underline{\mu_i = (\log e) \left(\frac{1}{\nu} - \frac{1}{\lambda_i} \right) \geq 0.}$$

Proposition 11.23 The problem

For given $\lambda_i \geq 0$, maximize $\sum_{i=1}^k \log(a_i + \lambda_i)$ subject to

$$\sum_i a_i \leq P \quad (1)$$

$$-a_i \leq 0. \quad (2)$$

has the solution

$$a_i^* = (\nu - \lambda_i)^+, \quad 1 \leq i \leq k,$$

where ν satisfies

$$\sum_{i=1}^k (\nu - \lambda_i)^+ = P.$$

Proof

1. We will prove the proposition by verifying that the proposed solution satisfies the KKT condition. This is done by finding nonnegative μ and μ_i satisfying the equations

$$\frac{\log e}{(a_i^* + \lambda_i)} - \mu + \mu_i = 0 \quad (3)$$

$$\mu \left(P - \sum_{i=1}^k a_i^* \right) = 0 \quad (4)$$

$$\mu_i a_i^* = 0, \quad 1 \leq i \leq k, \quad (5)$$

where μ and μ_i are the multipliers associated with the constraints in (1) and (2), respectively.

2. To avoid triviality, assume $P > 0$ so that $\nu > 0$, and observe that there exists at least one i such that $a_i^* > 0$.

3. For i such that $a_i^* > 0$:

- (5) implies $\mu_i = 0$
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5. Thus we have obtained nonnegative μ and μ_i satisfying the KKT condition.