

Structure of Sequential State Discrimination

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○ Part I : Theory

- Scenario of sequential state discrimination
- Constructing optimization problem
- Comparison with other scenarios

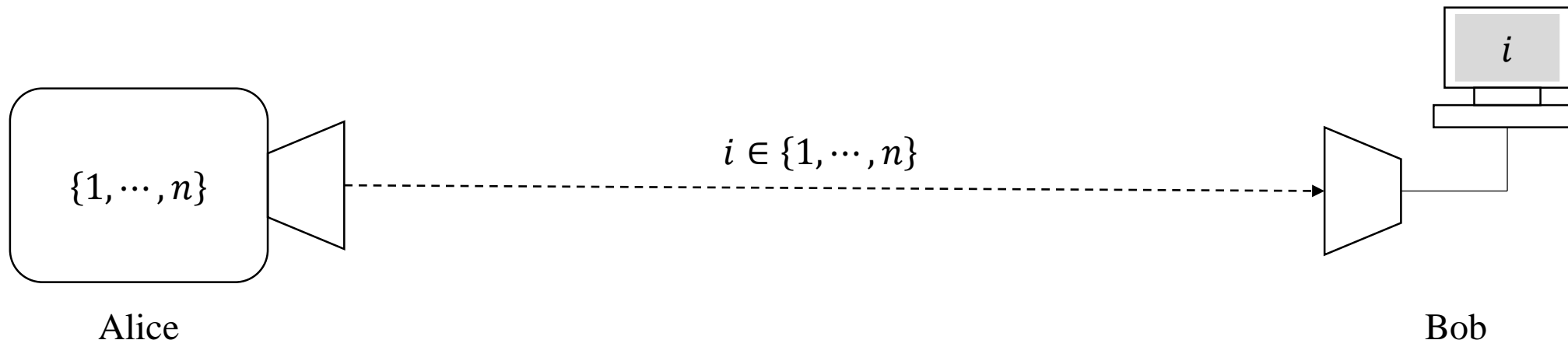
○ Part II : Application

- Realistic QKD
- Implementing sequential state discrimination
- Sequential state discrimination in noisy channel
- Comparison with probabilistic cloning strategy

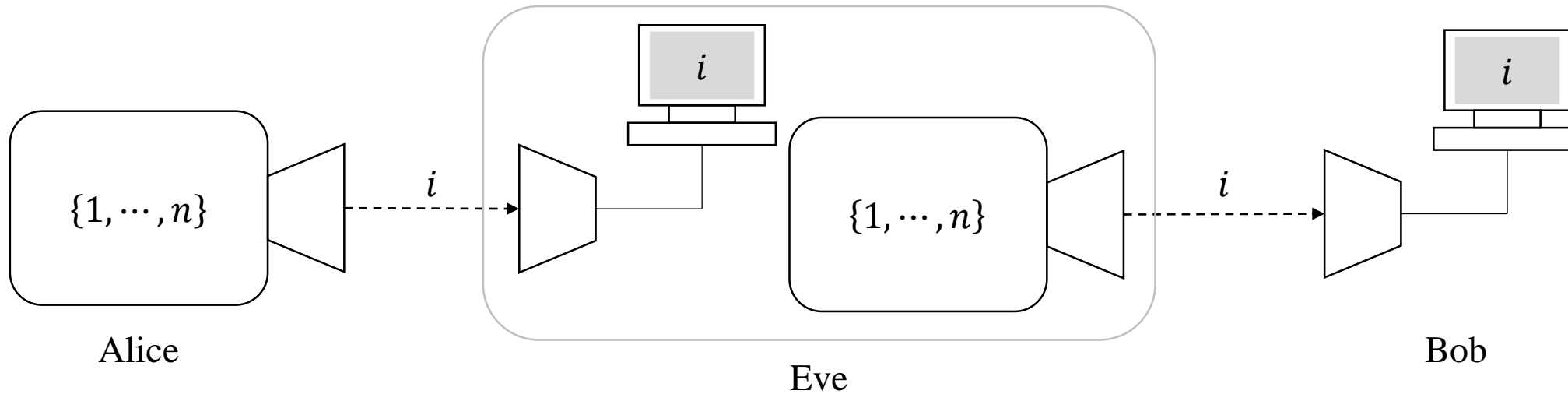
Part I

Theory

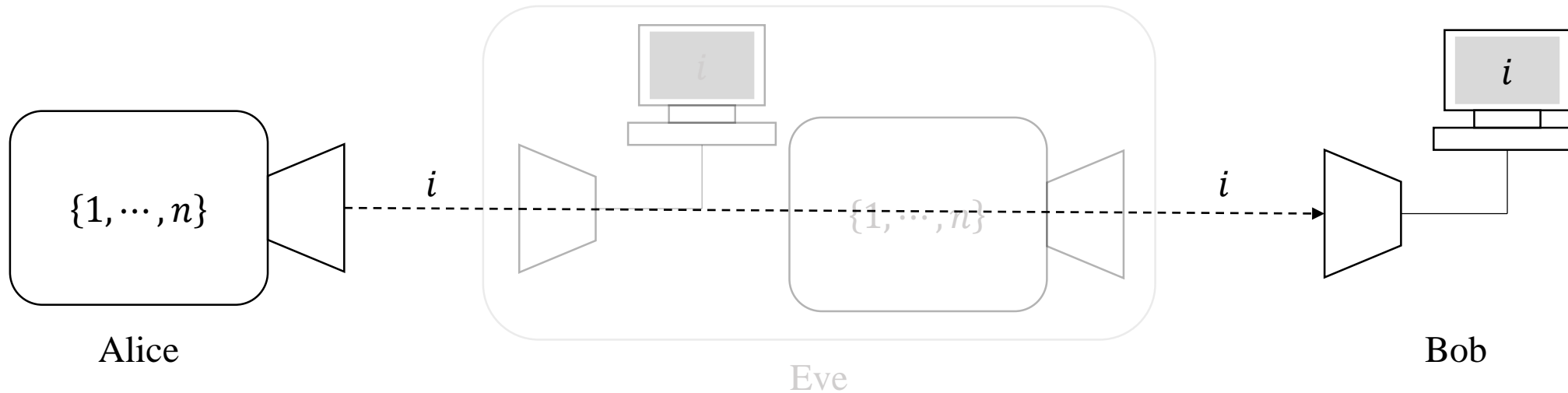
Classical scheme of communication



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Classical scheme of communication

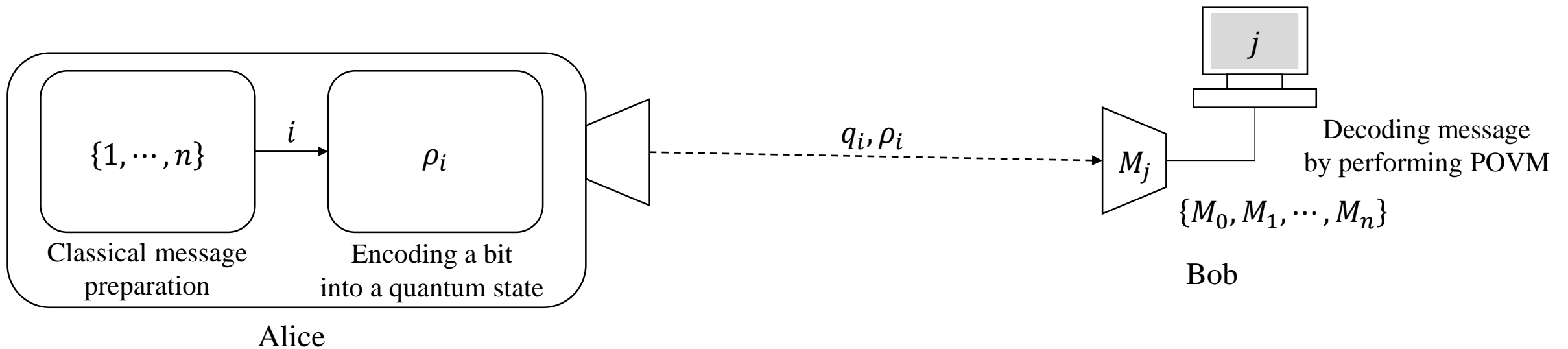


Insecurity of classical scheme

Even Eve eavesdrops secure message during Alice and Bob communicate, Alice and Bob cannot notice Eve.

Scenario of sequential state discrimination

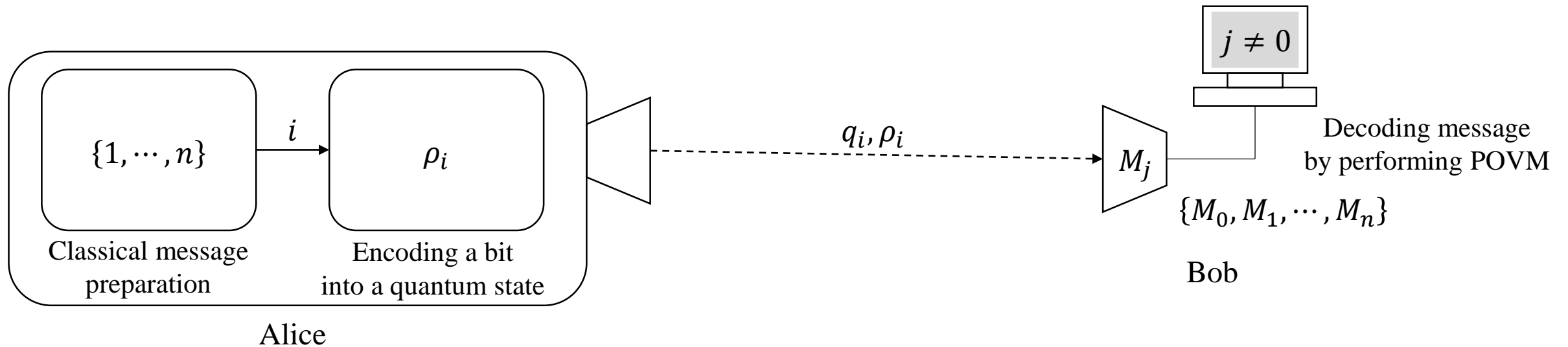
Quantum key distribution



G. Cariolaro, *Quantum Communications* (Springer, 2015).

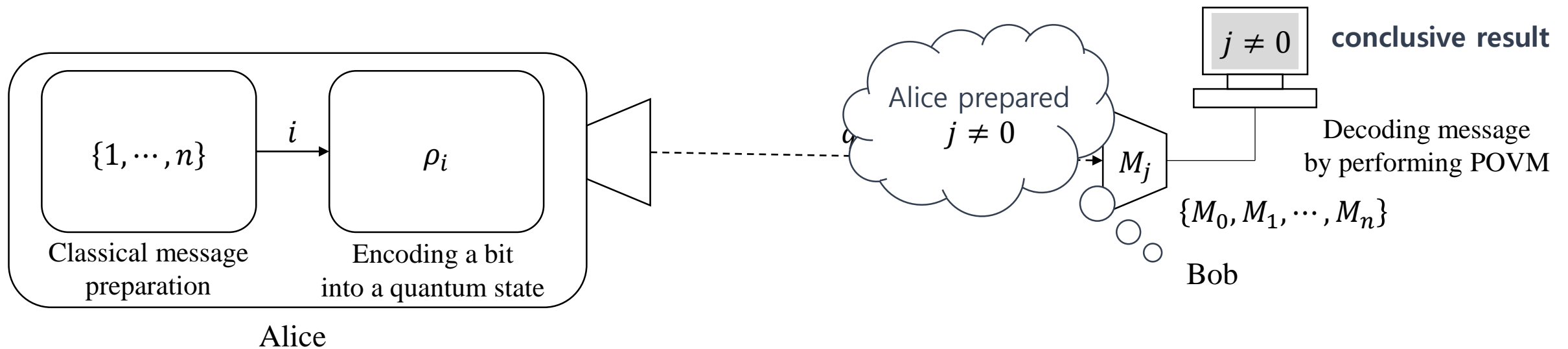
C. H. Bennett, Quantum cryptography using any two nonorthogonal states, *Phys. Rev. Lett.* **68**, 3121 (1992).

Quantum key distribution



Scenario of sequential state discrimination

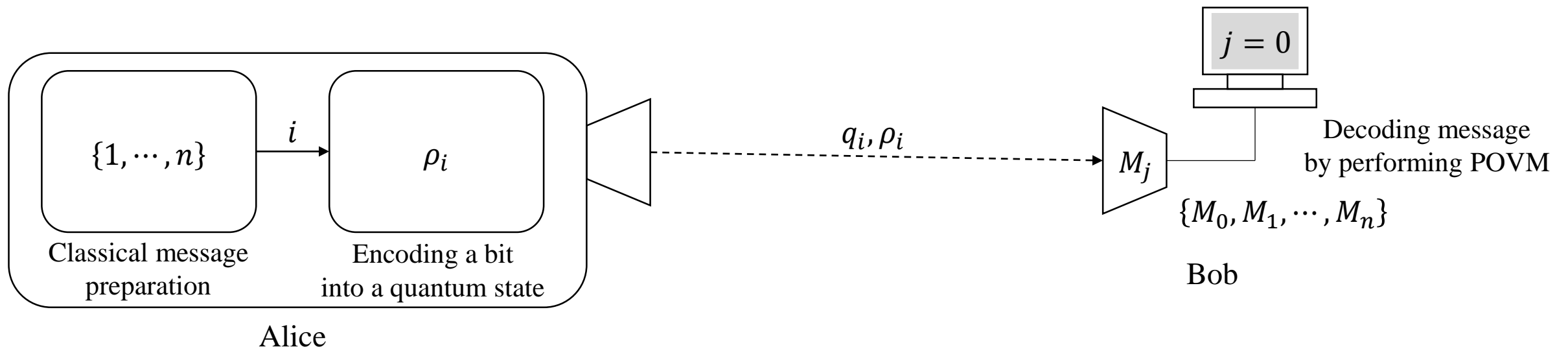
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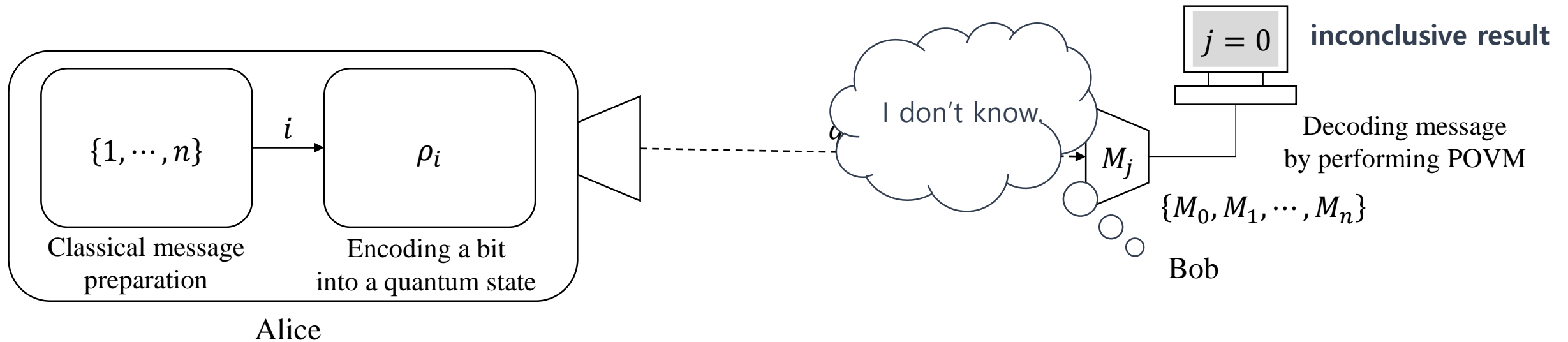
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Quantum key distribution



Scenario of sequential state discrimination

Quantum key distribution

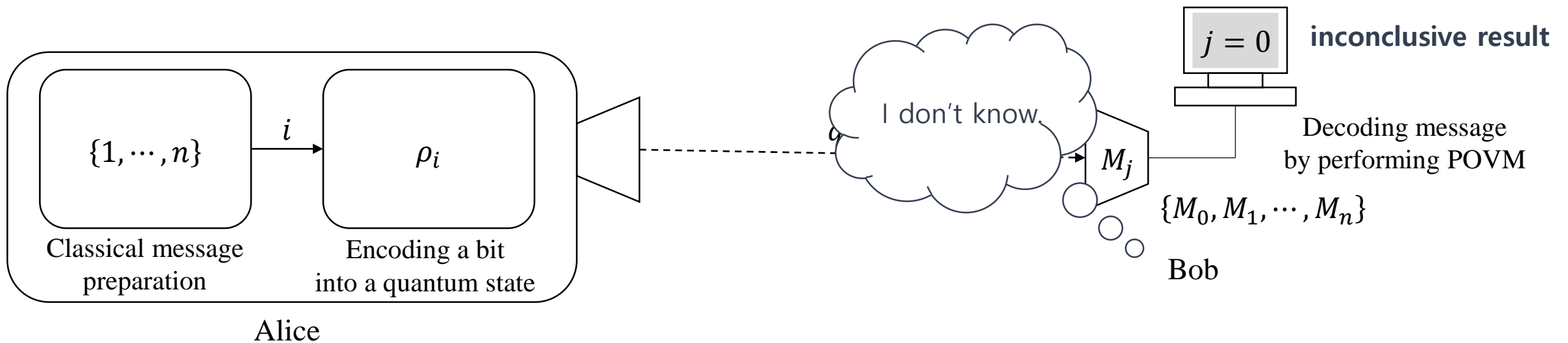


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Scenario of sequential state discrimination

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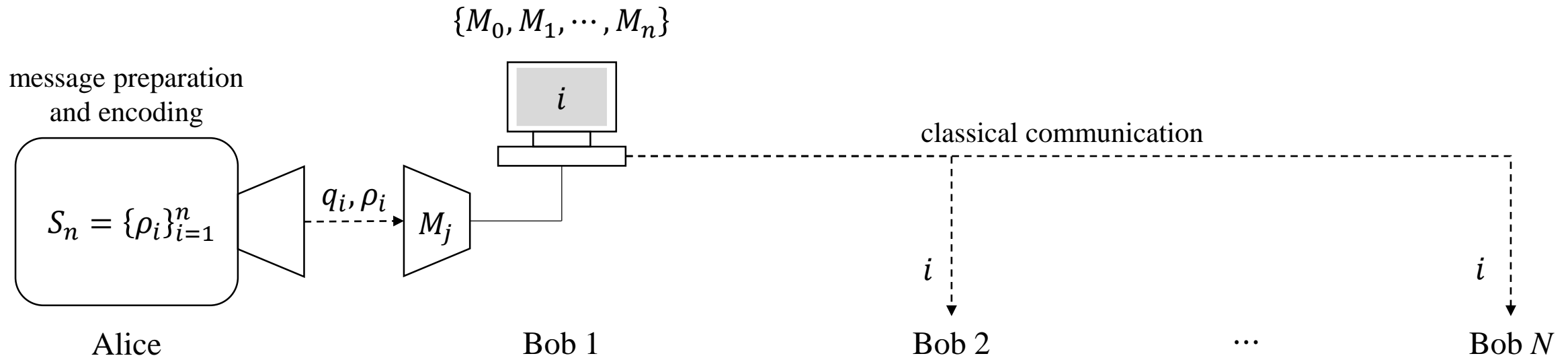
General structure of quantum key distribution

Quantum key distribution can be expressed as quantum state discrimination. For example, if Alice and Bob perform B92 protocol, then this scenario can be expressed as unambiguous discrimination.

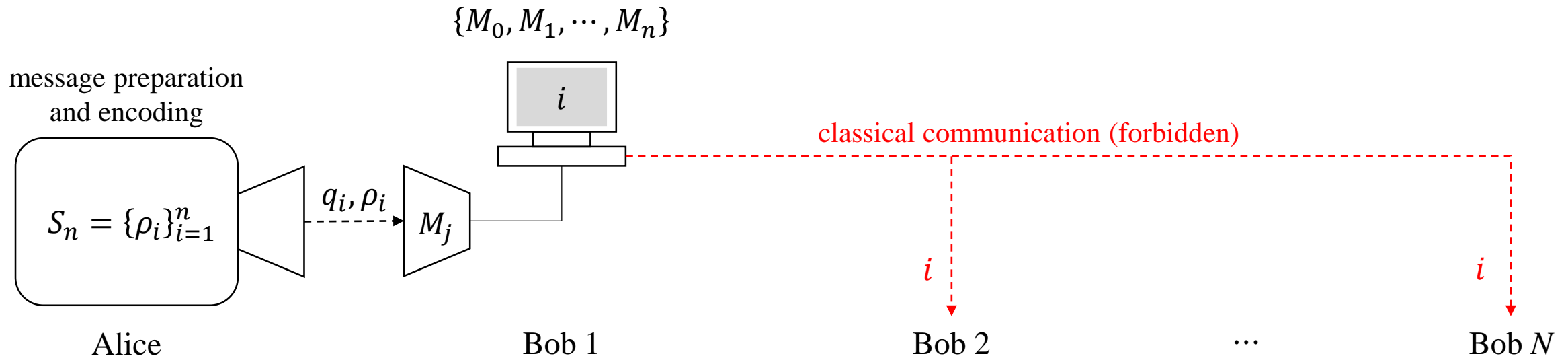
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Multiparty QKD based on sequential state discrimination

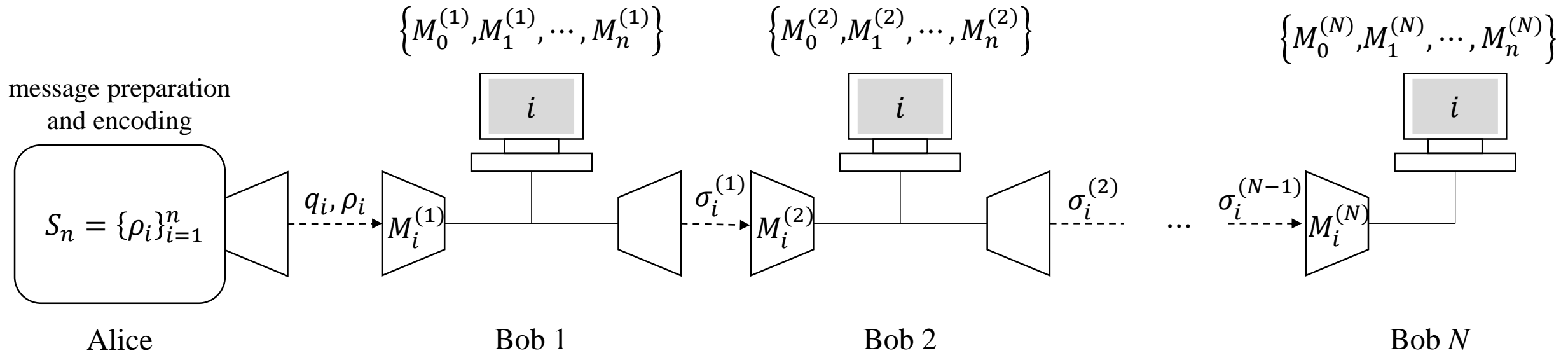


Multiparty QKD based on sequential state discrimination

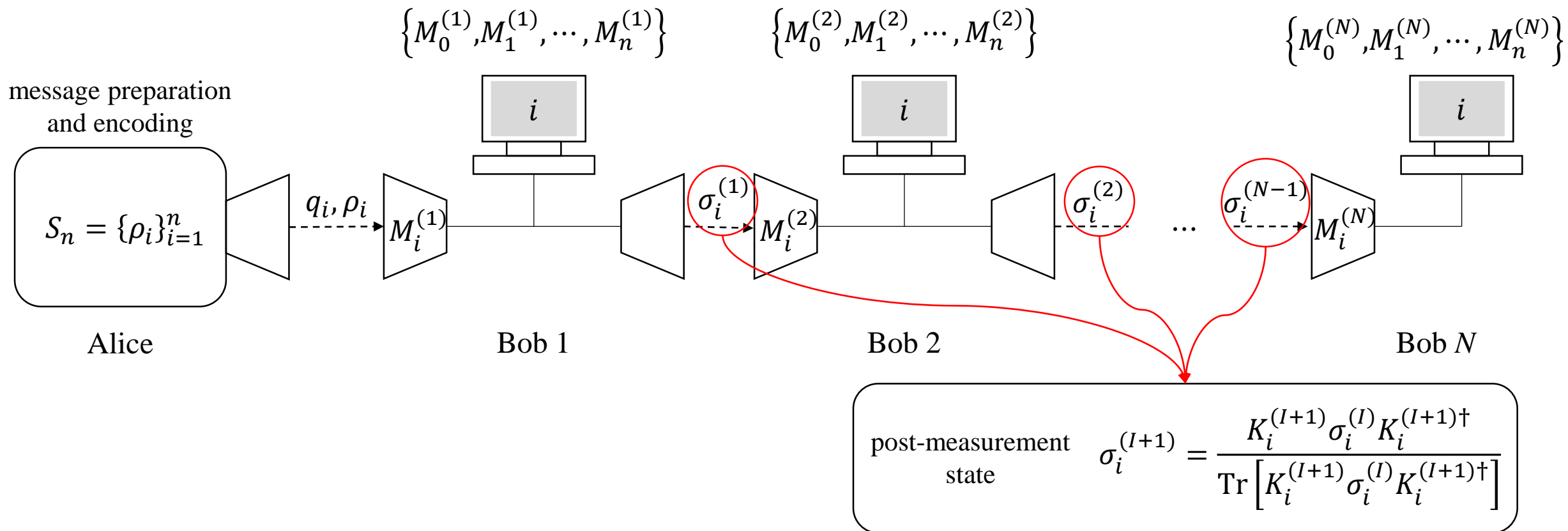


Scenario of sequential state discrimination

Multiparty QKD based on sequential state discrimination

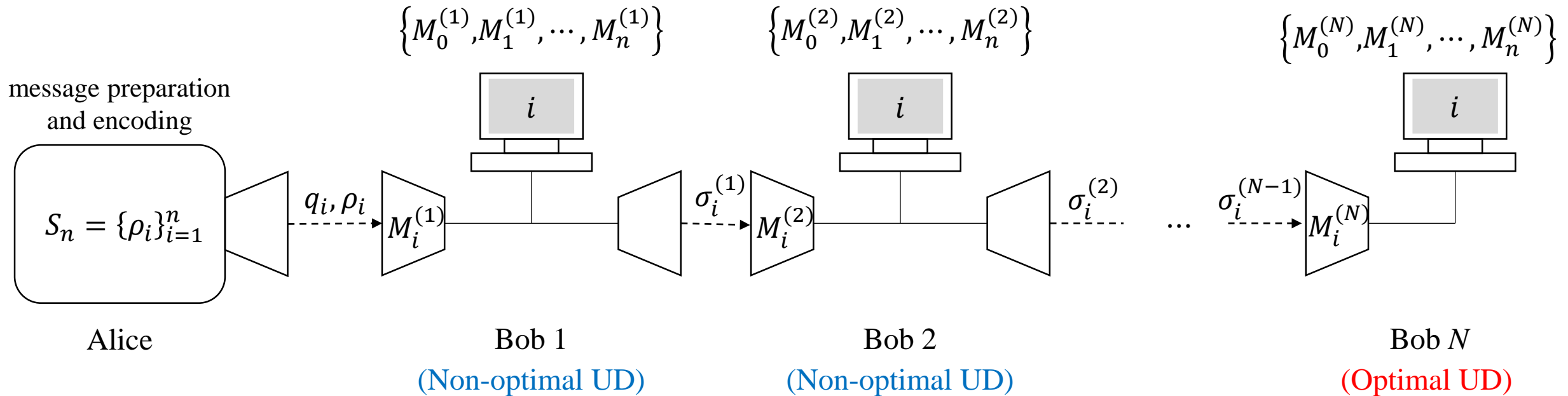


Multiparty QKD based on sequential state discrimination



Scenario of sequential state discrimination

Multiparty QKD based on sequential state discrimination



POVM for unambiguous discrimination

$p(j i) = \text{Tr} \rho_i M_j$		
\swarrow Born's rule	\searrow POVM condition	
$p(j i) \geq 0 \quad \forall i, j$	$M_i \geq 0 \quad \forall i \in \{0, \dots, n\}$	Positivity
$p(j i) \in \mathbb{R} \quad \forall i, j$	$M_i = M_i^\dagger \quad \forall i \in \{0, \dots, n\}$	Hermitian
$\sum_j p(j i) = 1 \quad \forall i$	$M_0 + M_1 + \dots + M_n = I$	Completeness
$p(j i) = 0 \quad \forall i \neq j$	$\text{Tr} \rho_i M_j = \delta_{ij} \text{Tr} \rho_i M_i \quad \forall i \neq j$	Unambiguous discrimination

POVM for unambiguous discrimination

- I. $M_i \geq 0, M_i = M_i^\dagger \quad \forall i \in \{0, \dots, n\}$
- II. $M_0 + M_1 + \dots + M_n = I$
- III. $\text{Tr} \rho_i M_j = \delta_{ij} \text{Tr} \rho_i M_i \quad \forall i \neq j$

Theorem 1. [T. Rudolph *et al.*] If $\text{supp}(\rho_i)$ satisfies $\text{supp}(\rho_i) \not\subseteq \bigcup_{j \neq i} \text{supp}(\rho_j)$ for all $\rho_i \in S_n$, then there exists POVM that performs unambiguous discrimination on S_n .

Constructing optimization problem

POVM for unambiguous discrimination

- I. $M_i \geq 0, M_i = M_i^\dagger \quad \forall i \in \{0, \dots, n\}$
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However, exploiting Theorem 1 is quite difficult.



- I. $M_i \geq 0, M_i = M_i^\dagger \quad \forall i \in \{0, \dots, n\}$
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Theorem 2. [A. Chefles] If \bar{S}_n is a set of linearly independent pure states, then there exists POVM that performs unambiguous discrimination on \bar{S}_n .

Constructing optimization problem

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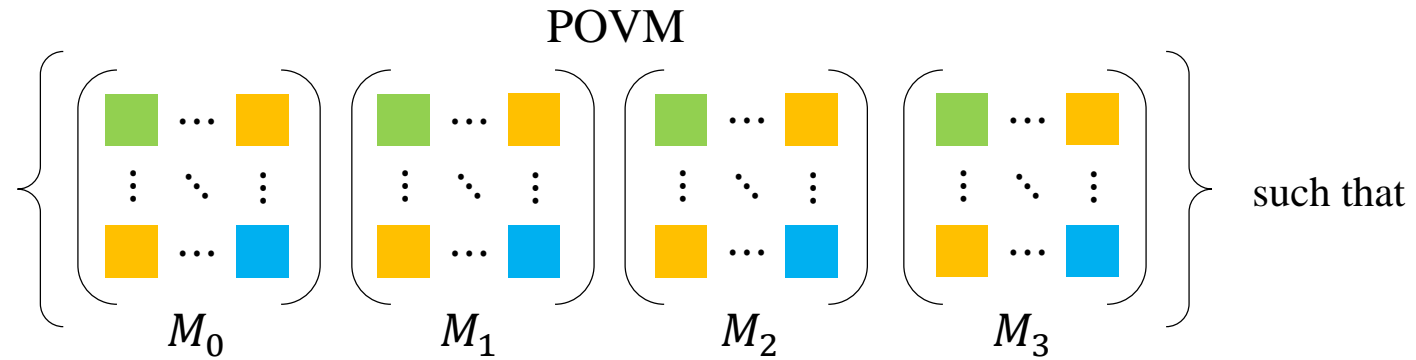
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Theorem 2. [A. Chefles] If \bar{S}_n is a set of linearly independent pure states, then there exists POVM that performs unambiguous discrimination on \bar{S}_n .

Simplified Proof. $M_i = \alpha_i |\tilde{\psi}_i\rangle \langle \tilde{\psi}_i| \quad (i = 1, \dots, n), \quad \alpha_i \geq 0 \wedge \alpha_i \in \mathbb{R}$
[D. Ha and Y. Kwon]
 $|\tilde{\psi}_i\rangle = \sum_{j=1}^n G_{ji}^{-1} |\psi_j\rangle, \quad G = \{\langle \psi_i | \psi_j \rangle\}_{i,j=1}^n$: Gram matrix
 $M_0 = I - M_1 - M_2 - \dots - M_n$

POVM for unambiguous discrimination

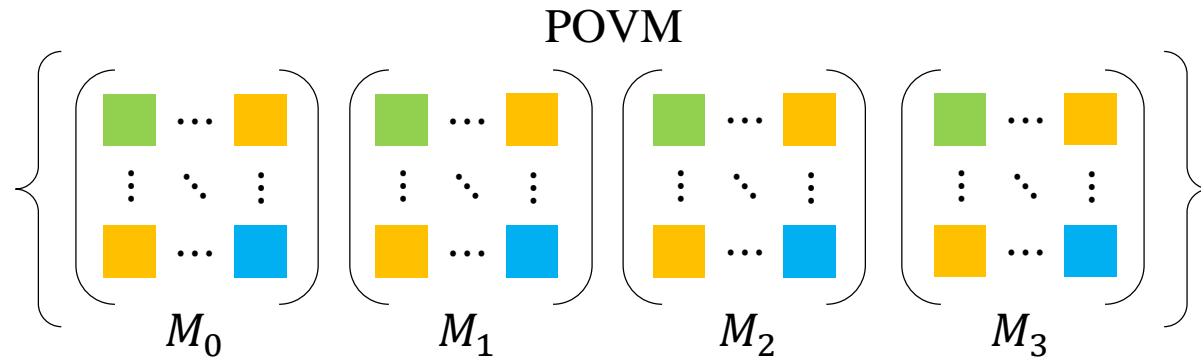
Power of Theorem 2.



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POVM for unambiguous discrimination

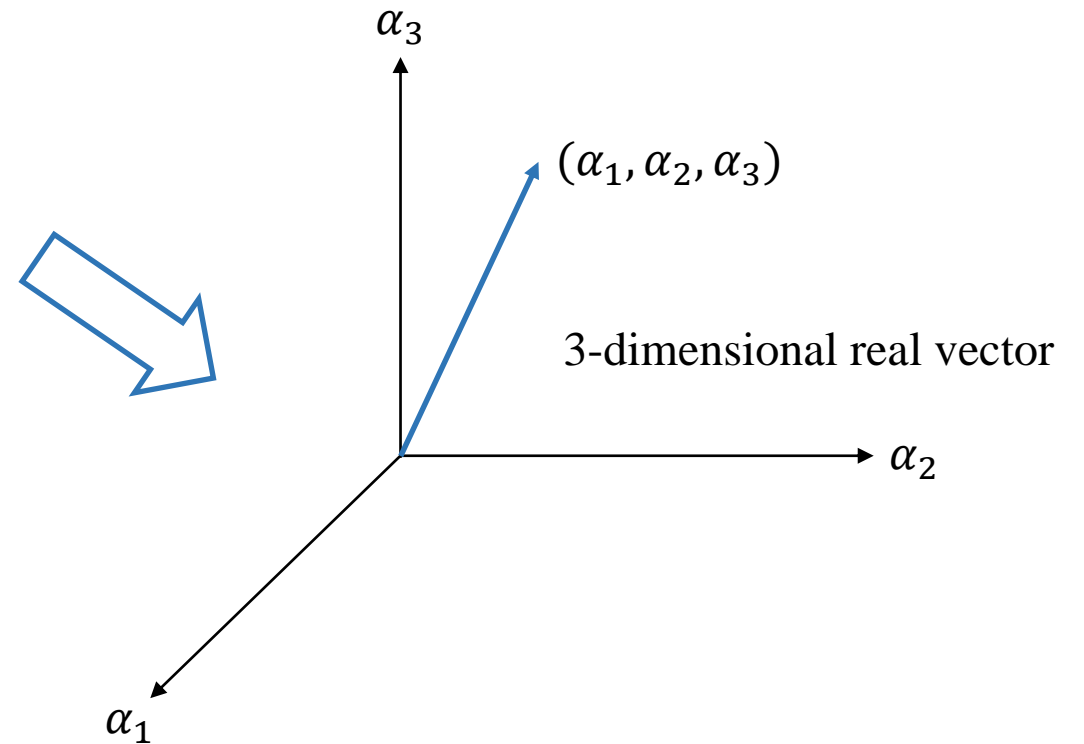
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$$M_i = \alpha_i |\tilde{\psi}_i\rangle\langle\tilde{\psi}_i| \quad (i = 1, \dots, n), \quad \alpha_i \geq 0 \wedge \alpha_i \in \mathbb{R}$$

$$|\tilde{\psi}_i\rangle = \sum_{j=1}^n G_{ji}^{-1} |\psi_j\rangle, \quad G = \{\langle\psi_i|\psi_j\rangle\}_{i,j=1}^n: \text{Gram matrix}$$

$$M_0 = I - M_1 - M_2 - \dots - M_n$$



POVM for unambiguous discrimination

Theorem 3. [D. Ha and Y. Kwon] Let define Hermitian matrix $\bar{M} = \{\langle \psi_i | M_0 | \psi_j \rangle\}_{i,j=1}^n$ and all $m \times m$ ($m < n$) principal submatrices \bar{M}_m .
 M_0 is positive-semidefinite if and only if every \bar{M} and $\forall \bar{M}_m$ is positive-semidefinite.

Constructing optimization problem

POVM for unambiguous discrimination

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Example. Let consider POVM

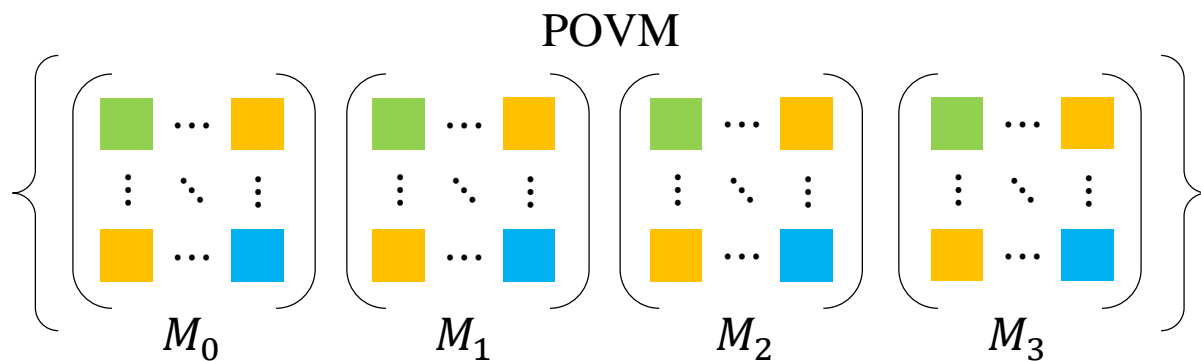
$$\left\{ \begin{matrix} \begin{pmatrix} \color{green}\blacksquare & \dots & \color{orange}\blacksquare \\ \vdots & \ddots & \vdots \\ \color{orange}\blacksquare & \dots & \color{blue}\blacksquare \end{pmatrix} & \begin{pmatrix} \color{green}\blacksquare & \dots & \color{orange}\blacksquare \\ \vdots & \ddots & \vdots \\ \color{orange}\blacksquare & \dots & \color{blue}\blacksquare \end{pmatrix} & \begin{pmatrix} \color{green}\blacksquare & \dots & \color{orange}\blacksquare \\ \vdots & \ddots & \vdots \\ \color{orange}\blacksquare & \dots & \color{blue}\blacksquare \end{pmatrix} & \begin{pmatrix} \color{green}\blacksquare & \dots & \color{orange}\blacksquare \\ \vdots & \ddots & \vdots \\ \color{orange}\blacksquare & \dots & \color{blue}\blacksquare \end{pmatrix} \\ M_0 & M_1 & M_2 & M_3 \end{matrix} \right\}$$

$\begin{pmatrix} \color{green}\blacksquare & \dots & \color{orange}\blacksquare \\ \vdots & \ddots & \vdots \\ \color{orange}\blacksquare & \dots & \color{blue}\blacksquare \end{pmatrix} M_0 \geq 0$ If and only if

$\bar{M} = \begin{pmatrix} \circ & \circ & \circ \\ \circ & \circ & \circ \\ \circ & \circ & \circ \end{pmatrix}$ $\bar{M}_1 = \begin{pmatrix} \circ \\ \circ \end{pmatrix}$ $\bar{M}_2 = \begin{pmatrix} \circ & \circ \\ \circ & \circ \end{pmatrix}$ $\bar{M}_3 = \begin{pmatrix} \circ \\ \circ \\ \circ \end{pmatrix}$ are positive-semidefinite.

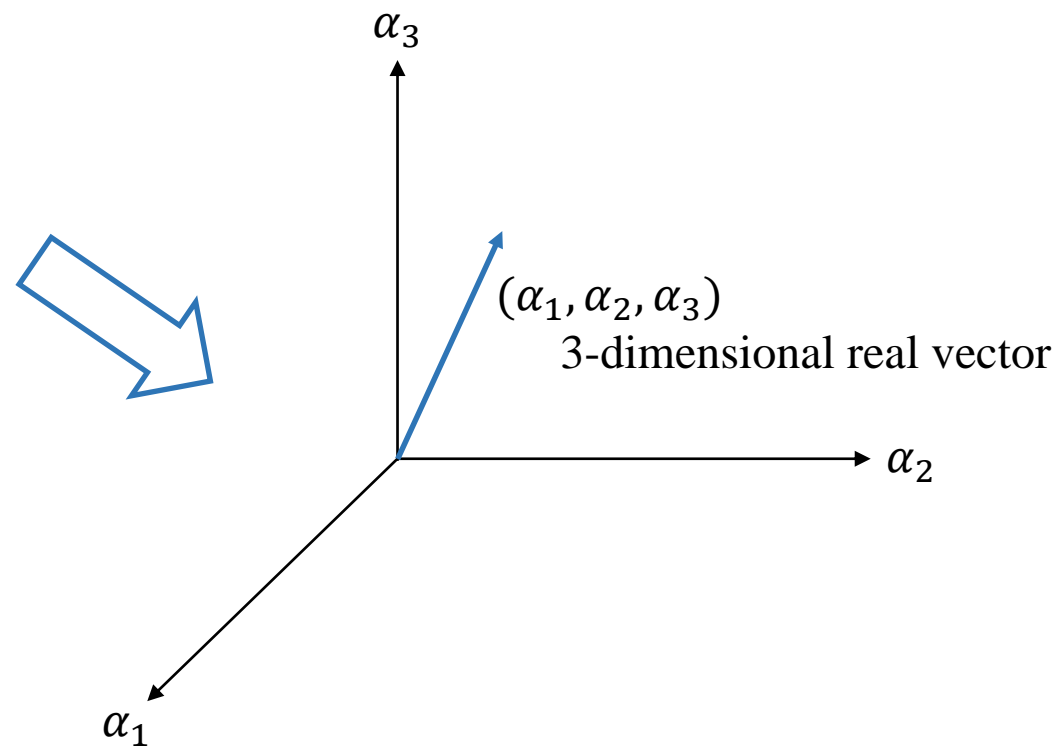
POVM for unambiguous discrimination

Power of Theorem 3.



Constraint

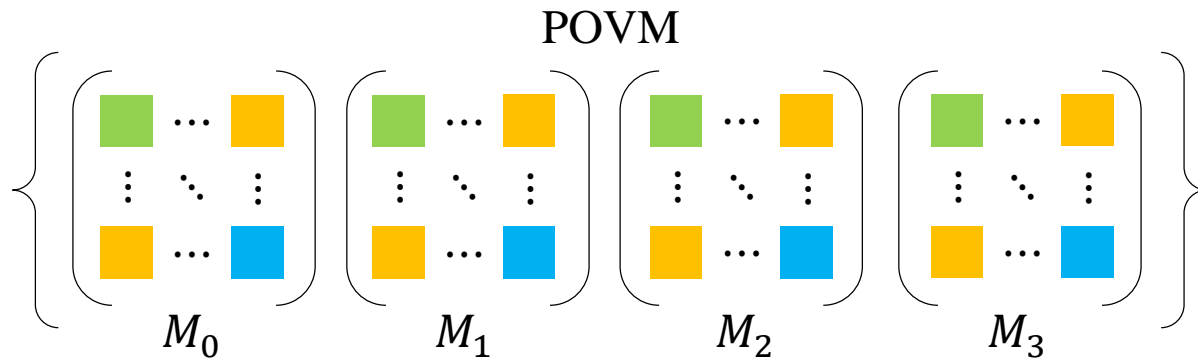
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Constructing optimization problem

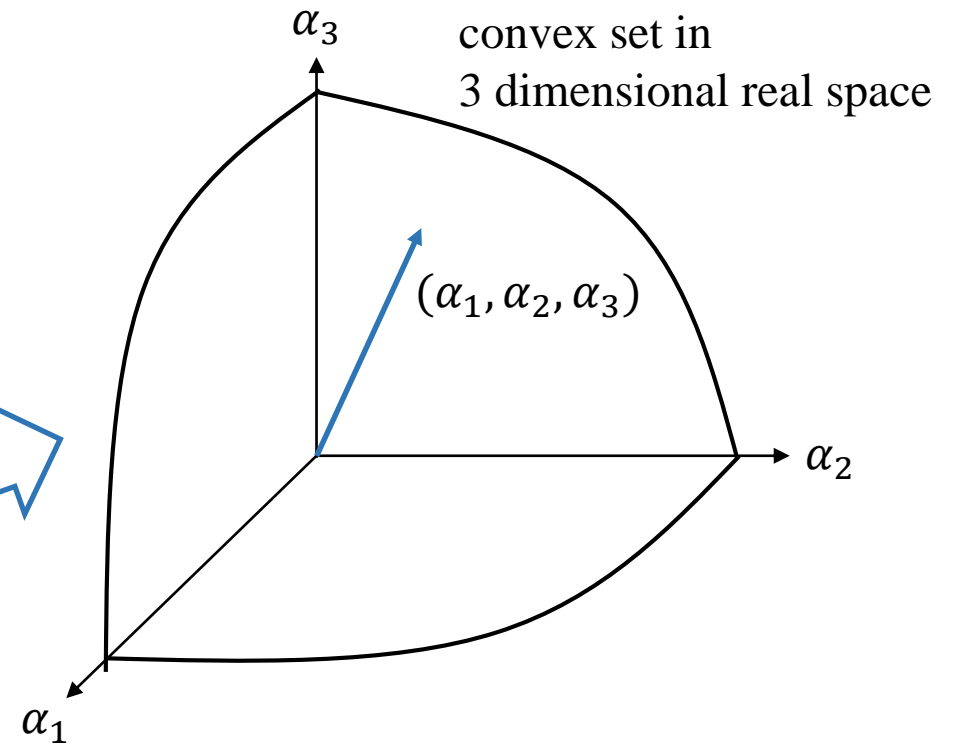
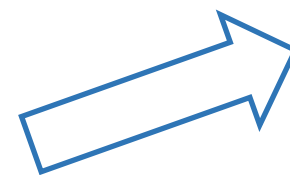
POVM for unambiguous discrimination

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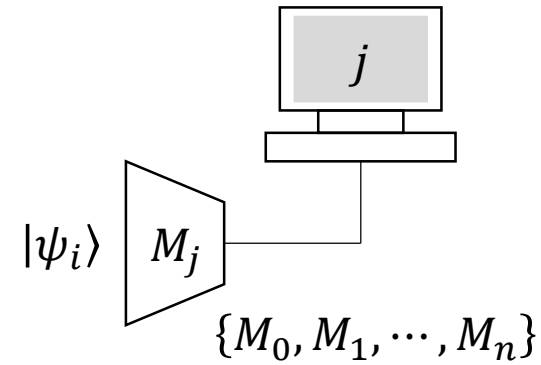
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POVM for unambiguous discrimination

POVM

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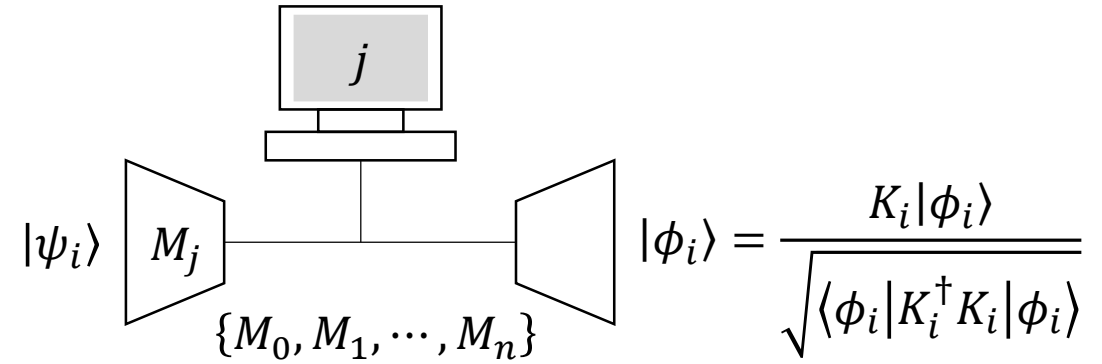


Constructing optimization problem

Constructing Kraus operator

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Kraus operator

- I. $M_i = K_i^\dagger K_i$
- II. $K_i |\psi_i\rangle \propto |\phi_i\rangle$ where
 $\{|\phi_1\rangle, \dots, |\phi_n\rangle\}$: linearly independent

Constructing optimization problem

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POVM

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$i \neq 0$

K_i can be obtained using singular value decomposition:

$$K_i = U_i \sqrt{M_i}$$

Kraus operator

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Constructing optimization problem

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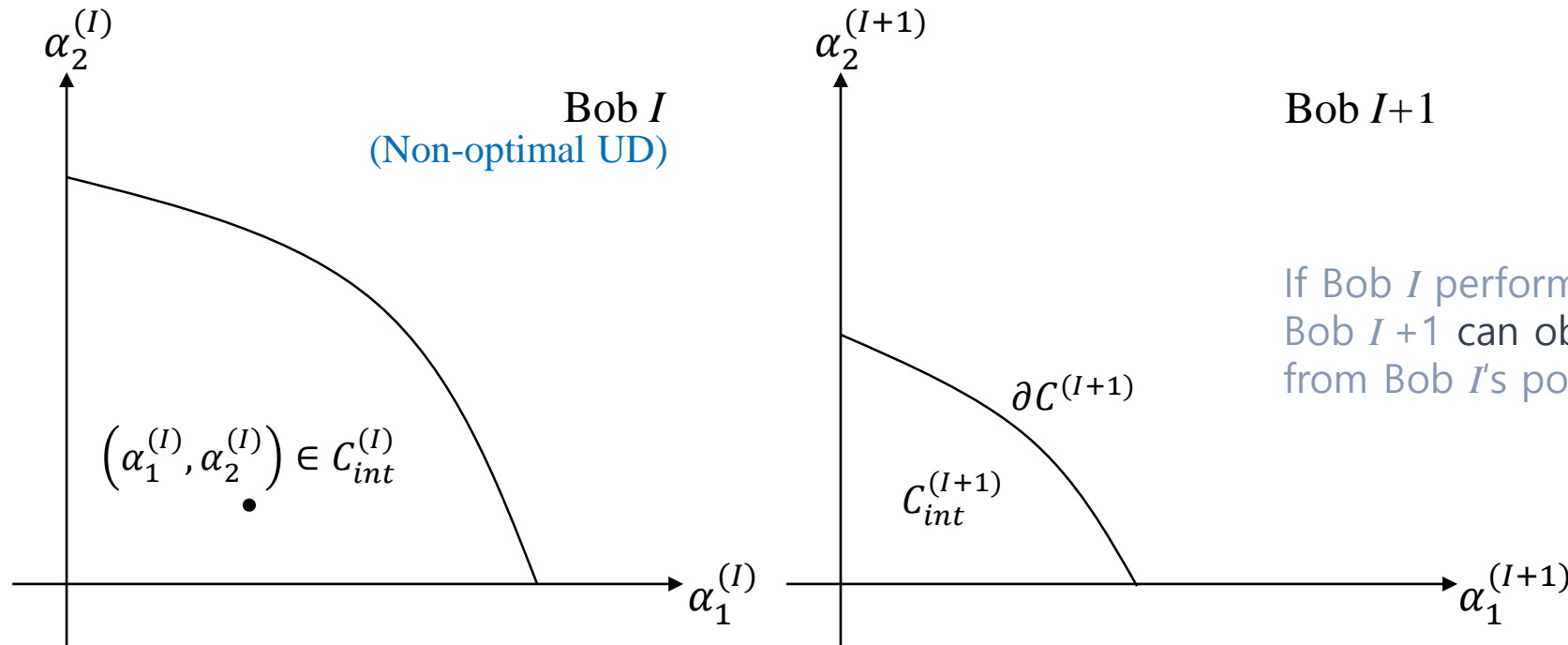
$i = 0$

K_i can be obtained using following lemma:

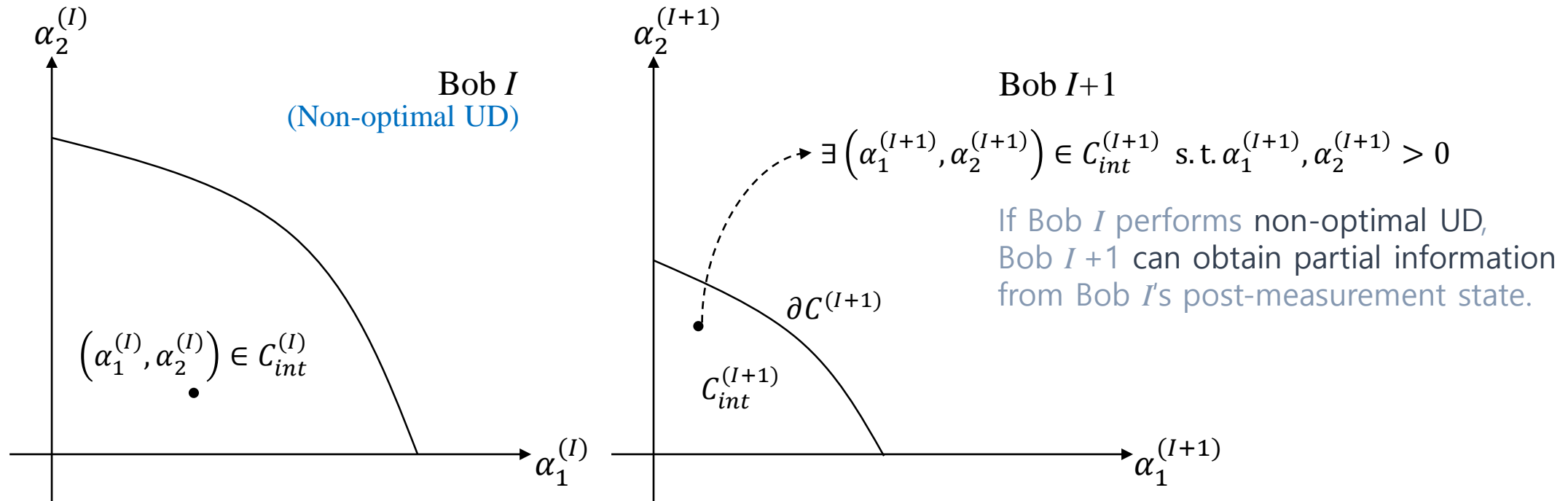
Lemma. [M. Namkung and Y. Kwon]

$$M_0 = K_0^\dagger K_0 \text{ if and only if } \langle \psi_i | M_0 | \psi_j \rangle = \langle \psi_i | K_0^\dagger K_0 | \psi_j \rangle.$$

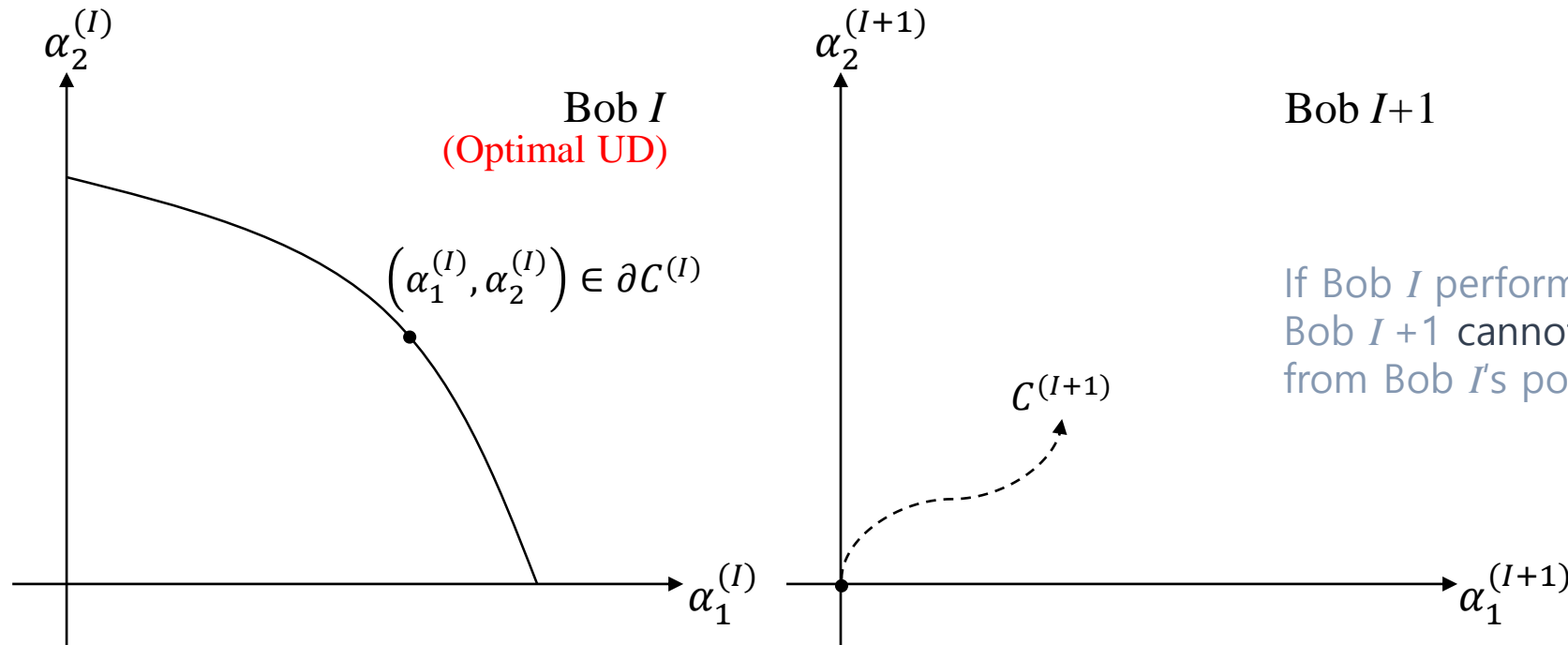
Property of Kraus operator



Property of Kraus operator



Property of Kraus operator



If Bob I performs optimal UD,
Bob $I+1$ cannot obtain partial information
from Bob I 's post-measurement state.

Optimization problem (result)

$$\begin{aligned} \text{maximize } & P_S^{(B_1, \dots, B_N)} = \sum_{i=1}^n q_i \alpha_i^{(1)} \alpha_i^{(2)} \alpha_i^{(3)} \times \dots \times \alpha_i^{(N)} \\ \text{subject to } & (\alpha_1^{(I)}, \dots, \alpha_n^{(I)}) \in C_{int}^{(I)} \quad \forall I \leq N-1 \\ & (\alpha_1^{(I)}, \dots, \alpha_n^{(I)}) \in \partial C^{(N)} \end{aligned}$$

Remark: $\alpha_i^{(I)}$ is probability that Bob I obtains outcome i , given that Alice prepares $|\psi_i\rangle$.

Optimization problem (two pure states, three receivers)

$$\text{maximize } P_S^{(B_1, B_2, B_3)} = q_1 \alpha_1 \beta_1 \gamma_1 + q_2 \alpha_2 \beta_2 \gamma_2$$

$$\begin{aligned} \text{subject to } (1 - \alpha_1)(1 - \alpha_2) &> |\langle \psi_1 | \psi_2 \rangle|^2 \\ (1 - \beta_1)(1 - \beta_2) &> \left| \left\langle \phi_1^{(B_1)} \middle| \phi_2^{(B_1)} \right\rangle \right|^2 \\ (1 - \gamma_1)(1 - \gamma_2) &= \left| \left\langle \phi_1^{(B_2)} \middle| \phi_2^{(B_2)} \right\rangle \right|^2 \end{aligned}$$

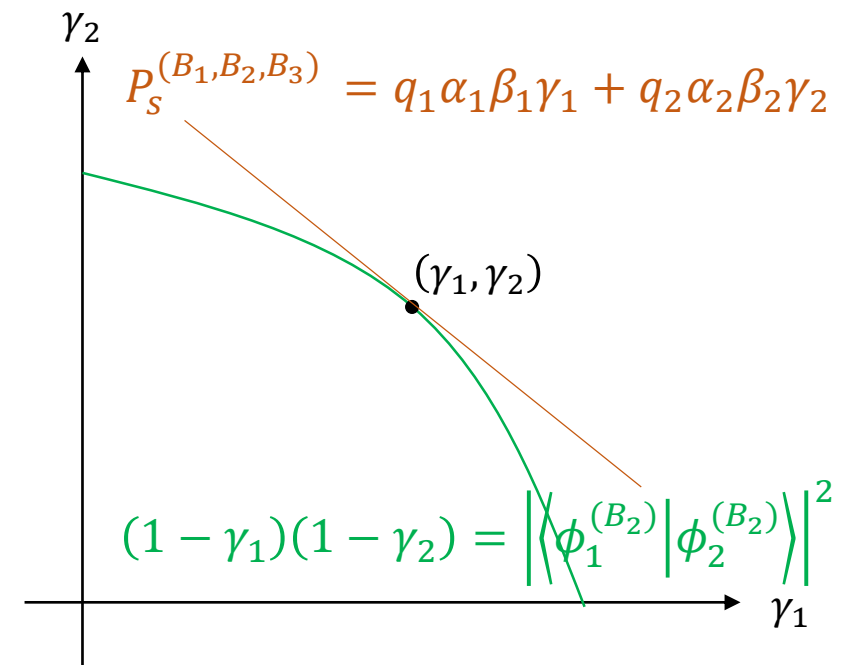
Optimization problem (two pure states, three receivers)

$$\text{maximize } P_S^{(B_1, B_2, B_3)} = q_1 \alpha_1 \beta_1 \gamma_1 + q_2 \alpha_2 \beta_2 \gamma_2$$

$$\text{subject to } (1 - \alpha_1)(1 - \alpha_2) > |\langle \psi_1 | \psi_2 \rangle|^2$$

$$(1 - \beta_1)(1 - \beta_2) > \left| \left\langle \phi_1^{(B_1)} \middle| \phi_2^{(B_1)} \right\rangle \right|^2$$

$$(1 - \gamma_1)(1 - \gamma_2) = \left| \left\langle \phi_1^{(B_2)} \middle| \phi_2^{(B_2)} \right\rangle \right|^2$$



Remark: (γ_1, γ_2) is obtained by finding a tangential point between a plane $P_S^{(B_1, B_2, B_3)} = q_1 \alpha_1 \beta_1 \gamma_1 + q_2 \alpha_2 \beta_2 \gamma_2$

and a surface $(1 - \gamma_1)(1 - \gamma_2) = \left| \left\langle \phi_1^{(B_1)} \middle| \phi_2^{(B_2)} \right\rangle \right|^2$.

Constructing optimization problem

Optimization problem (two pure states, three receivers)

$$\begin{aligned} \text{maximize } P_S^{(B_1, B_2, B_3)} &= q_1 \alpha_1 \beta_1 + q_2 \alpha_2 \beta_2 - 2 |\langle \psi_1 | \psi_2 \rangle| \sqrt{\frac{q_1 q_2 \alpha_1 \alpha_2 \beta_1 \beta_2}{(1 - \alpha_1)(1 - \alpha_2)(1 - \beta_1)(1 - \beta_2)}} \\ \text{subject to } (1 - \alpha_1)(1 - \alpha_2) &> |\langle \psi_1 | \psi_2 \rangle|^2 \\ \beta_2 &\leq \frac{\beta_1(1 - \beta_1)}{\beta_1(1 - \beta_1) + X(\alpha_1, \alpha_2)}, \quad X(\alpha_1, \alpha_2) = \frac{q_2 \alpha_2}{q_1 \alpha_1} \frac{|\langle \psi_1 | \psi_2 \rangle|^2}{(1 - \alpha_1)(1 - \alpha_2)} \\ \beta_1 &\leq \frac{\beta_2(1 - \beta_2)}{\beta_2(1 - \beta_2) + Y(\alpha_1, \alpha_2)}, \quad Y(\alpha_1, \alpha_2) = \frac{q_1 \alpha_1}{q_2 \alpha_2} \frac{|\langle \psi_1 | \psi_2 \rangle|^2}{(1 - \alpha_1)(1 - \alpha_2)} \end{aligned}$$

Remark: In general, this optimization problem can be solved by using nonlinear programming, including random search method, sequential linear programming, and penalty function method.

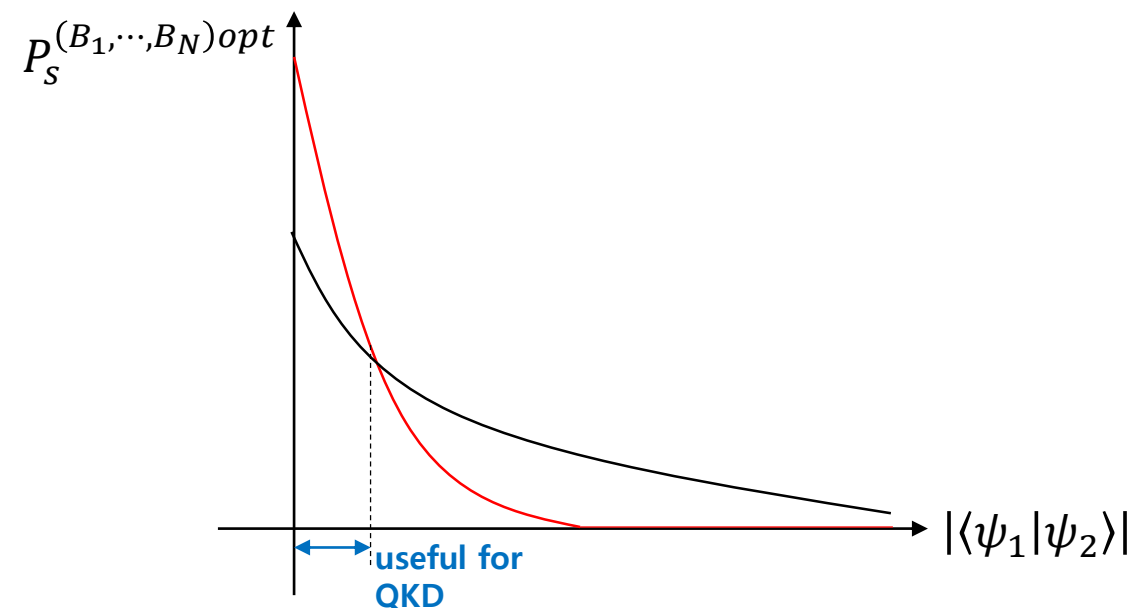
Optimal success probability (two pure states with equal prior probabilities, N receivers)

Discriminating two states: $P_S^{(B_1, \dots, B_N)opt} = (1 - |\langle \psi_1 | \psi_2 \rangle|^{1/N})^N$ $|\langle \psi_1 | \psi_2 \rangle| < (2^{1/N} - 1)^N$

Discriminating one out of two states: $P_S^{(B_1, \dots, B_N)opt} = \frac{1}{2} (1 - |\langle \psi_1 | \psi_2 \rangle|^{2/N})^N$ $|\langle \psi_1 | \psi_2 \rangle| \geq (2^{1/N} - 1)^N$

Remark 1: Optimal success probability satisfies the result of [J. A. Bergou *et al.*], and [C.-Q. Pang *et al.*].

Remark 2: Sequential state discrimination of two pure states is suitable for multiparty QKD, when the number of receivers is not too many.

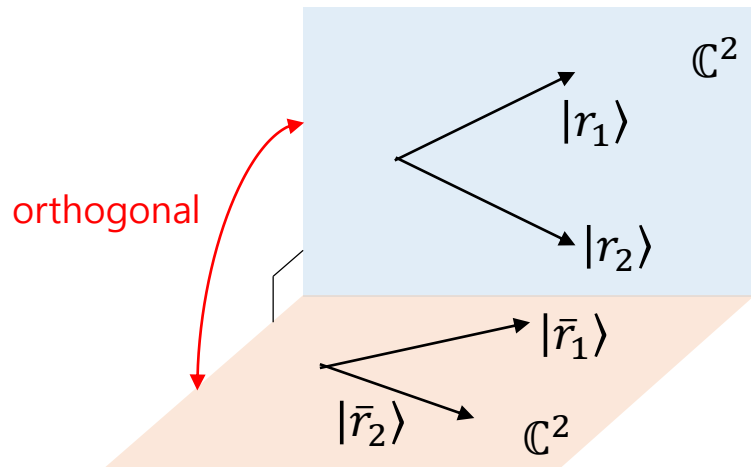


Extending problem to mixed states case

$$\rho_1 = r_1 |r_1\rangle\langle r_1| + \bar{r}_1 |\bar{r}_1\rangle\langle \bar{r}_1|$$

$$\rho_2 = r_2 |r_2\rangle\langle r_2| + \bar{r}_2 |\bar{r}_2\rangle\langle \bar{r}_2|$$

(Sketch of POVM)



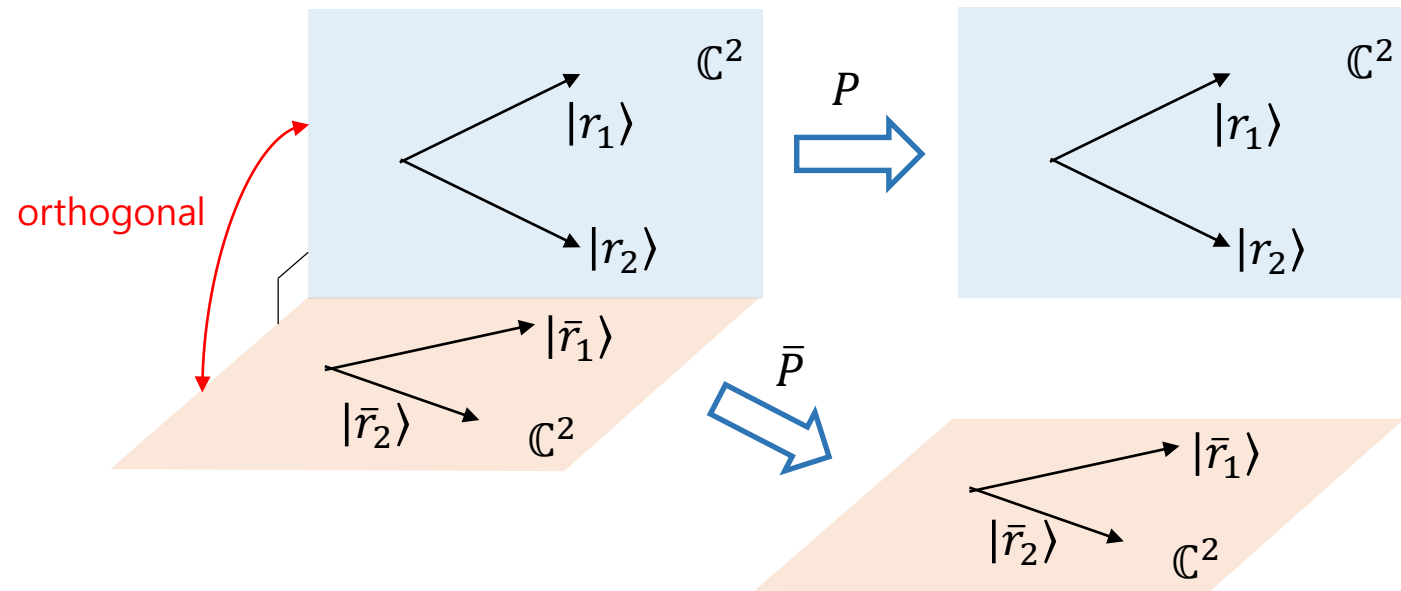
Constructing optimization problem

Extending problem to mixed states case

$$\rho_1 = r_1 |r_1\rangle\langle r_1| + \bar{r}_1 |\bar{r}_1\rangle\langle \bar{r}_1|$$

$$\rho_2 = r_2 |r_2\rangle\langle r_2| + \bar{r}_2 |\bar{r}_2\rangle\langle \bar{r}_2|$$

(Sketch of POVM)



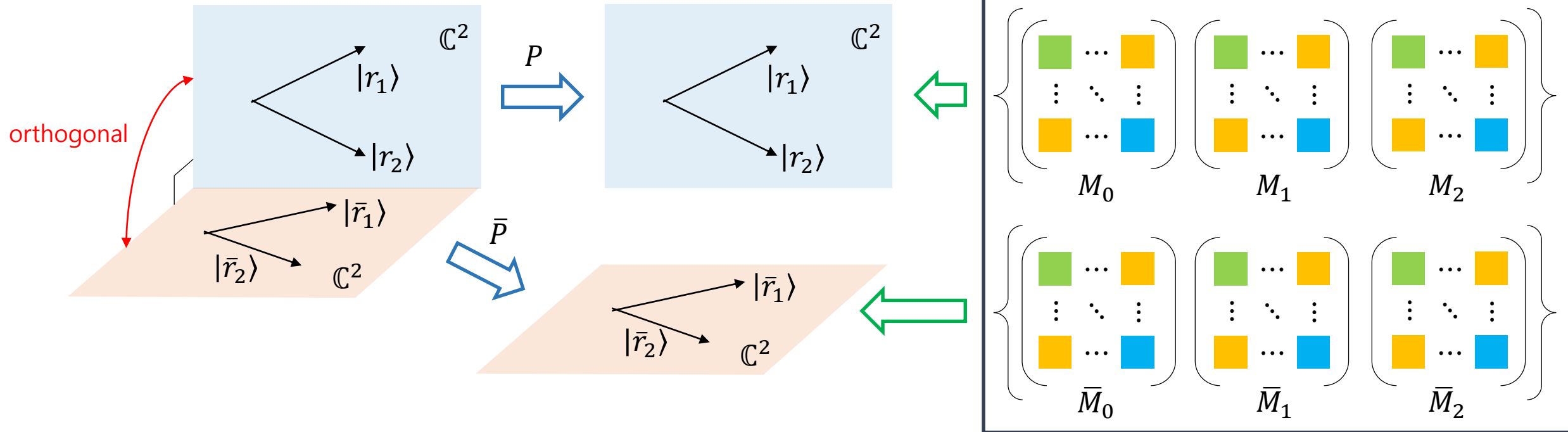
Constructing optimization problem

Extending problem to mixed states case

$$\rho_1 = r_1 |r_1\rangle\langle r_1| + \bar{r}_1 |\bar{r}_1\rangle\langle \bar{r}_1|$$

$$\rho_2 = r_2 |r_2\rangle\langle r_2| + \bar{r}_2 |\bar{r}_2\rangle\langle \bar{r}_2|$$

(Sketch of POVM)

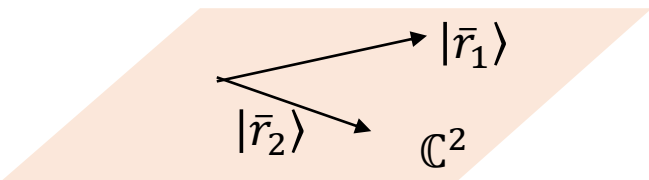
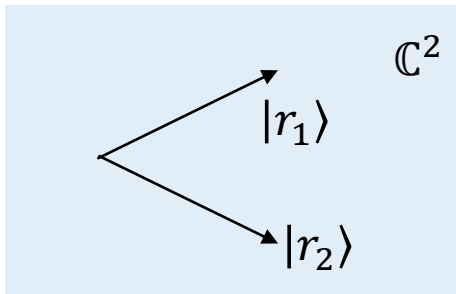


U. Herzog, Optimum unambiguous discrimination of two mixed states and application to a class of similar states, Phys. Rev. A **75**, 052309 (2007).

Extending problem to mixed states case

$$\rho_1 = r_1|r_1\rangle\langle r_1| + \bar{r}_1|\bar{r}_1\rangle\langle\bar{r}_1|$$
$$\rho_2 = r_2|r_2\rangle\langle r_2| + \bar{r}_2|\bar{r}_2\rangle\langle\bar{r}_2|$$

(Sketch of Kraus operator)



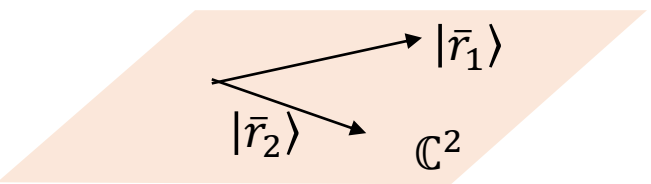
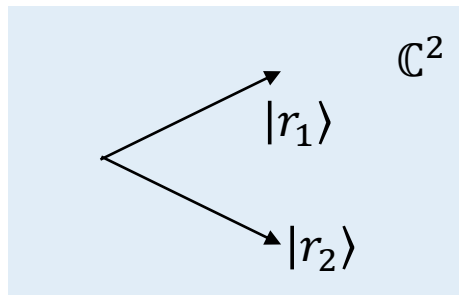
Constructing optimization problem

Extending problem to mixed states case

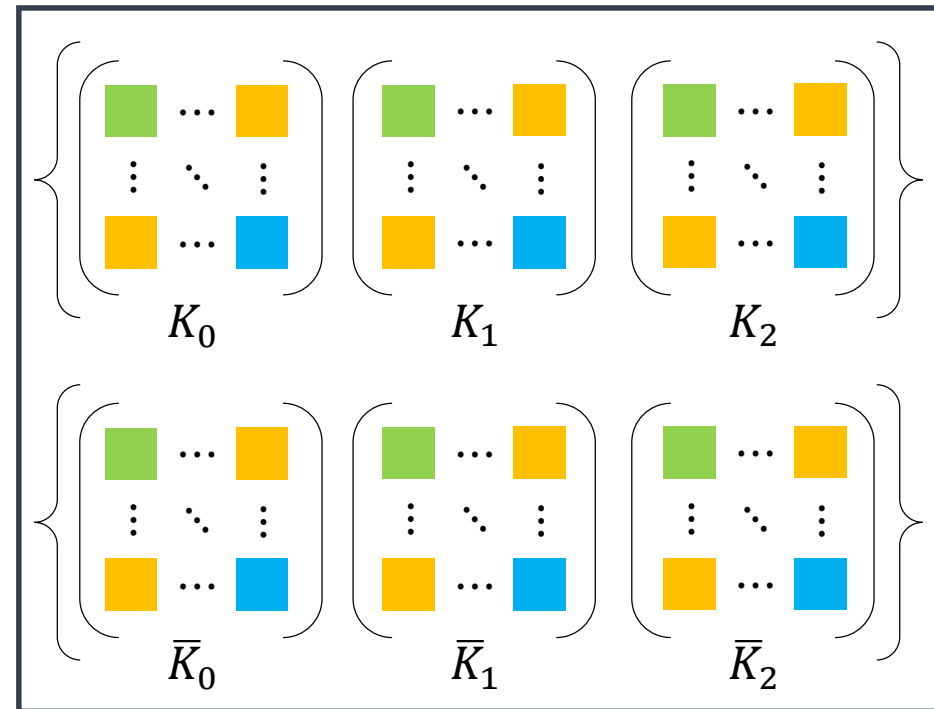
$$\rho_1 = r_1|r_1\rangle\langle r_1| + \bar{r}_1|\bar{r}_1\rangle\langle\bar{r}_1|$$

$$\rho_2 = r_2|r_2\rangle\langle r_2| + \bar{r}_2|\bar{r}_2\rangle\langle\bar{r}_2|$$

(Sketch of Kraus operator)



$$\{K_i \oplus \bar{K}_j\}_{i,j=0}^2$$



Constructing optimization problem

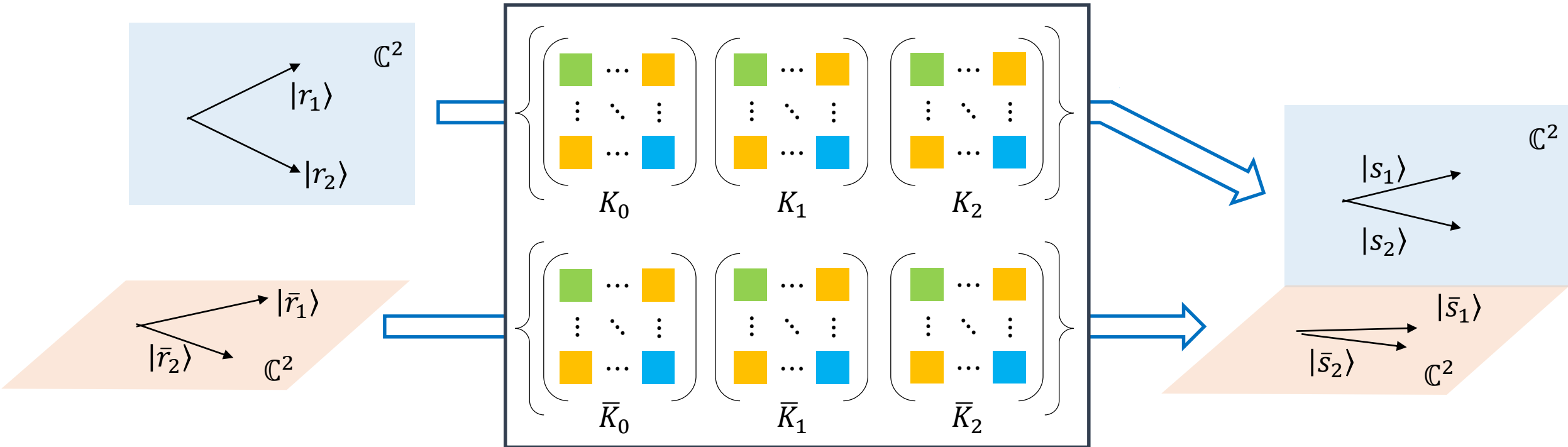
Extending problem to mixed states case

$$\rho_1 = r_1|r_1\rangle\langle r_1| + \bar{r}_1|\bar{r}_1\rangle\langle\bar{r}_1|$$

$$\rho_2 = r_2|r_2\rangle\langle r_2| + \bar{r}_2|\bar{r}_2\rangle\langle\bar{r}_2|$$

(Sketch of Kraus operator)

$$\{K_i \oplus \bar{K}_j\}_{i,j=0}^2$$



M. Namkung and Y. Kwon, Optimal sequential state discrimination between two mixed quantum states, Phys. Rev. A **96**, 022318 (2017).

Extending problem to mixed states case (result)

$$\begin{aligned}
 &\text{maximize } P_S^{(B_1, \dots, B_N)} = q_1 \left(r_1 \prod_{l=1}^N \alpha_1^{(l)} + r_1 \prod_{l=1}^N \bar{\alpha}_1^{(l)} \right) + q_2 \left(r_2 \prod_{l=1}^N \alpha_2^{(l)} + r_2 \prod_{l=1}^N \bar{\alpha}_2^{(l)} \right) \\
 &\text{subject to } \begin{aligned}
 &(1 - \alpha_1^{(1)}) (1 - \alpha_2^{(1)}) > |\langle r_1 | r_2 \rangle|^2 && (1 - \bar{\alpha}_1^{(1)}) (1 - \bar{\alpha}_2^{(1)}) > |\langle \bar{r}_1 | \bar{r}_2 \rangle|^2 \\
 &(1 - \alpha_1^{(I)}) (1 - \alpha_2^{(I)}) > \left| \langle s_1^{(I-1)} | s_2^{(I-1)} \rangle \right|^2 && (1 - \bar{\alpha}_1^{(I)}) (1 - \bar{\alpha}_2^{(I)}) > \left| \langle \bar{s}_1^{(I-1)} | \bar{s}_2^{(I-1)} \rangle \right|^2 \\
 &(1 - \alpha_1^{(N)}) (1 - \alpha_2^{(N)}) = \left| \langle s_1^{(N-1)} | s_2^{(N-1)} \rangle \right|^2 && (1 - \bar{\alpha}_1^{(N)}) (1 - \bar{\alpha}_2^{(N)}) = \left| \langle \bar{s}_1^{(N-1)} | \bar{s}_2^{(N-1)} \rangle \right|^2
 \end{aligned}
 \end{aligned}$$

Extending problem to mixed states case (result)

$$\text{maximize } p_s^{(B_1, \dots, B_N)} = q_1 r_1 \prod_{l=1}^N \alpha_1^{(l)} + q_2 r_2 \prod_{l=1}^N \alpha_2^{(l)}$$

$$\begin{aligned} \text{subject to } & (1 - \alpha_1^{(1)}) (1 - \alpha_2^{(1)}) > |\langle r_1 | r_2 \rangle|^2 \\ & (1 - \alpha_1^{(I)}) (1 - \alpha_2^{(I)}) > \left| \langle s_1^{(I-1)} | s_2^{(I-1)} \rangle \right|^2 \\ & (1 - \alpha_1^{(N)}) (1 - \alpha_2^{(N)}) = \left| \langle s_1^{(N-1)} | s_2^{(N-1)} \rangle \right|^2 \end{aligned}$$

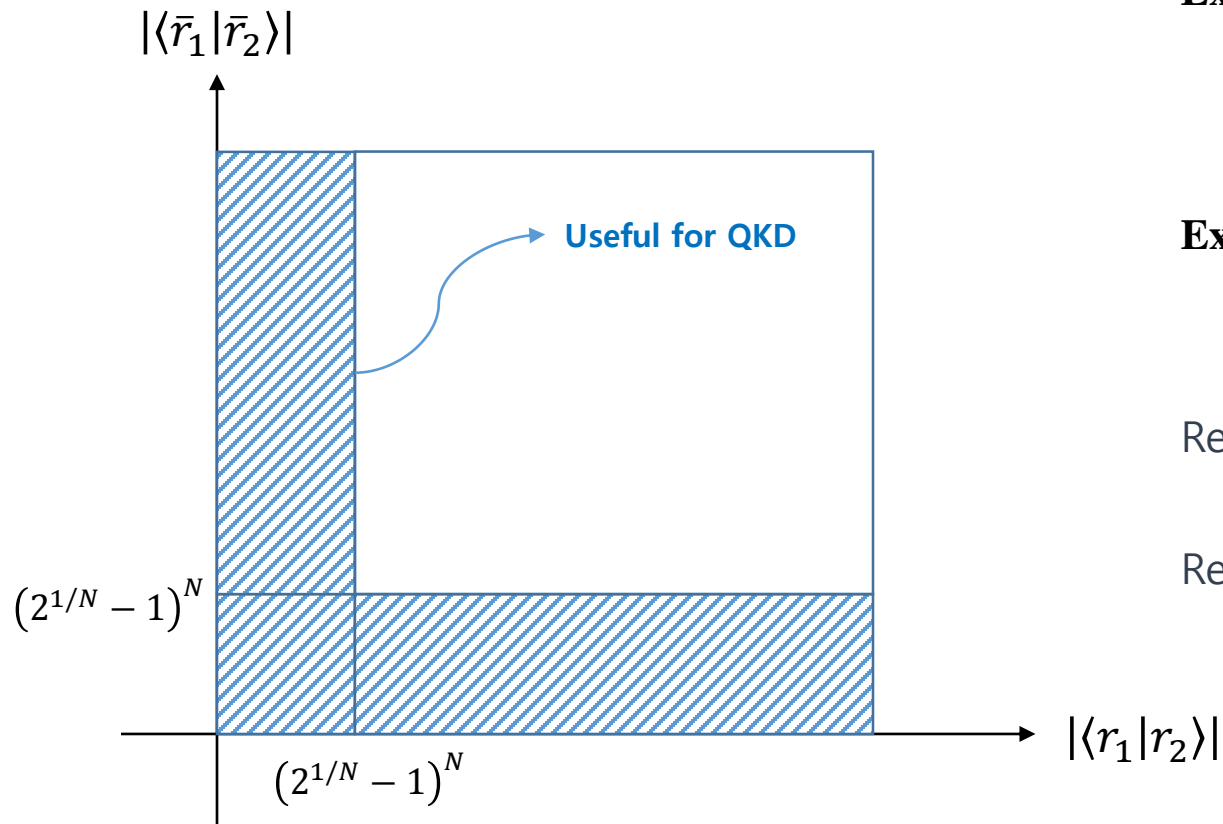
sub-optimization problem 1

$$\text{maximize } \bar{p}_s^{(B_1, \dots, B_N)} = q_1 \bar{r}_1 \prod_{l=1}^N \bar{\alpha}_1^{(l)} + q_2 \bar{r}_2 \prod_{l=1}^N \bar{\alpha}_2^{(l)}$$

$$\begin{aligned} \text{subject to } & (1 - \bar{\alpha}_1^{(1)}) (1 - \bar{\alpha}_2^{(1)}) > |\langle \bar{r}_1 | \bar{r}_2 \rangle|^2 \\ & (1 - \bar{\alpha}_1^{(I)}) (1 - \bar{\alpha}_2^{(I)}) > \left| \langle \bar{s}_1^{(I-1)} | \bar{s}_2^{(I-1)} \rangle \right|^2 \\ & (1 - \bar{\alpha}_1^{(N)}) (1 - \bar{\alpha}_2^{(N)}) = \left| \langle \bar{s}_1^{(N-1)} | \bar{s}_2^{(N-1)} \rangle \right|^2 \end{aligned}$$

sub-optimization problem 2

Extending problem to mixed states case (in case of $q_1 = q_2$, $r_1 = r_2 = r$, $\bar{r}_1 = \bar{r}_2 = \bar{r}$)



Example 1. $r = 0.6, \bar{r} = 0.4, s = 0.7, \bar{s} = 0.0001, N = 3$

$$P_s^{(B_1, \dots, B_N)opt} = 0.294443356418367$$

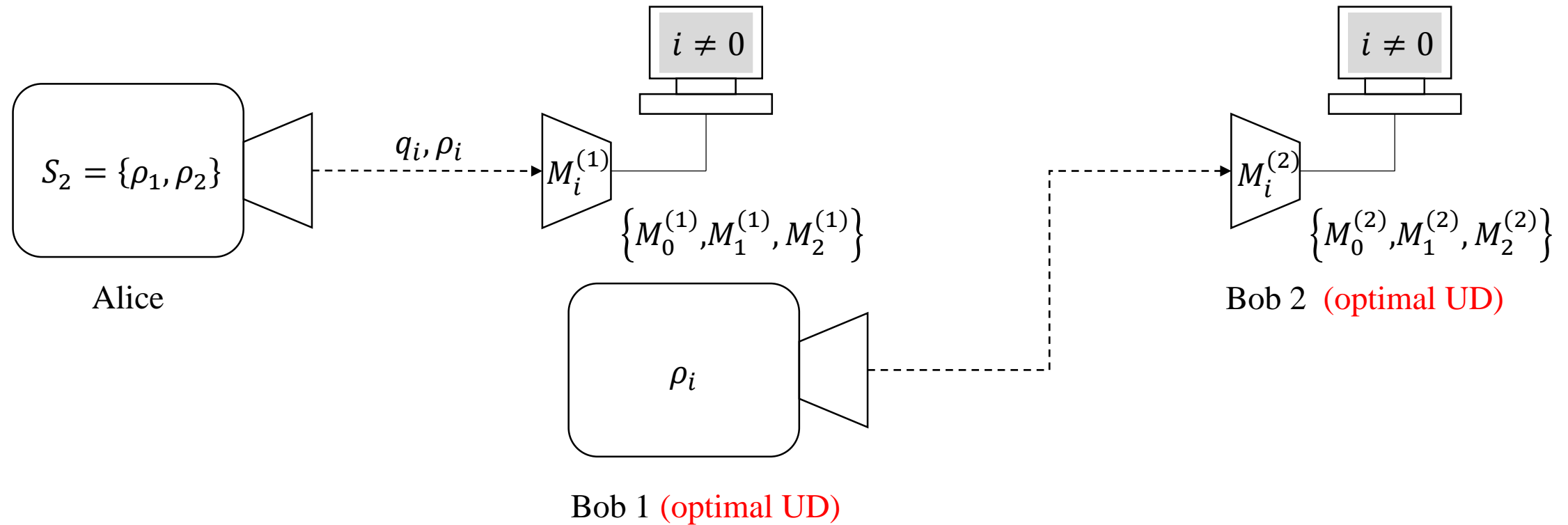
Example 2. $r = 0.6, \bar{r} = 0.4, s = 0.7, \bar{s} = 0.0001, N = 4$

$$P_s^{(B_1, \dots, B_N)opt} = 0.182986042905254$$

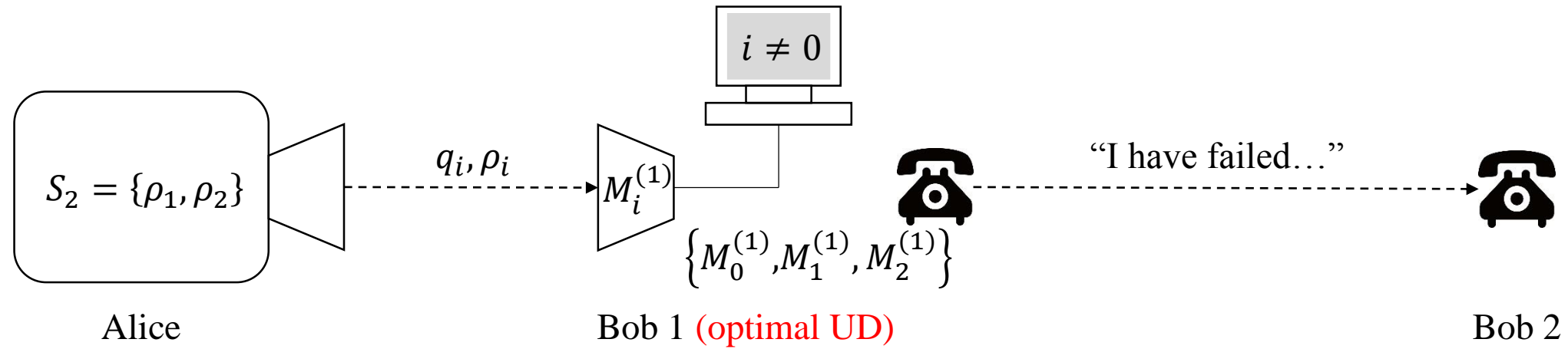
Remark 1: Sequential state discrimination of two mixed states can be performed with large success probability.

Remark 2: Sequential state discrimination can be applied to multiparty QKD, even if the number of receivers is too many.

Quantum reproducing

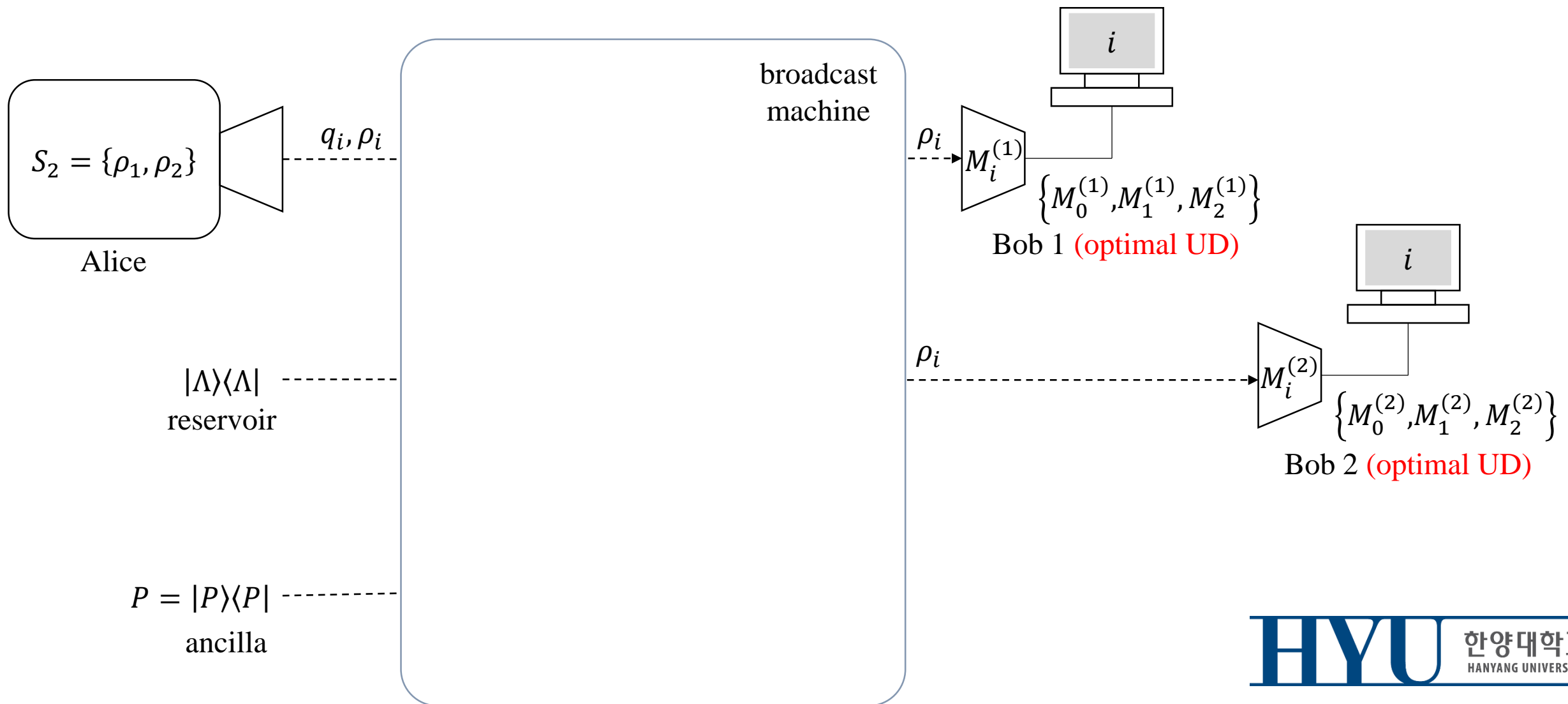


Quantum reproducing



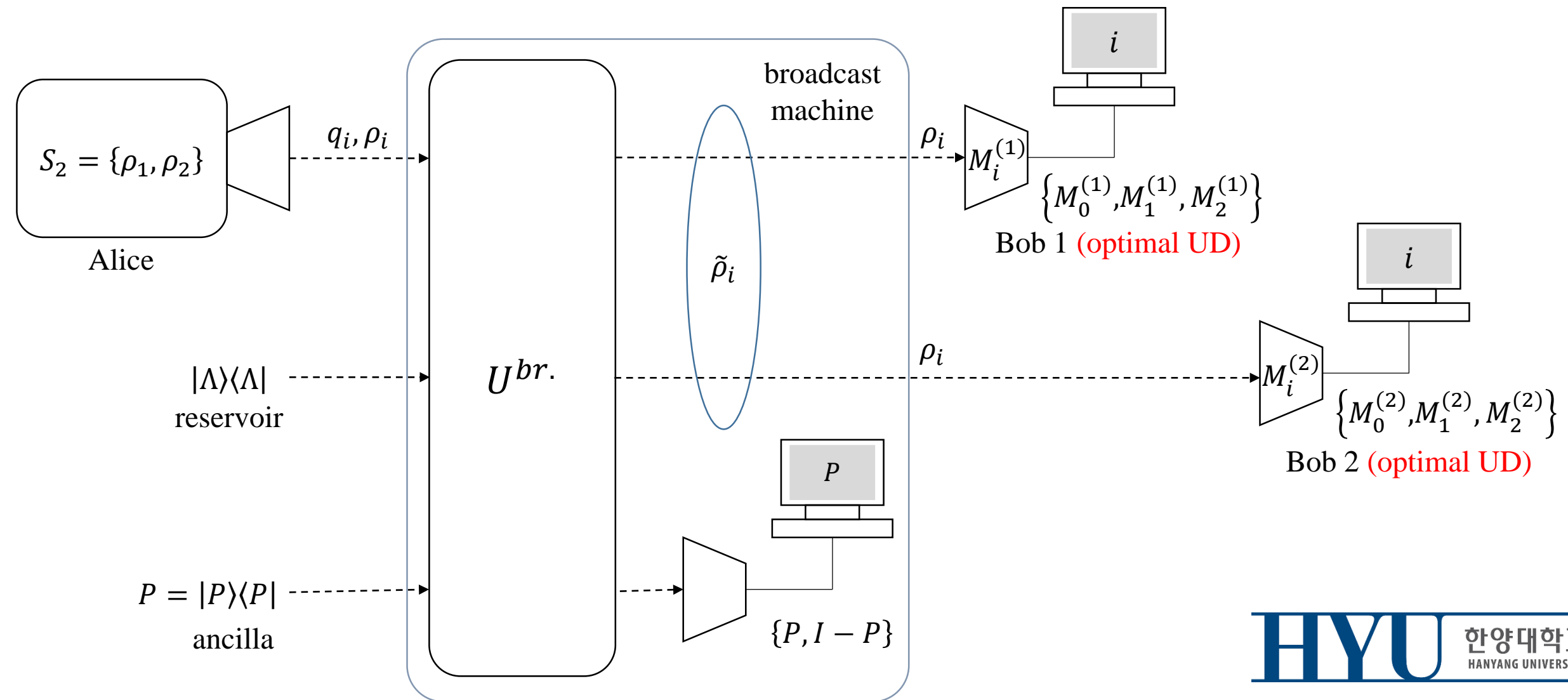
Probabilistic Quantum Broadcasting

L. Li et al., Probabilistic broadcasting of mixed states, J. Phys. A: Math. Theor. 42, 175302 (2009).



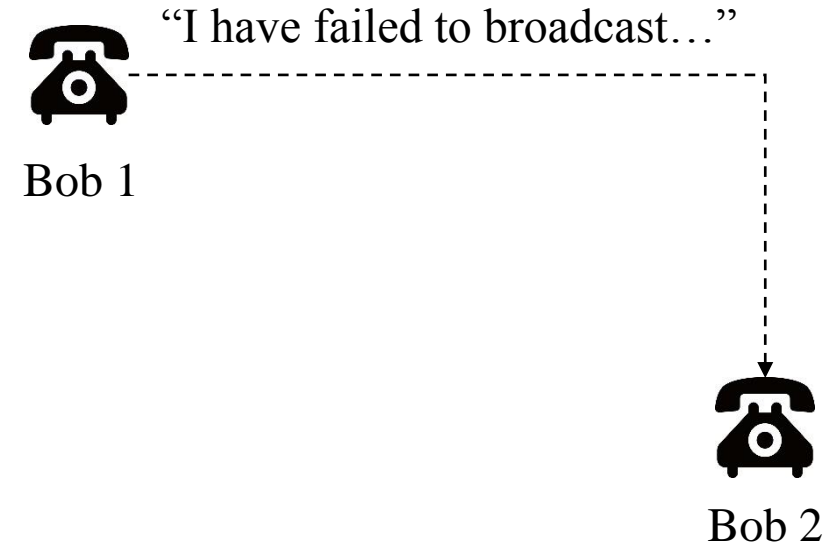
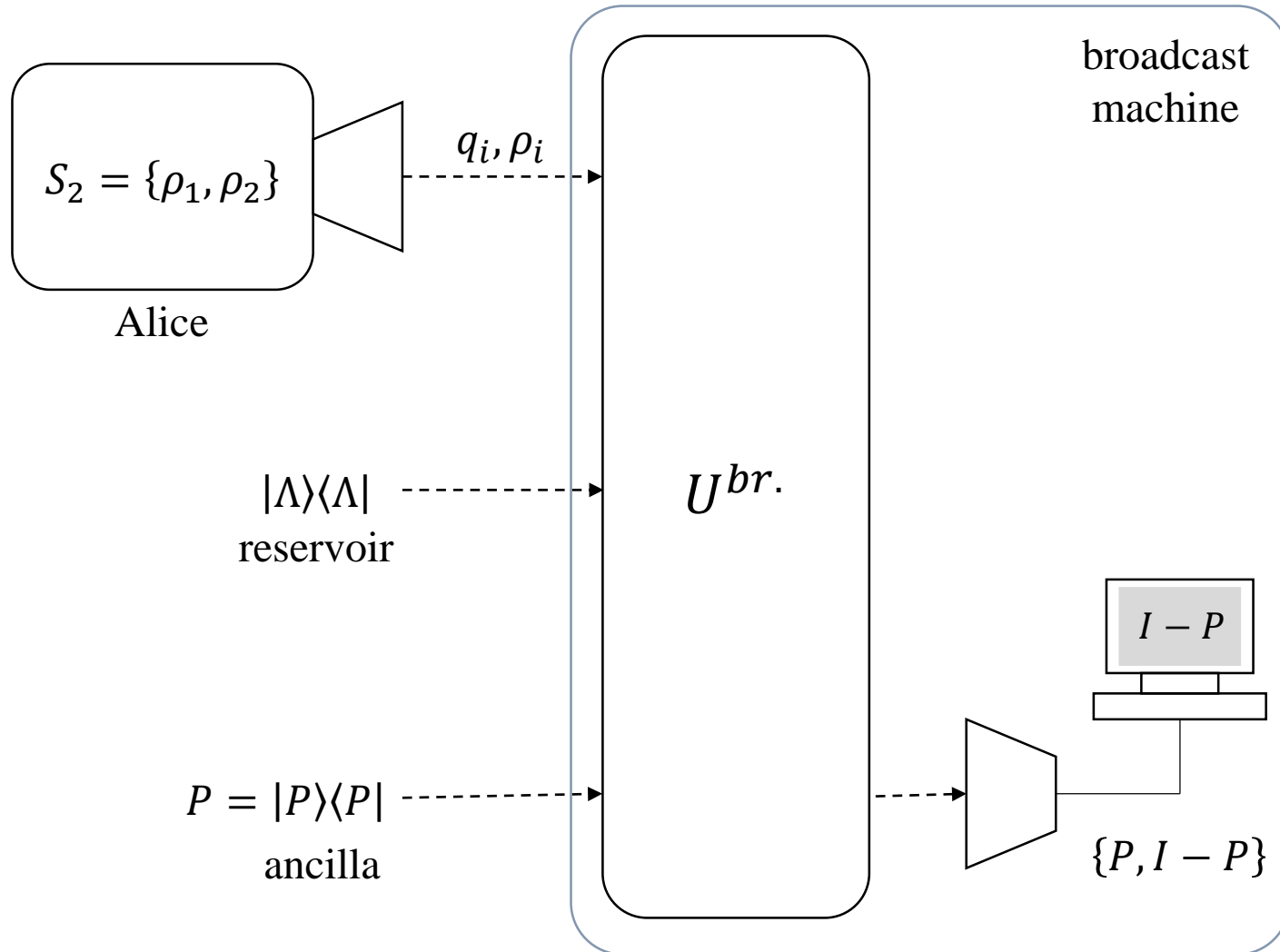
Probabilistic Quantum Broadcasting

L. Li et al., Probabilistic broadcasting of mixed states, J. Phys. A: Math. Theor. 42, 175302 (2009).

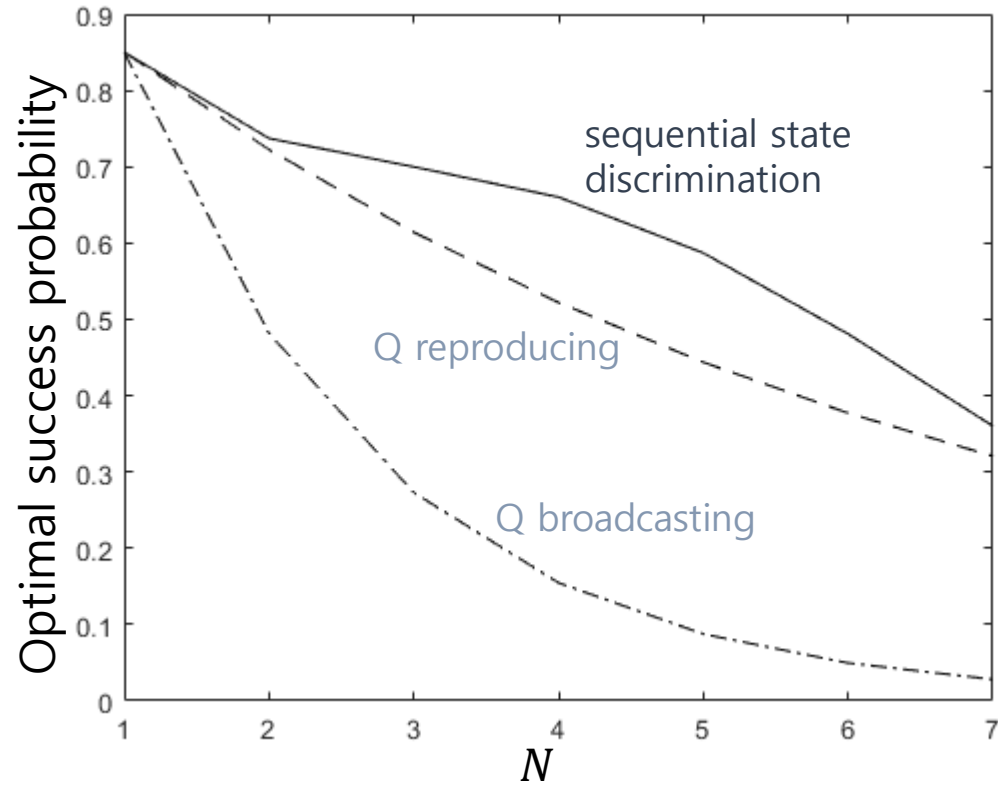


Probabilistic Quantum Broadcasting

L. Li et al., Probabilistic broadcasting of mixed states, J. Phys. A: Math. Theor. 42, 175302 (2009).

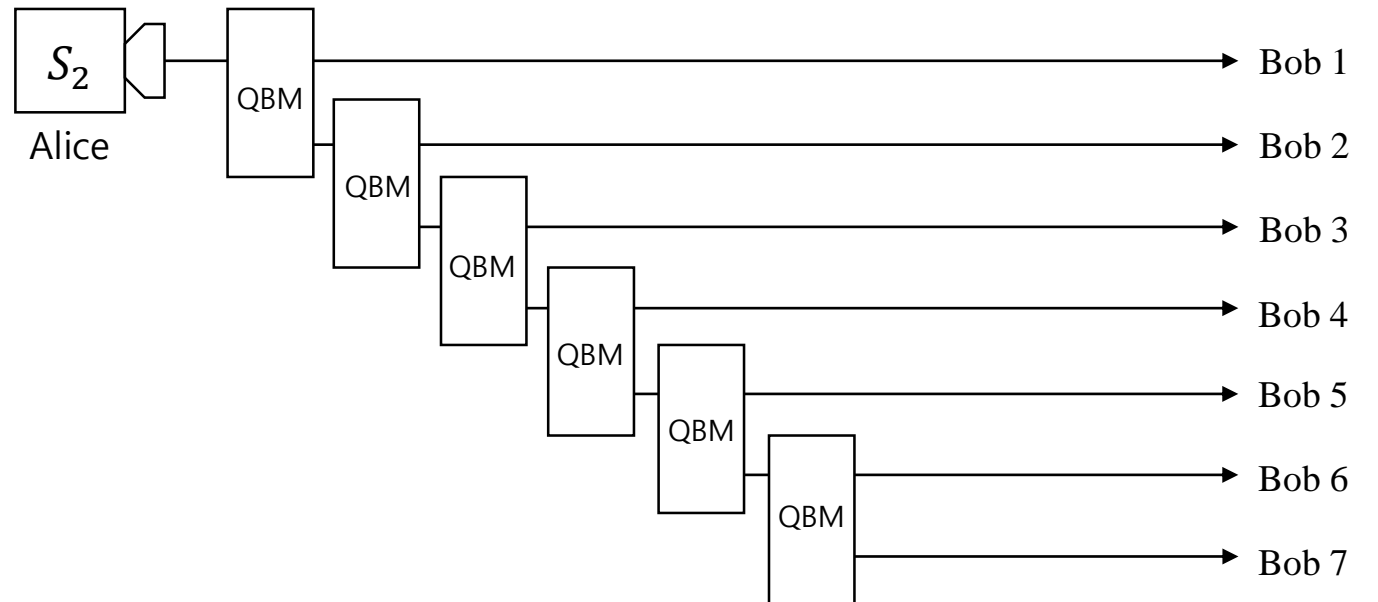


Optimal success probabilities



Example. $r = 0.3$, $\bar{r} = 0.7$, $s = 0.5$, $\bar{s} = 5 \times 10^{-8}$, $N \in \{1, \dots, 7\}$

(probabilistic quantum broadcasting)

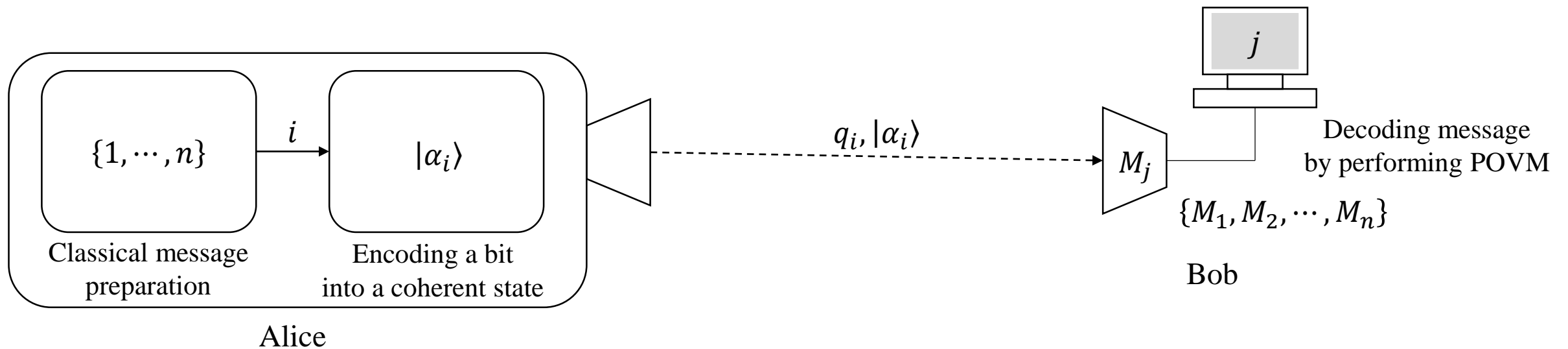


Remark : Sequential state discrimination outperform quantum reproducing and quantum broadcasting strategies.

Part II

Application

Realistic QKD based on coherent states



Realistic quantum key distribution

Realistic quantum key distribution can be expressed as *quantum state discrimination of coherent states*. For example, B92 protocol can be expressed as *unambiguous discrimination of coherent states*.

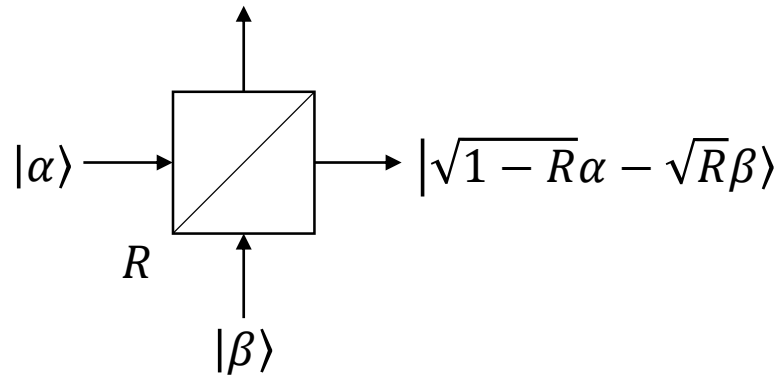
Introduction: coherent state and operations

Definition. [R. J. Glauber]

$$|\alpha\rangle = e^{-|\alpha|^2/2} \sum_{n=0}^{\infty} \frac{\alpha^n}{\sqrt{n!}} |n\rangle, \quad n : \text{number of photons}$$

Lemma 1. $\hat{D}(\gamma)|\alpha\rangle = e^{\gamma\hat{a}-\gamma^*\hat{a}^\dagger}|\alpha\rangle \simeq |\alpha + \gamma\rangle$

Lemma 2. $|\sqrt{R}\alpha + \sqrt{1-R}\beta\rangle$



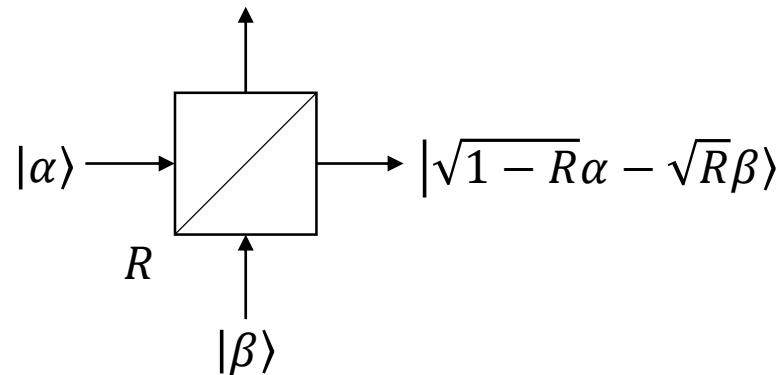
Introduction: coherent state and operations

Definition. [R. J. Glauber]

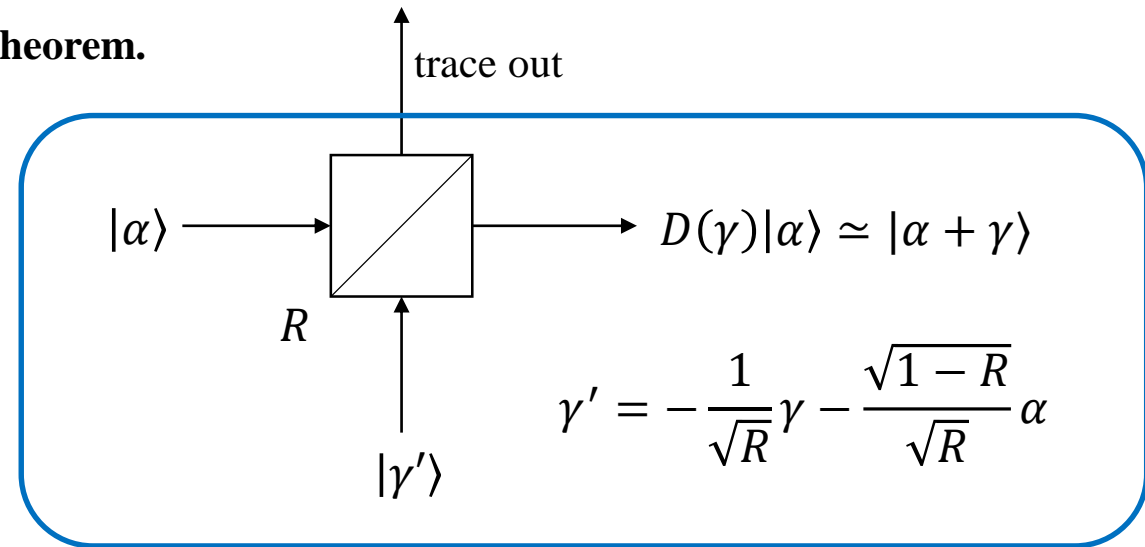
$$|\alpha\rangle = e^{-|\alpha|^2/2} \sum_{n=0}^{\infty} \frac{\alpha^n}{\sqrt{n!}} |n\rangle, \quad n : \text{number of photons}$$

Lemma 1. $\hat{D}(\gamma)|\alpha\rangle = e^{\gamma\hat{a}-\gamma^*\hat{a}^\dagger} |\alpha\rangle \simeq |\alpha + \gamma\rangle$

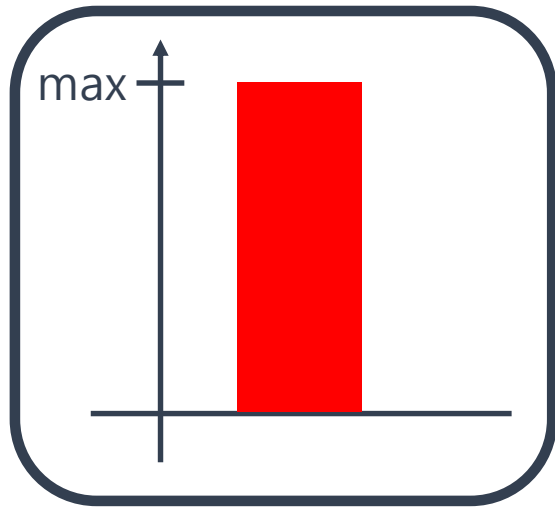
Lemma 2. $|\sqrt{R}\alpha + \sqrt{1-R}\beta\rangle$



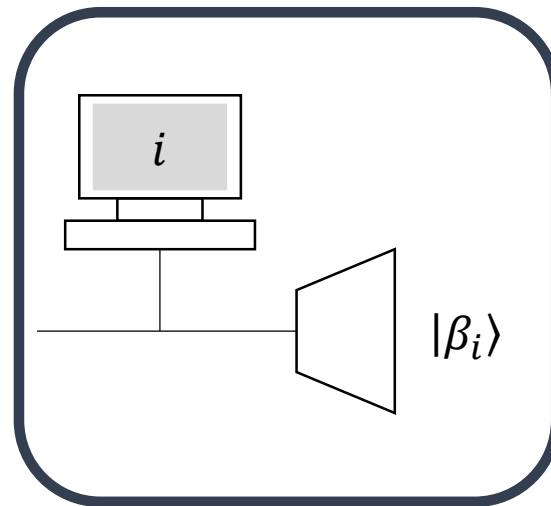
Theorem.



Requirement for implementing sequential state discrimination

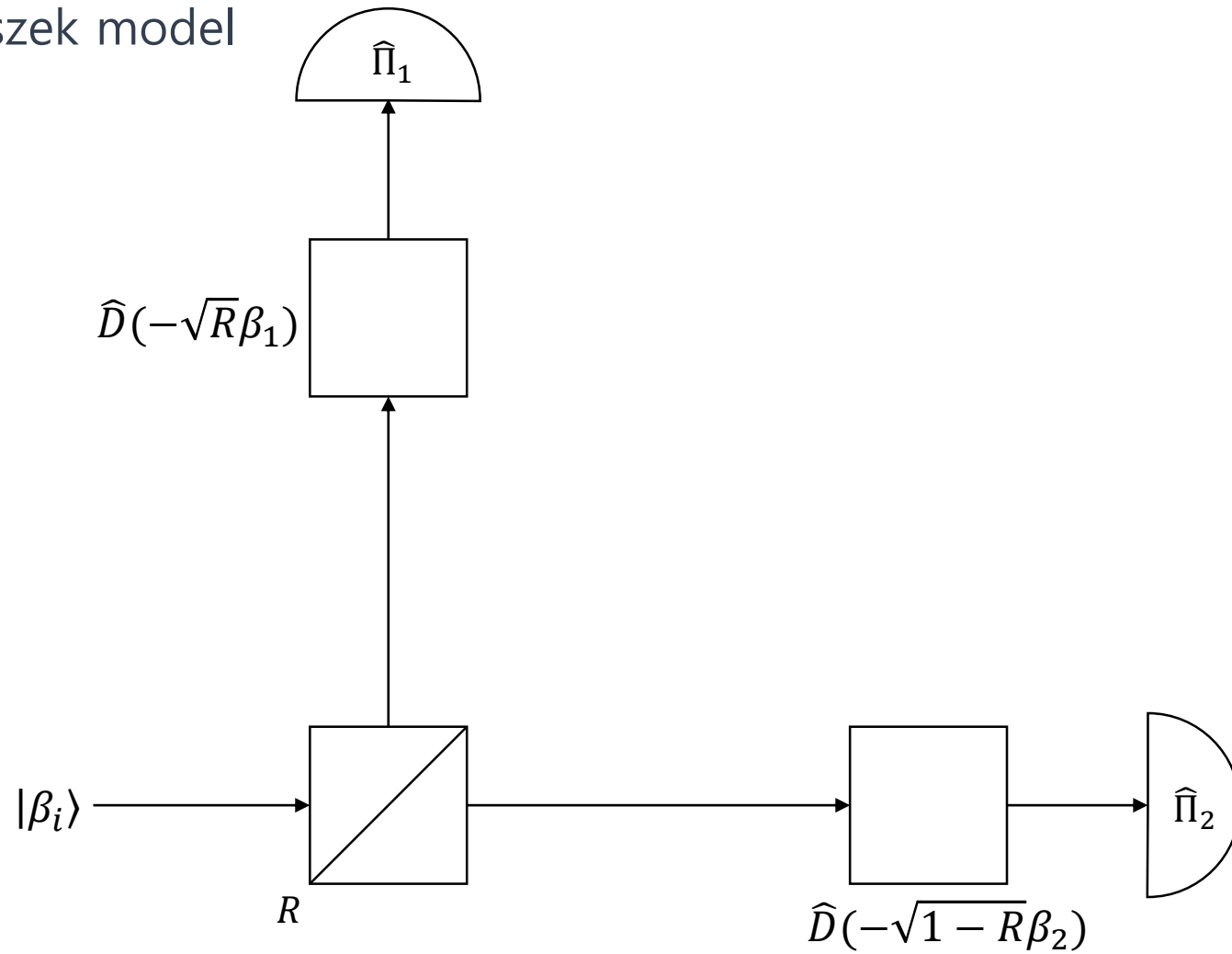


- Optical design should implement optimal unambiguous discrimination

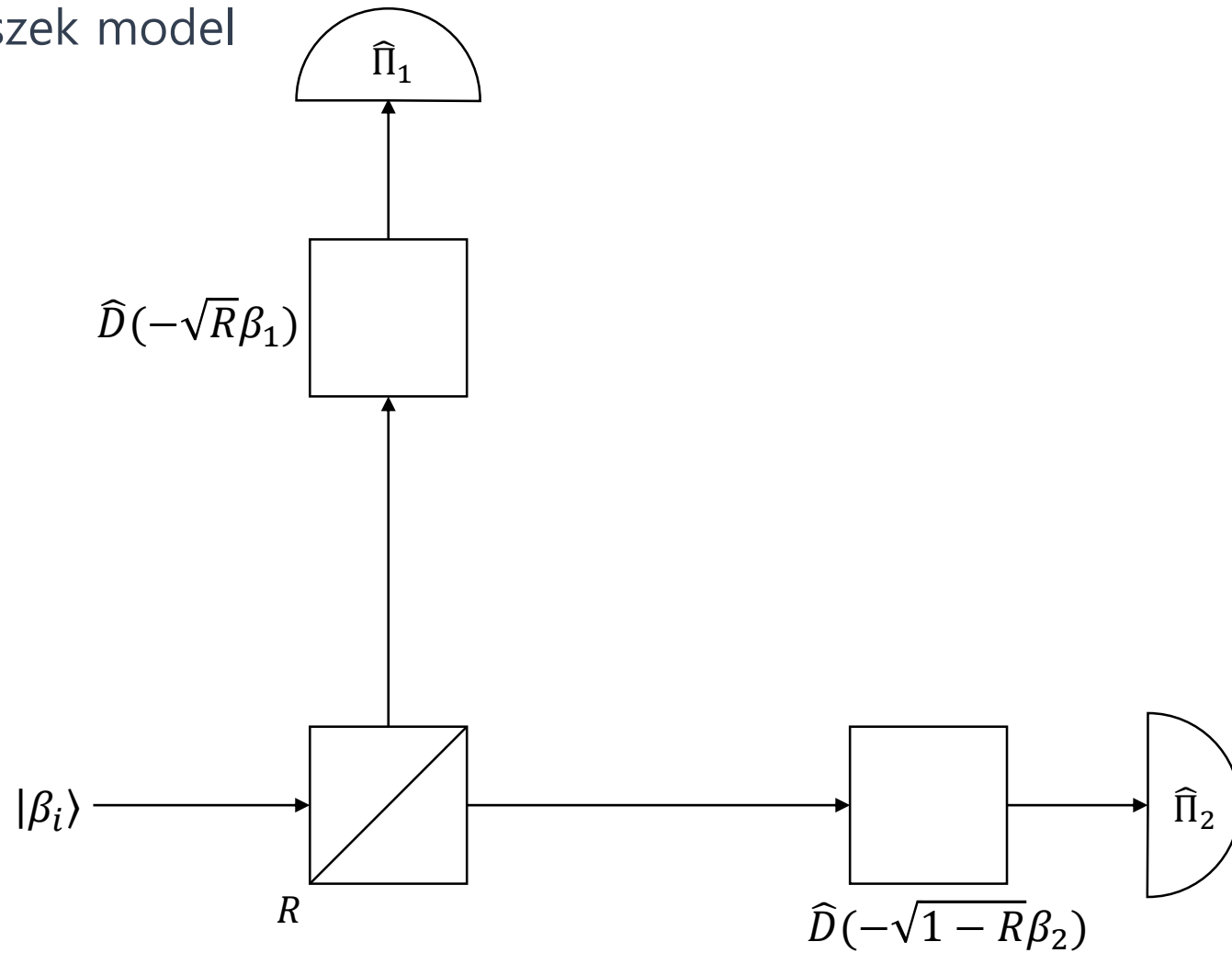


- Optical design should produce non-orthogonal post-measurement states

Banaszek model



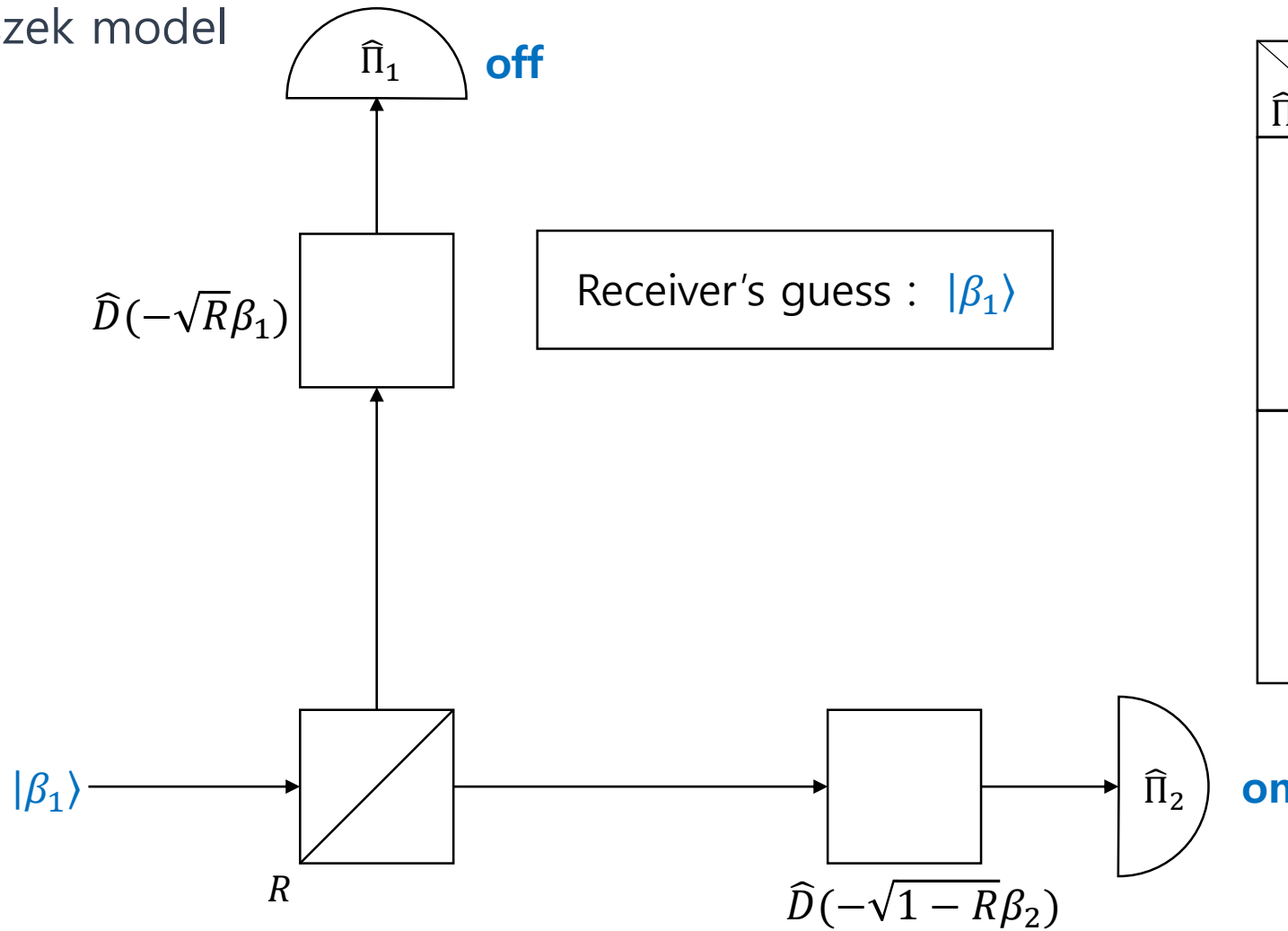
Banaszek model



$\hat{\Pi}_1 \backslash \hat{\Pi}_2$	off	on
off	inconclusive	correct
on	error (forbidden)	(forbidden)

Implementing sequential state discrimination

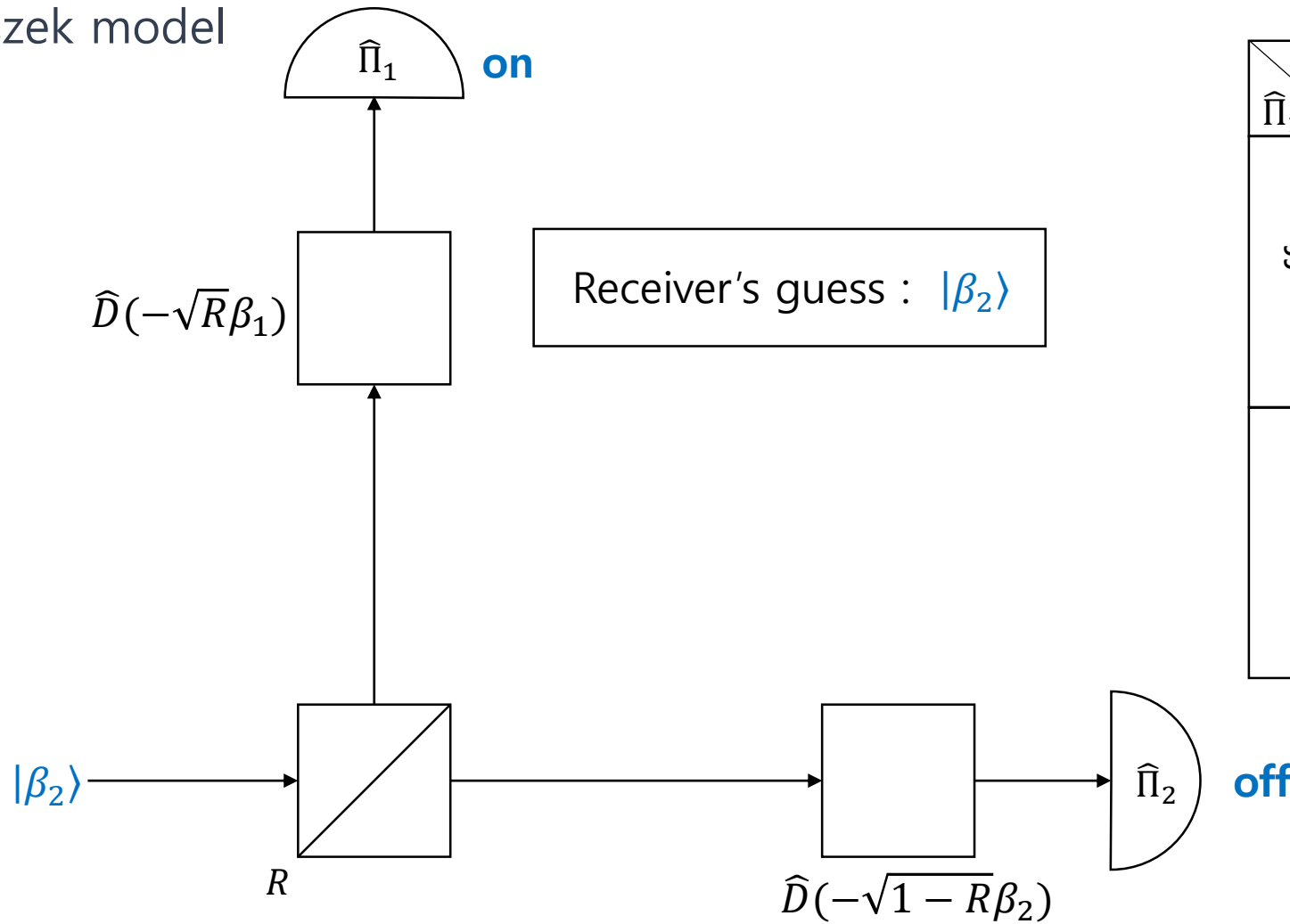
Banaszek model



$\hat{\Pi}_1 \backslash \hat{\Pi}_2$	off	on
off	inconclusive	correct
on	error (forbidden)	(forbidden)

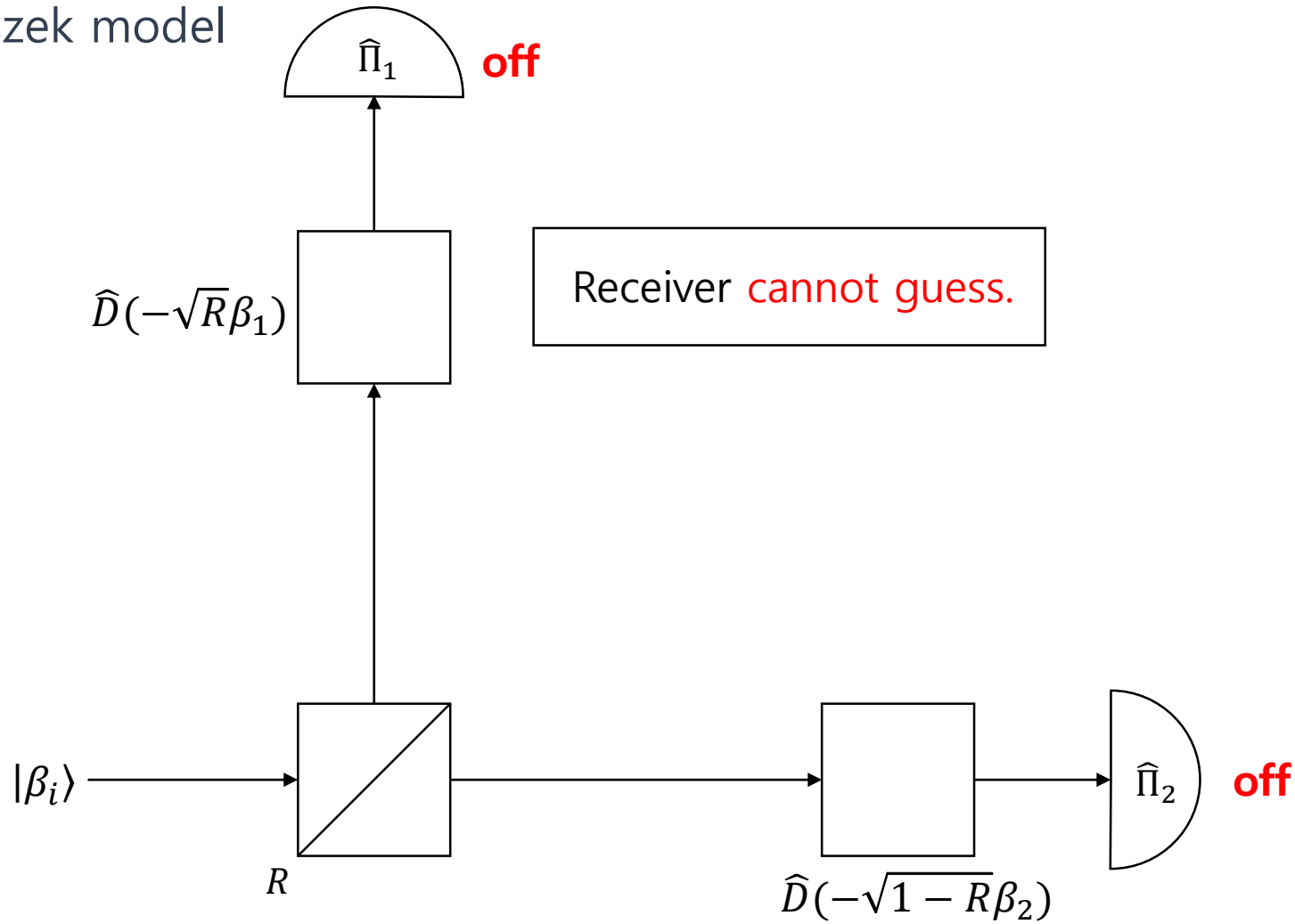
Implementing sequential state discrimination

Banaszek model



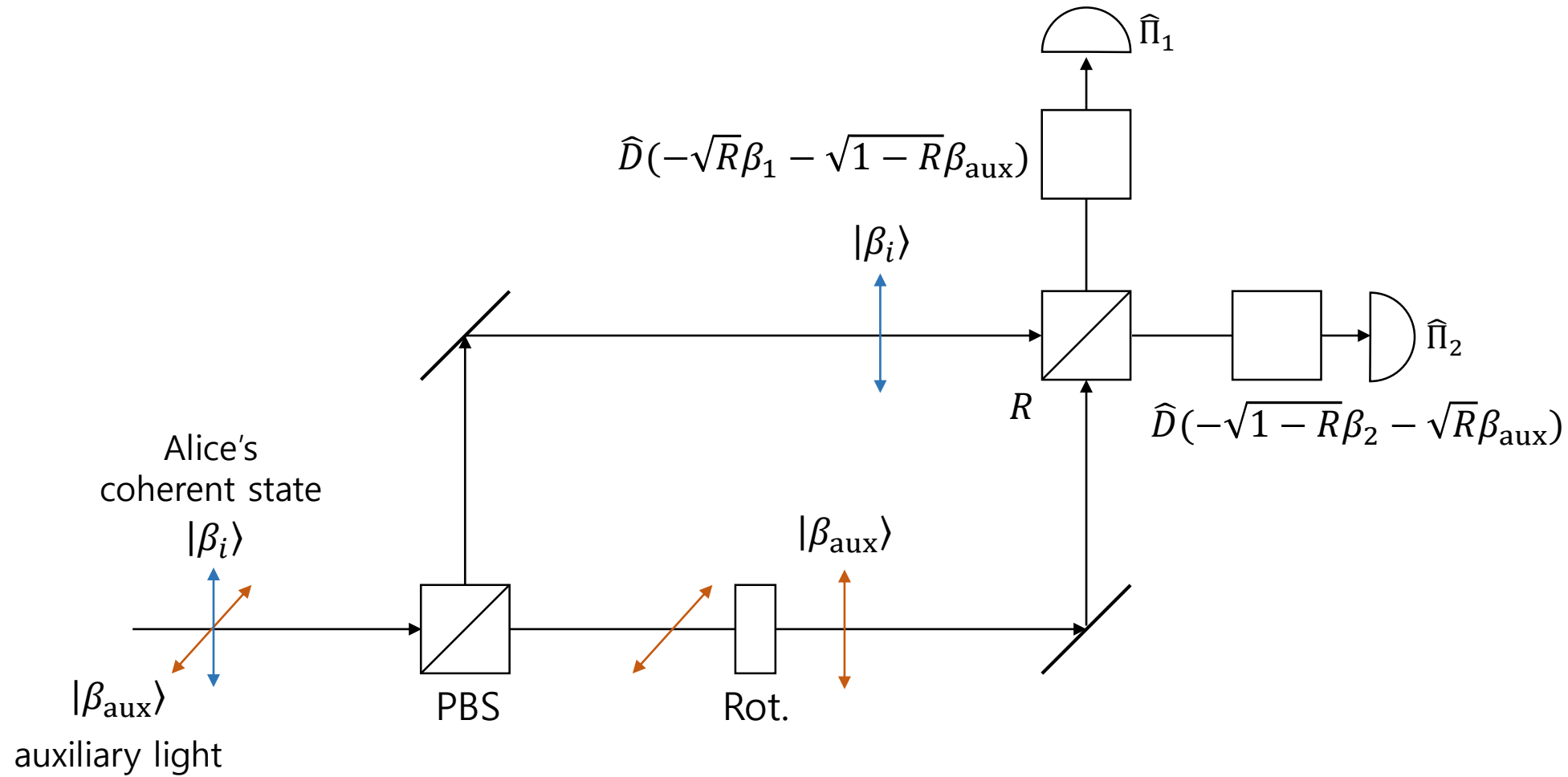
$\hat{\Pi}_1 \backslash \hat{\Pi}_2$	off	on
off	inconclusive	error (forbidden)
on	correct	(forbidden)

Banaszek model



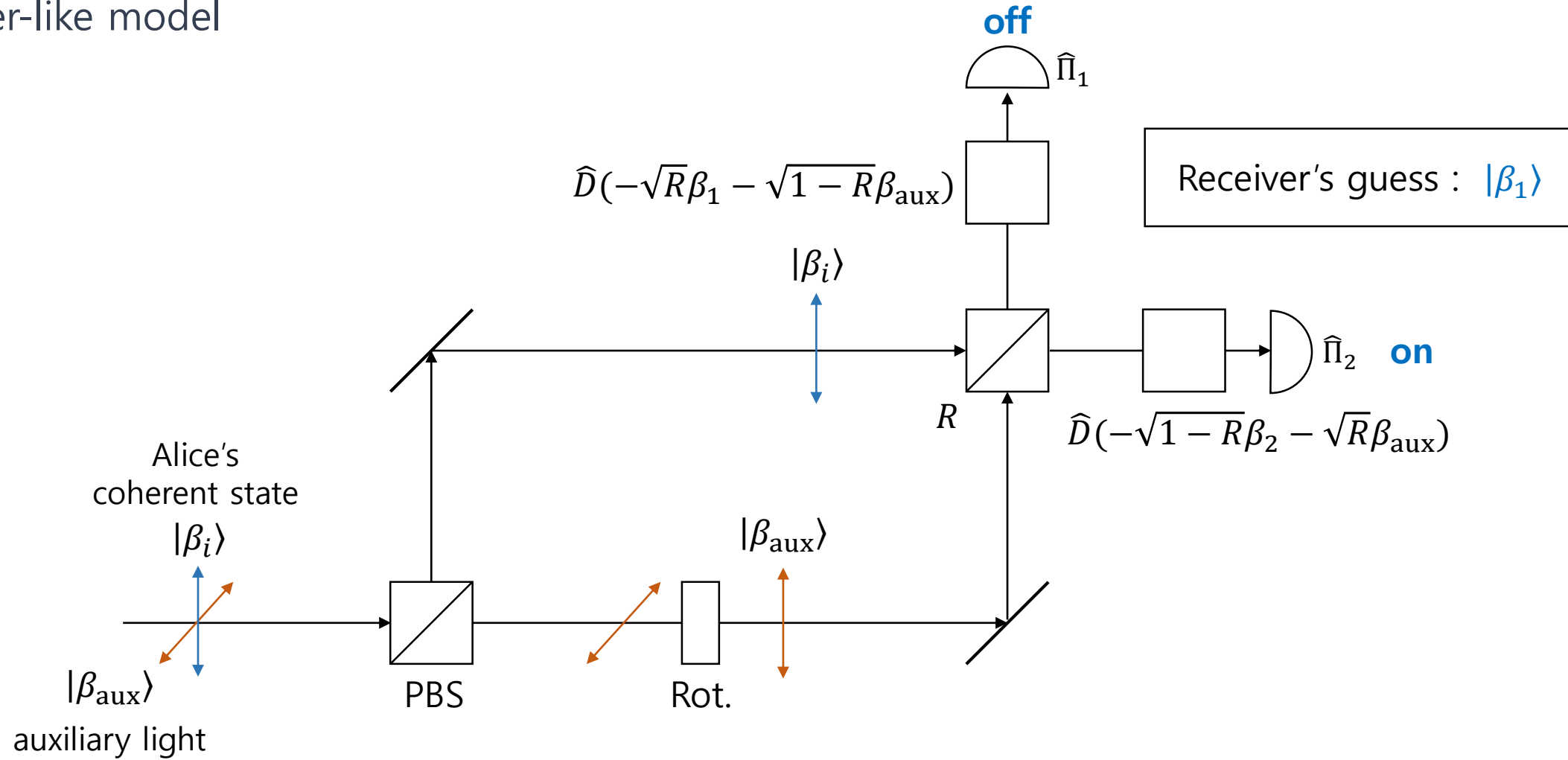
Implementing sequential state discrimination

Huttner-like model



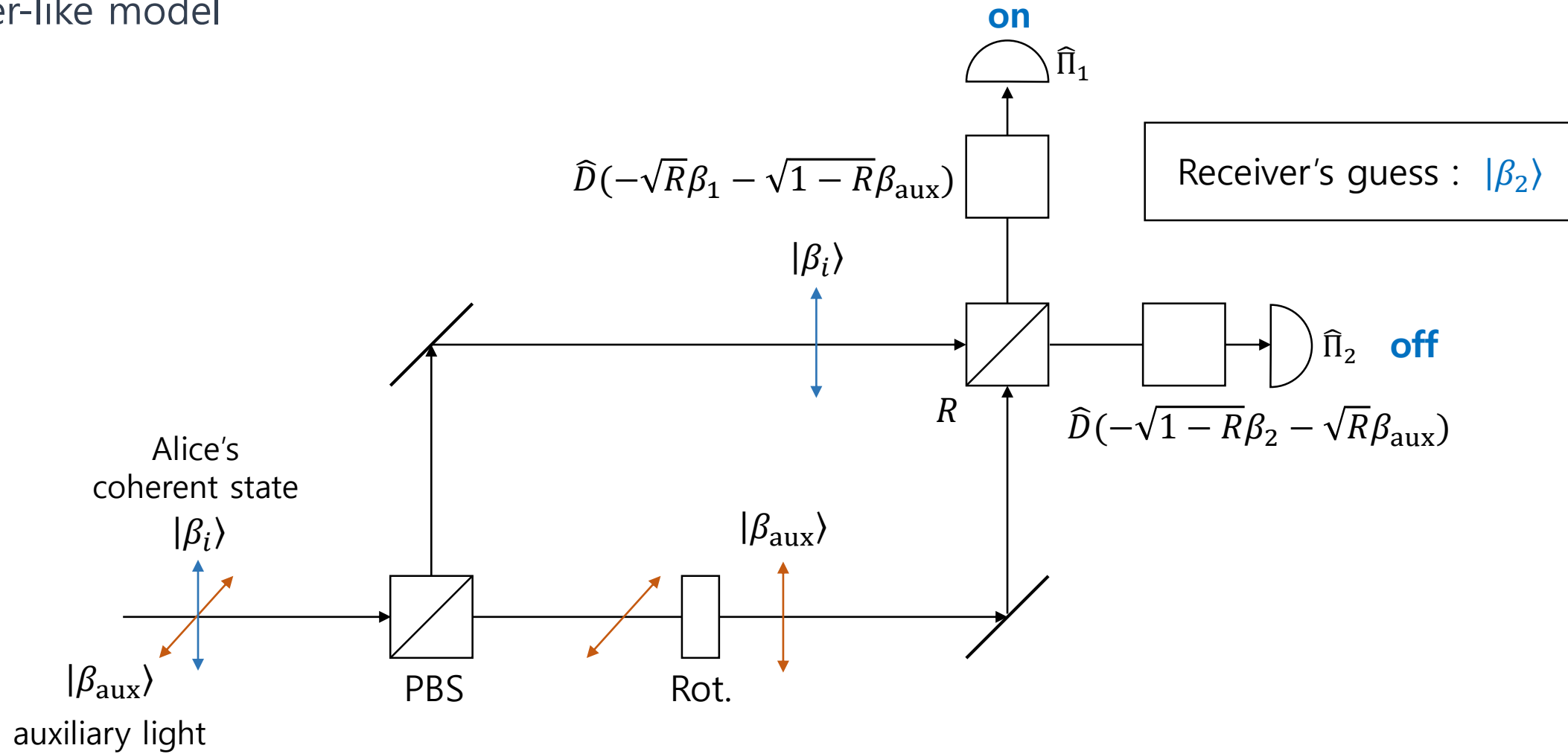
Implementing sequential state discrimination

Huttner-like model



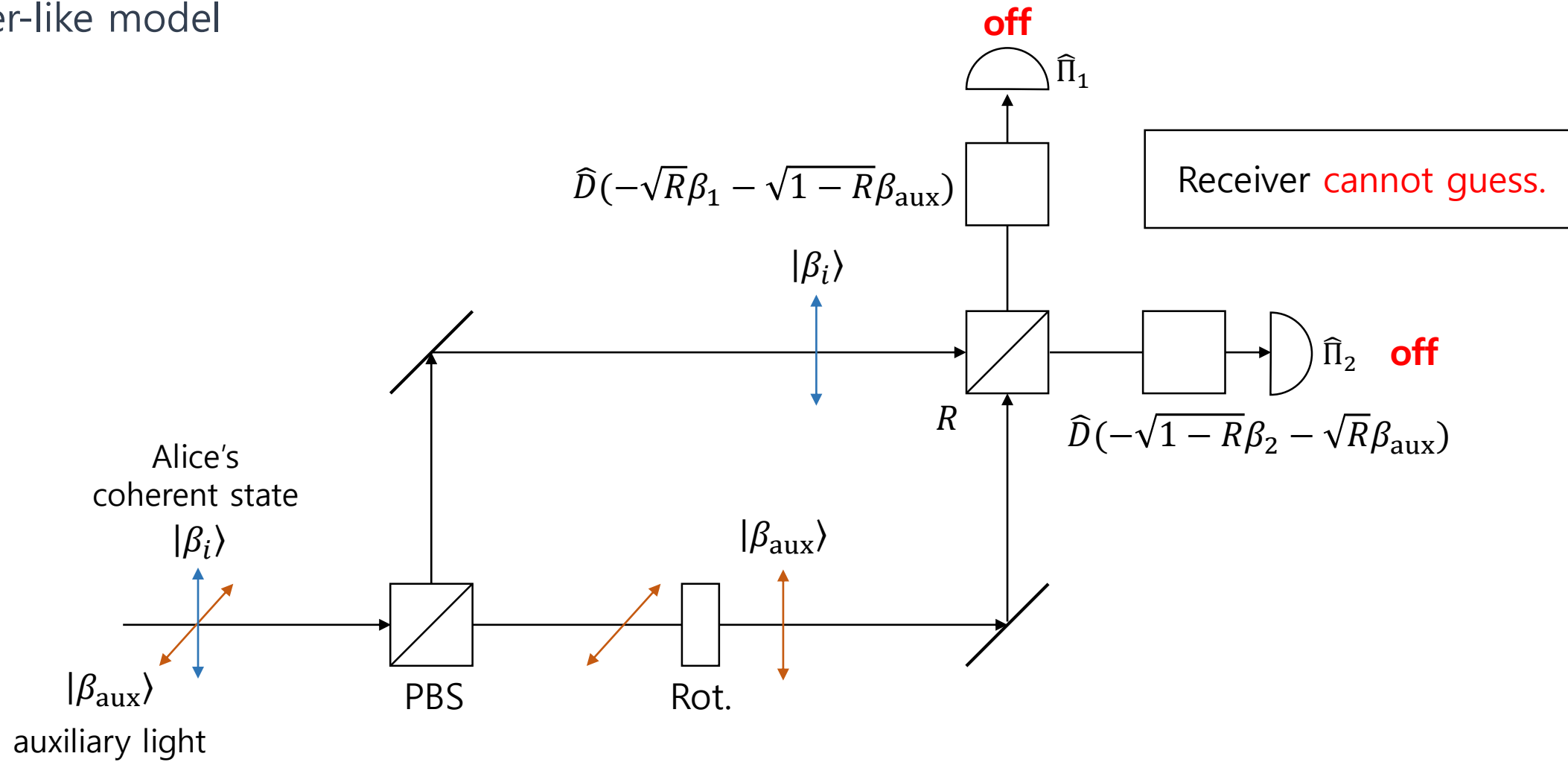
Implementing sequential state discrimination

Huttner-like model



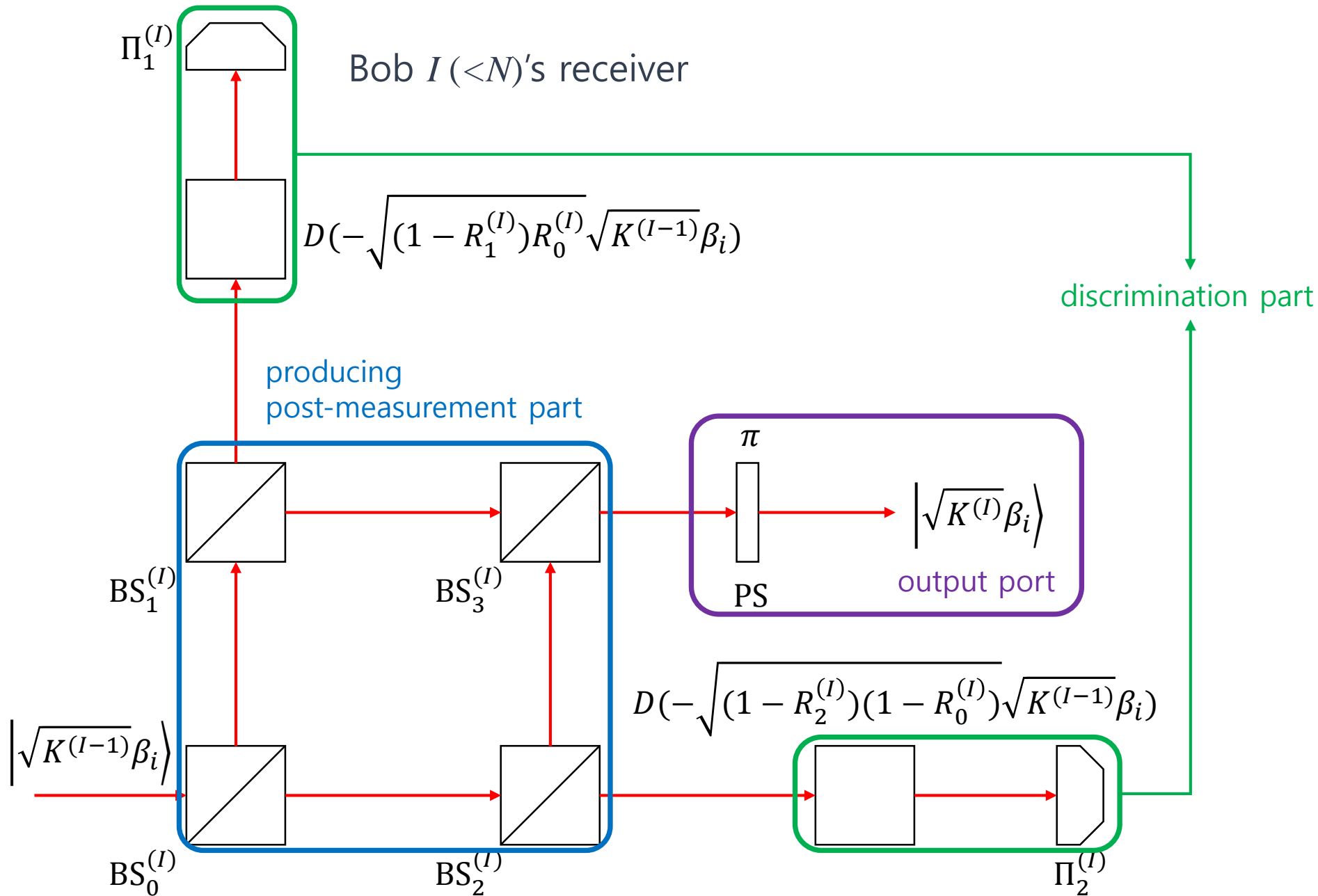
Implementing sequential state discrimination

Huttner-like model



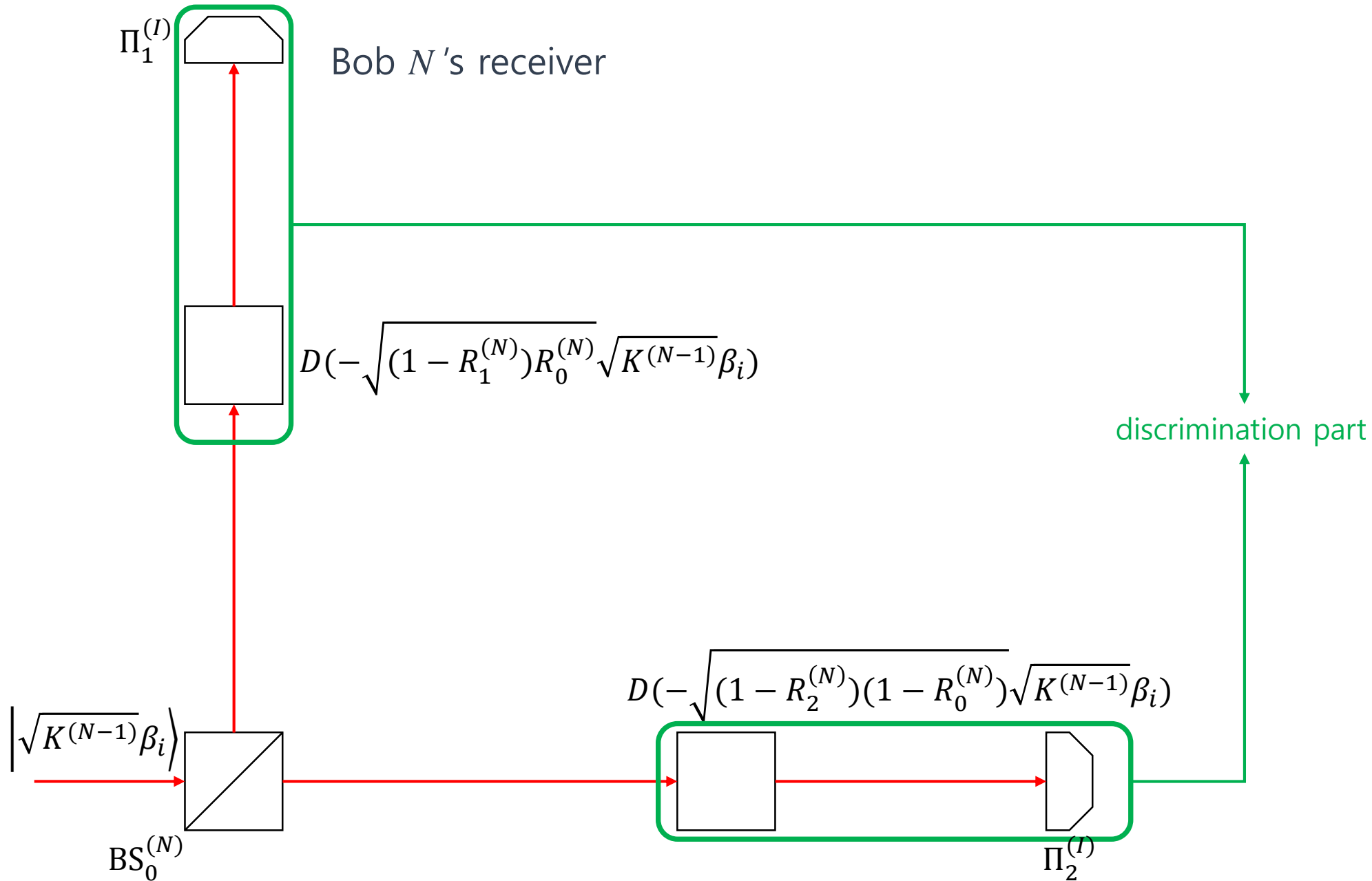
Implementing sequential state discrimination

M. Namkung and Y. Kwon,
Sequential state discrimination
of coherent states,
Scientific Reports **8**, 16915 (2018).



Implementing sequential state discrimination

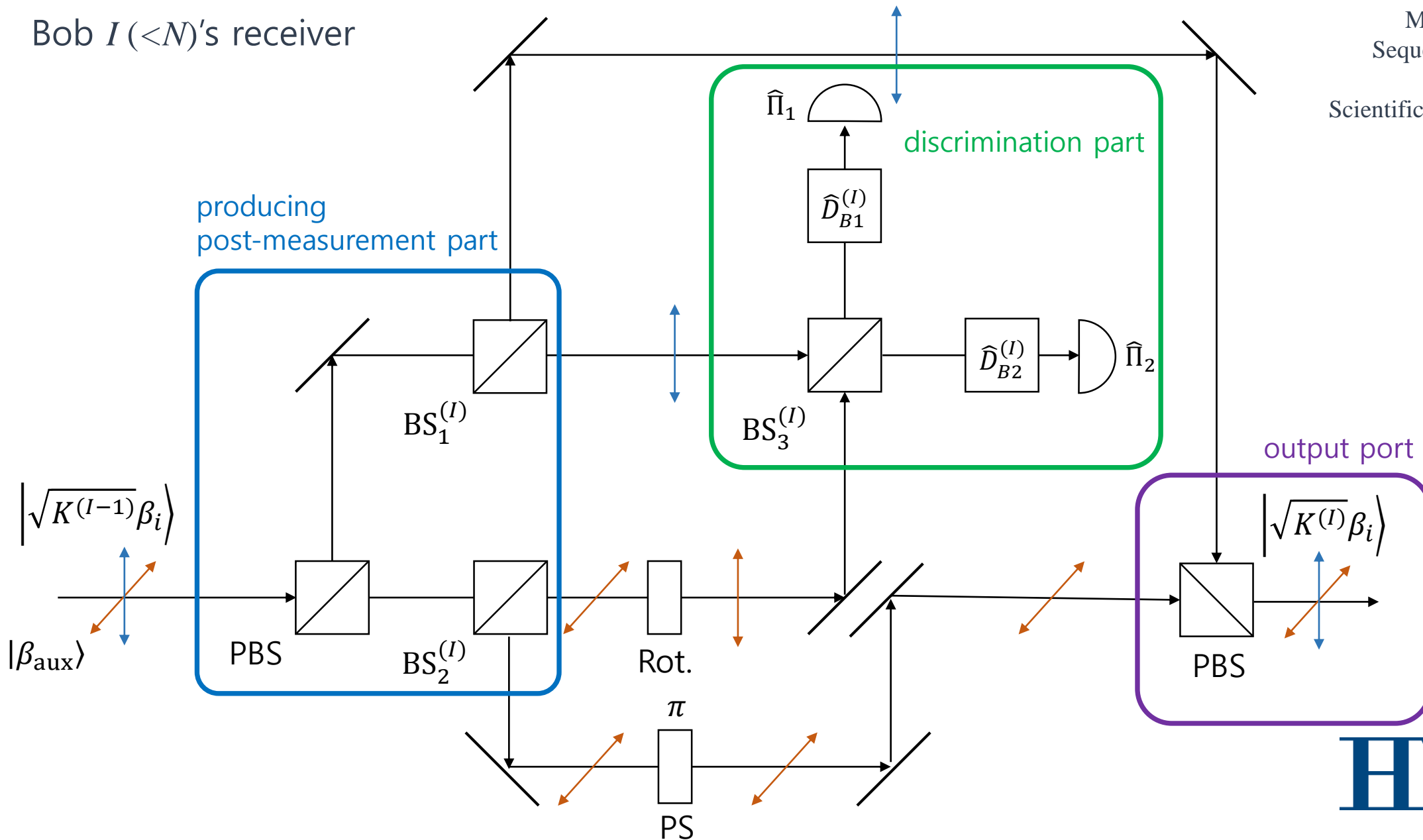
M. Namkung and Y. Kwon,
Sequential state discrimination
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Scientific Reports **8**, 16915 (2018).



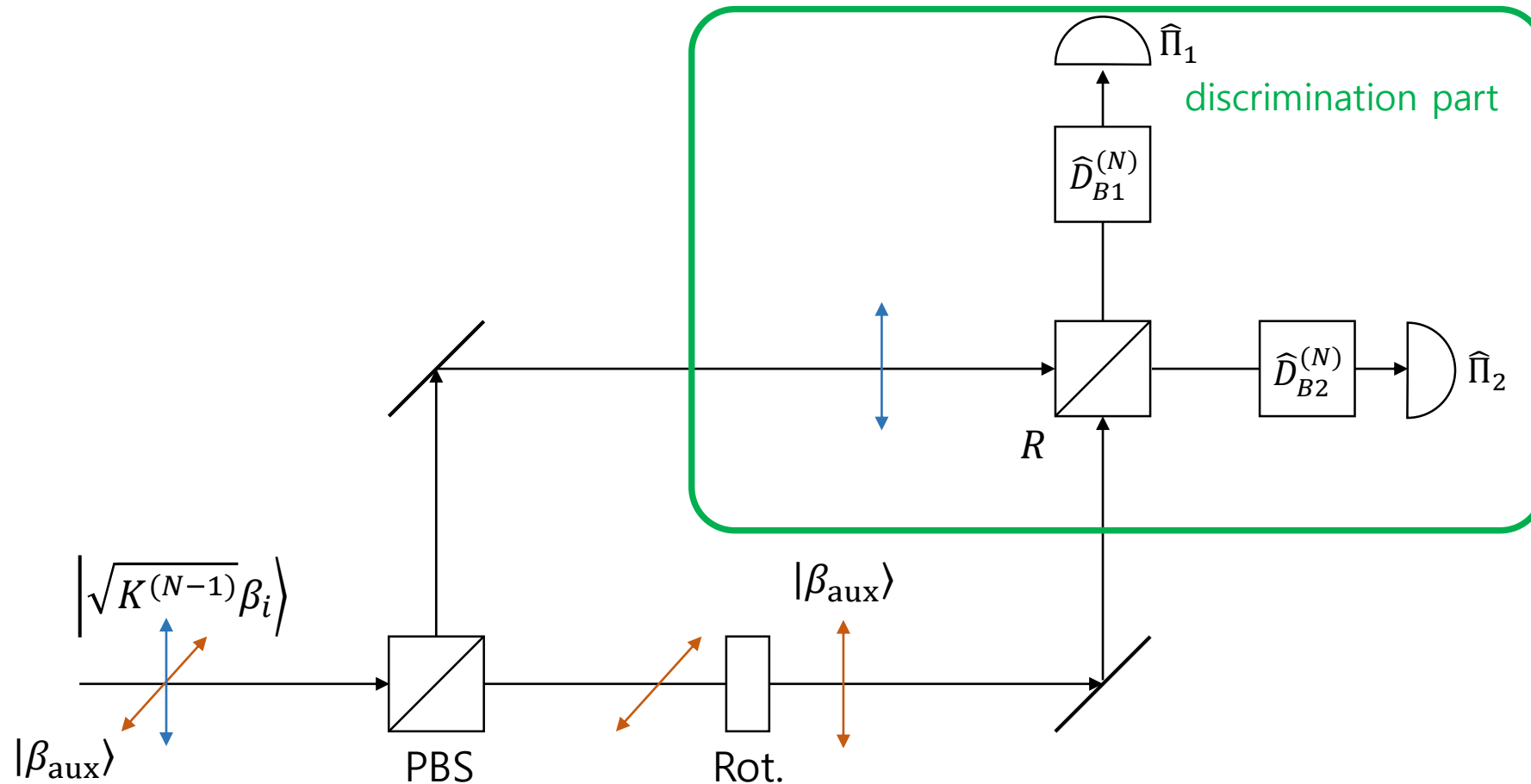
Implementing sequential state discrimination

Bob I ($<N$)'s receiver

M. Namkung and Y. Kwon,
Sequential state discrimination
of coherent states,
Scientific Reports **8**, 16915 (2018).



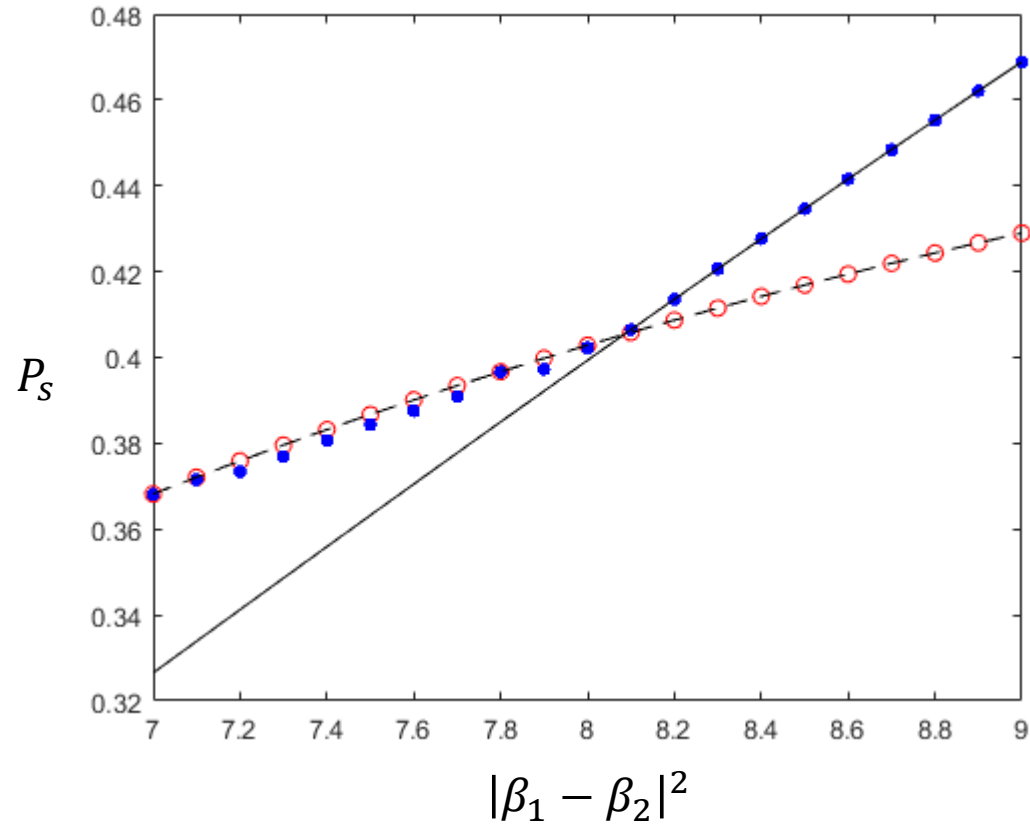
Bob N' 's receiver



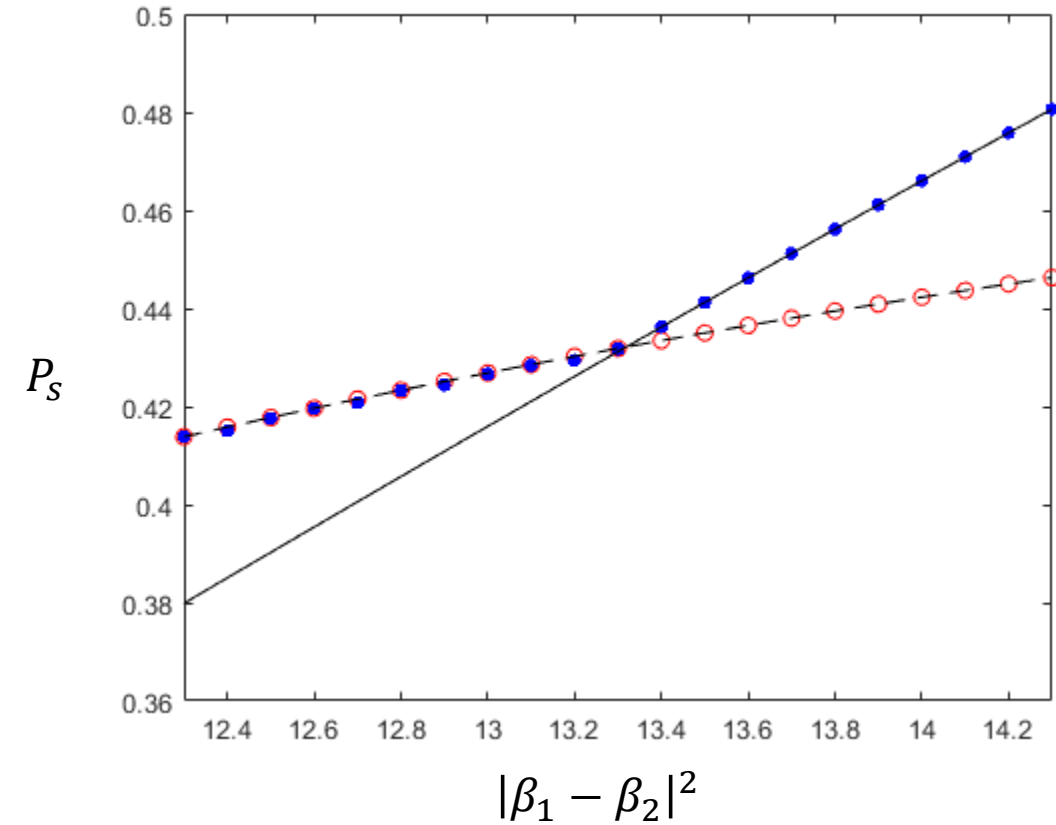
M. Namkung and Y. Kwon,
Sequential state discrimination
of coherent states,
Scientific Reports **8**, 16915 (2018).

Optimal success probability

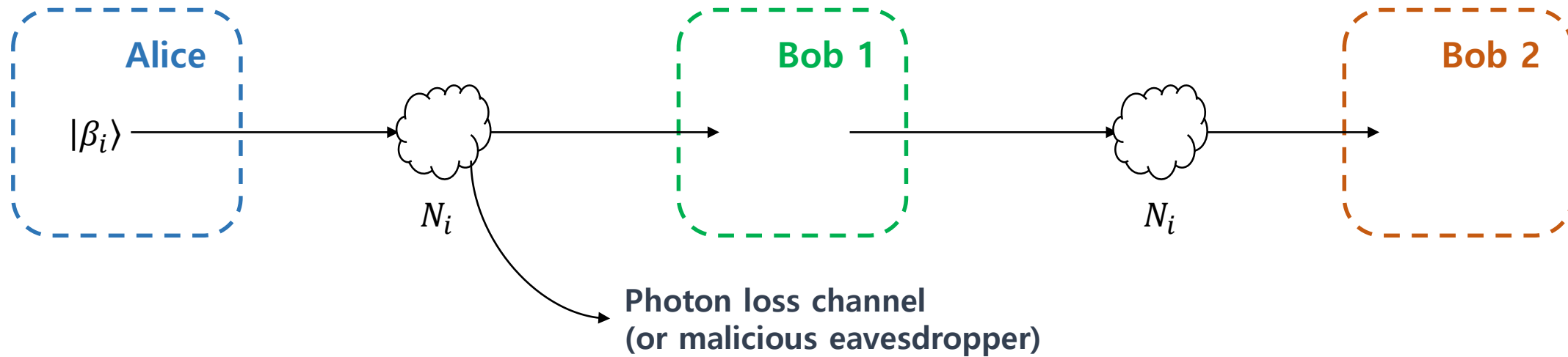
(three receivers)



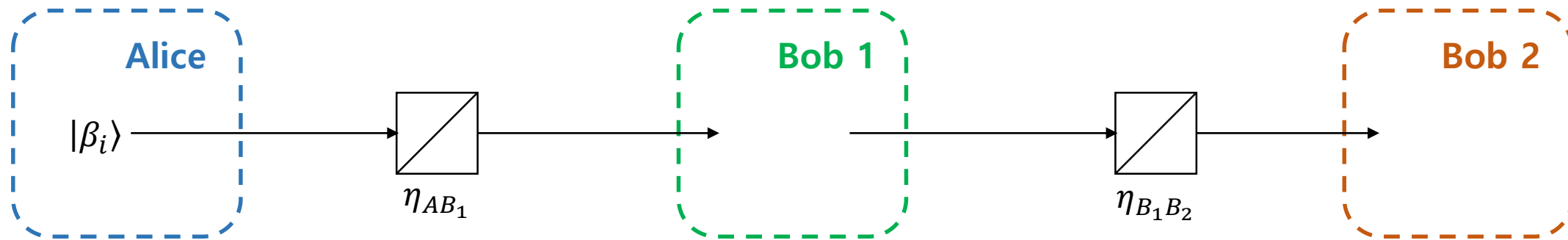
(four receivers)



Noisy channel in Banaszek model



Noisy channel in Banaszek model

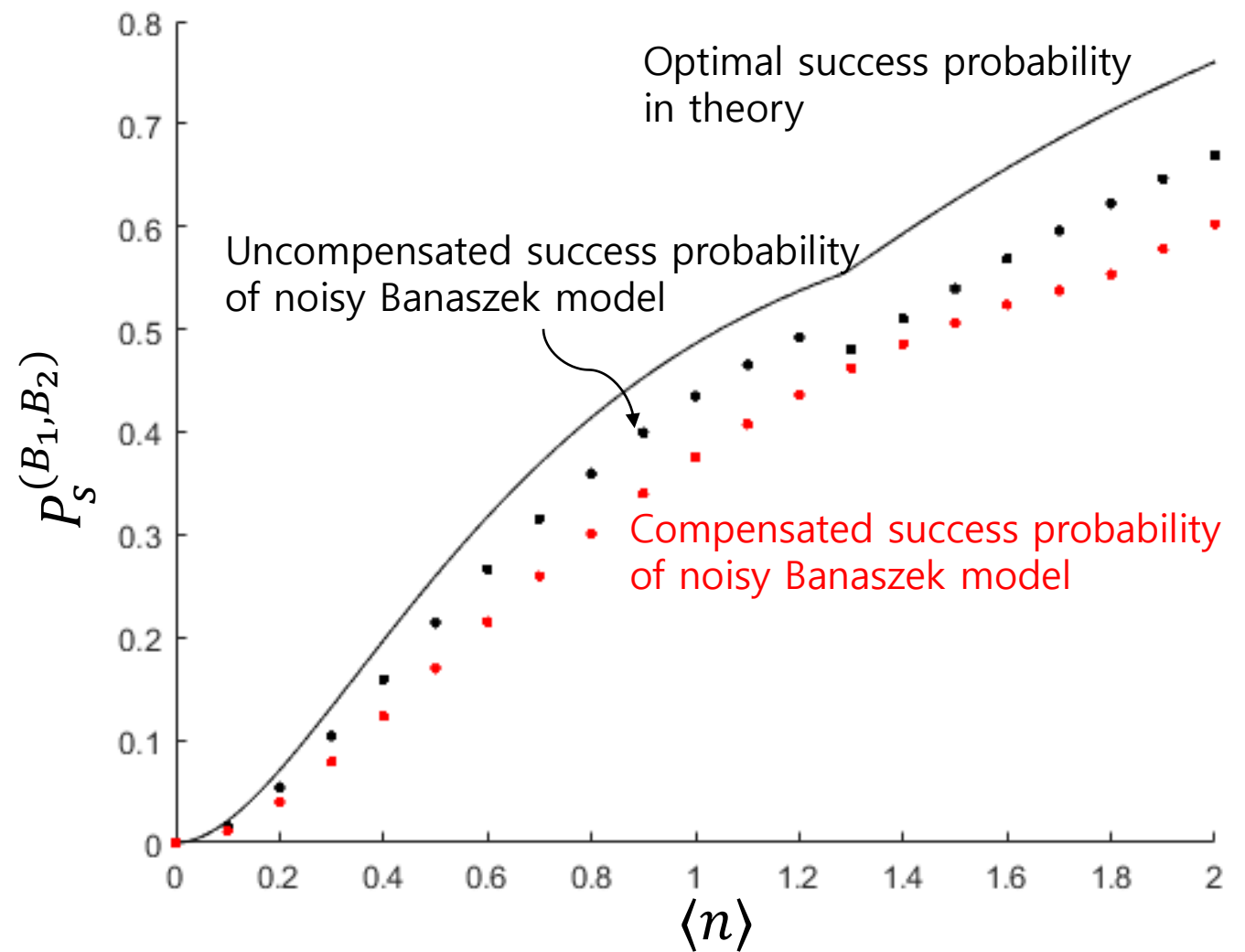


Noisy channel in Banaszek model

Example. $|\beta_1\rangle = |\alpha\rangle$, $|\beta_2\rangle = |-\alpha\rangle$

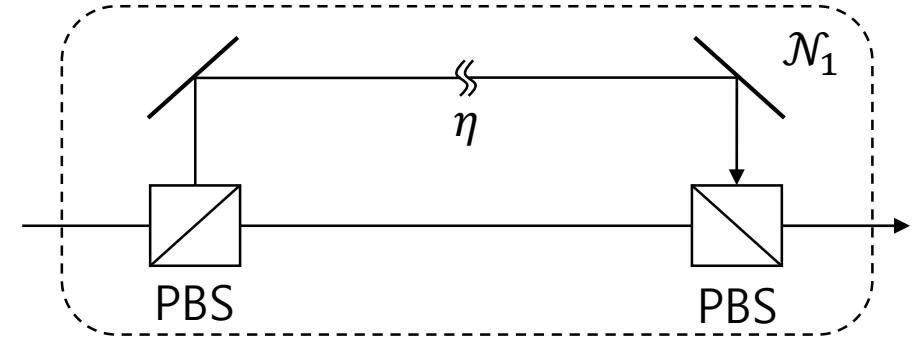
In view of noisy channel:
Photon loss noise can be compensated,
but success probability decreases.

In view of eavesdropping:
Eavesdropping decreases success probability.

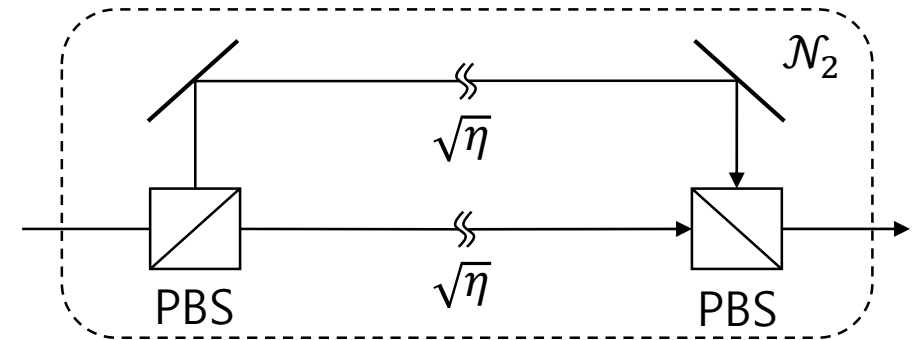


Noisy channel in Huttner-like model

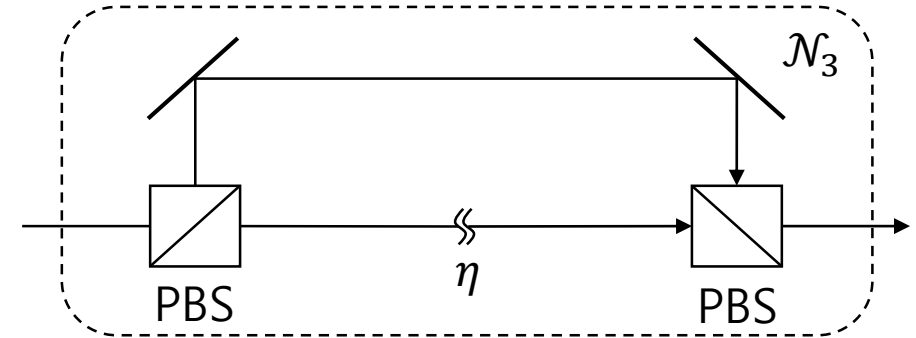
Photon loss(eavesdropping)
on Alice's coherent state only



Photon loss(eavesdropping)
on Alice's coherent state
and auxiliary light

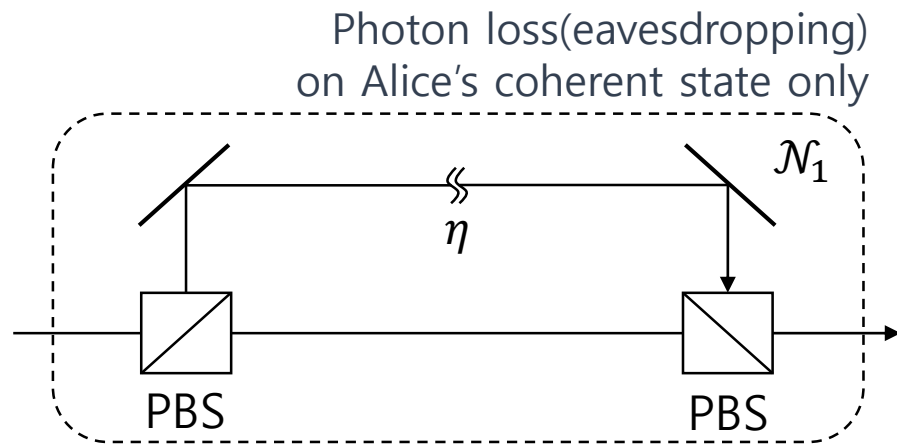


Photon loss(eavesdropping)
on auxiliary light only



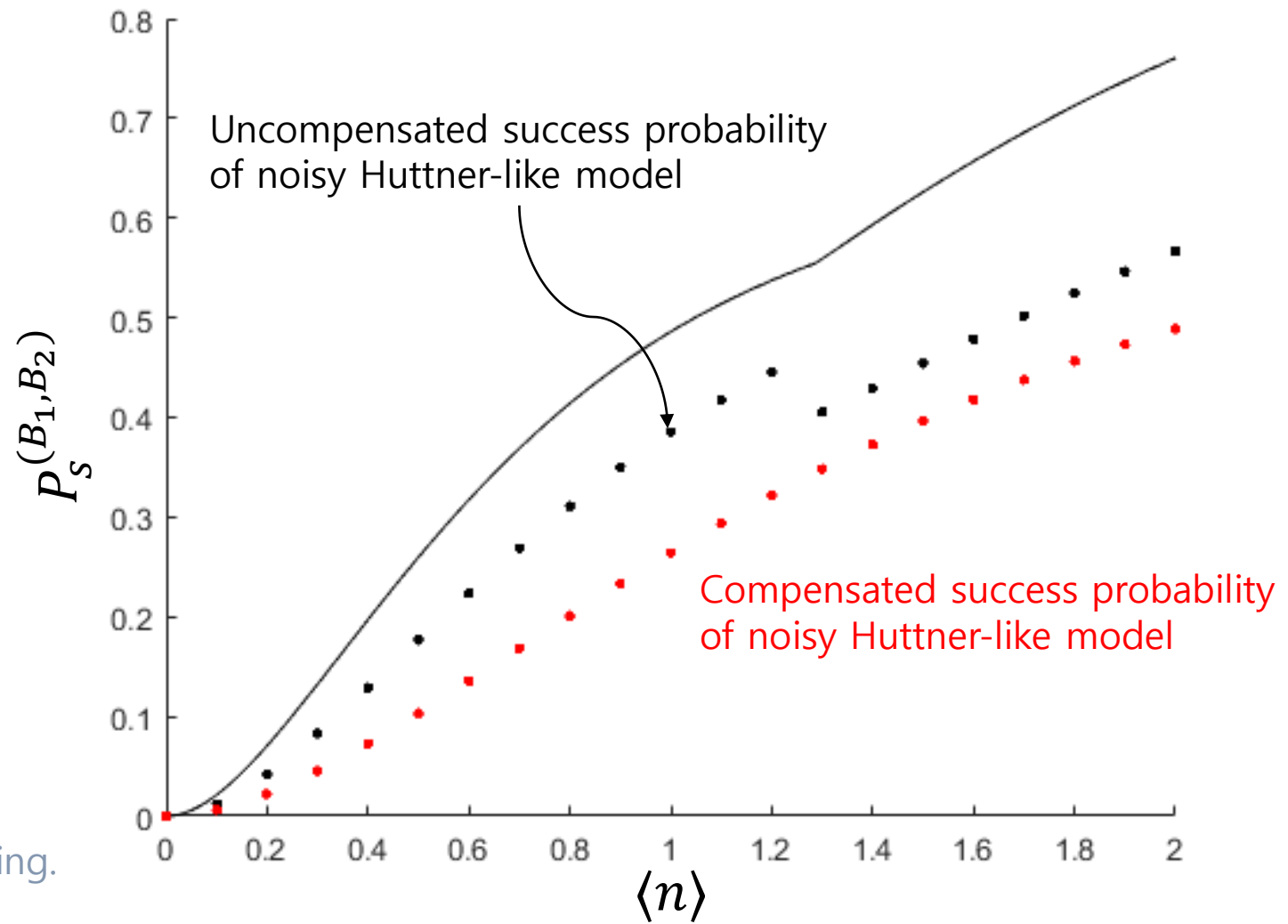
Sequential state discrimination in noisy channel

Noisy channel in Huttner-like model



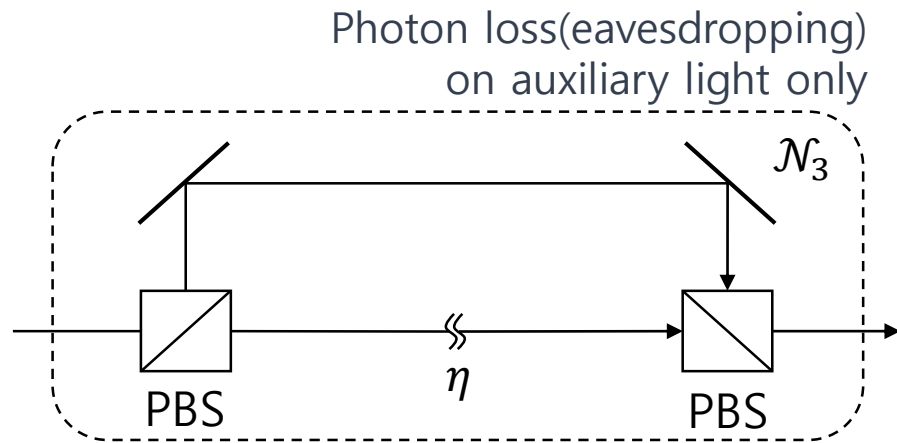
In view of noisy channel:
Compensation decreases large amount of success probability.

In view of eavesdropping:
Because of decreasing, receivers notice eavesdropping.



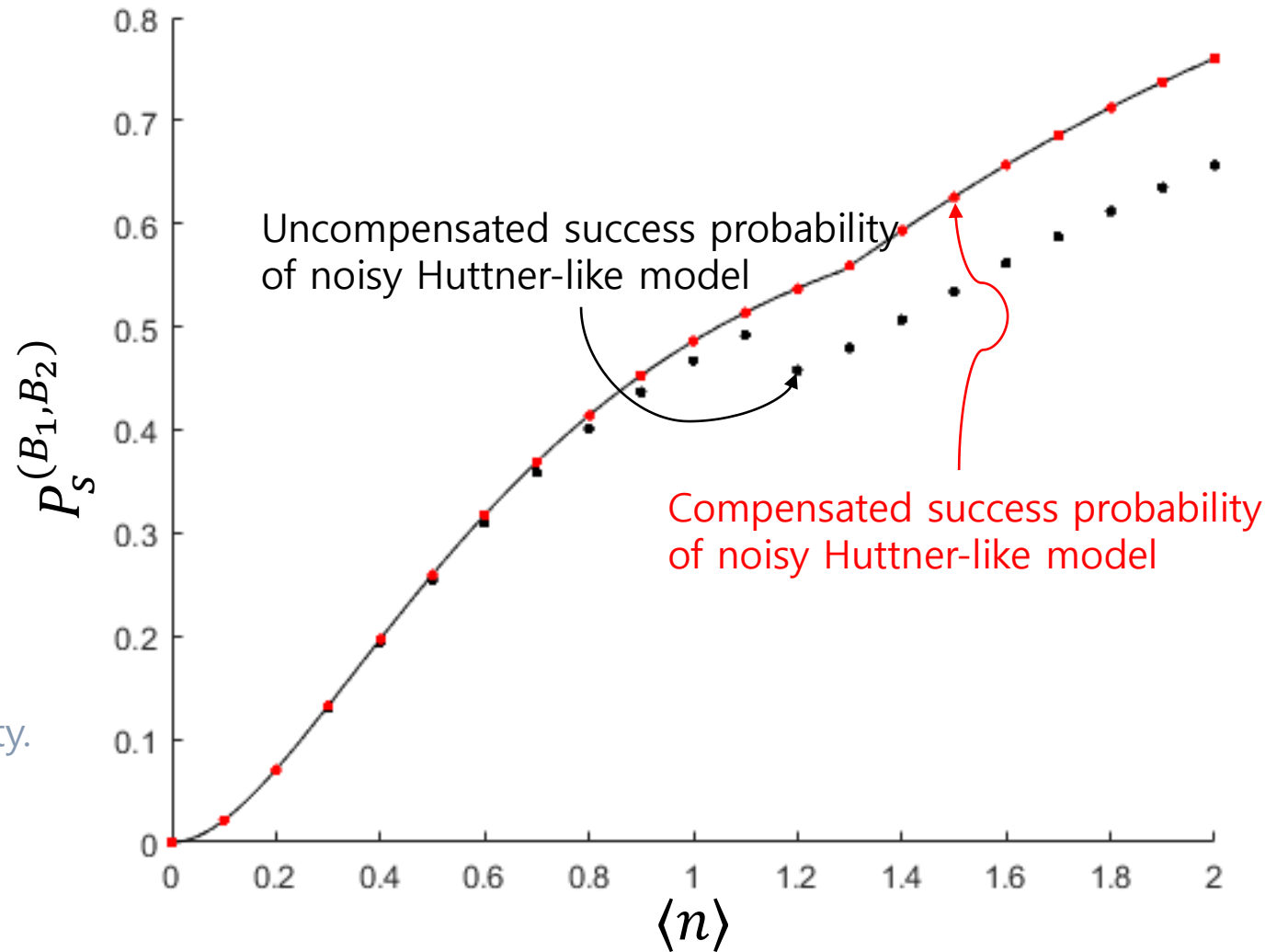
Sequential state discrimination in noisy channel

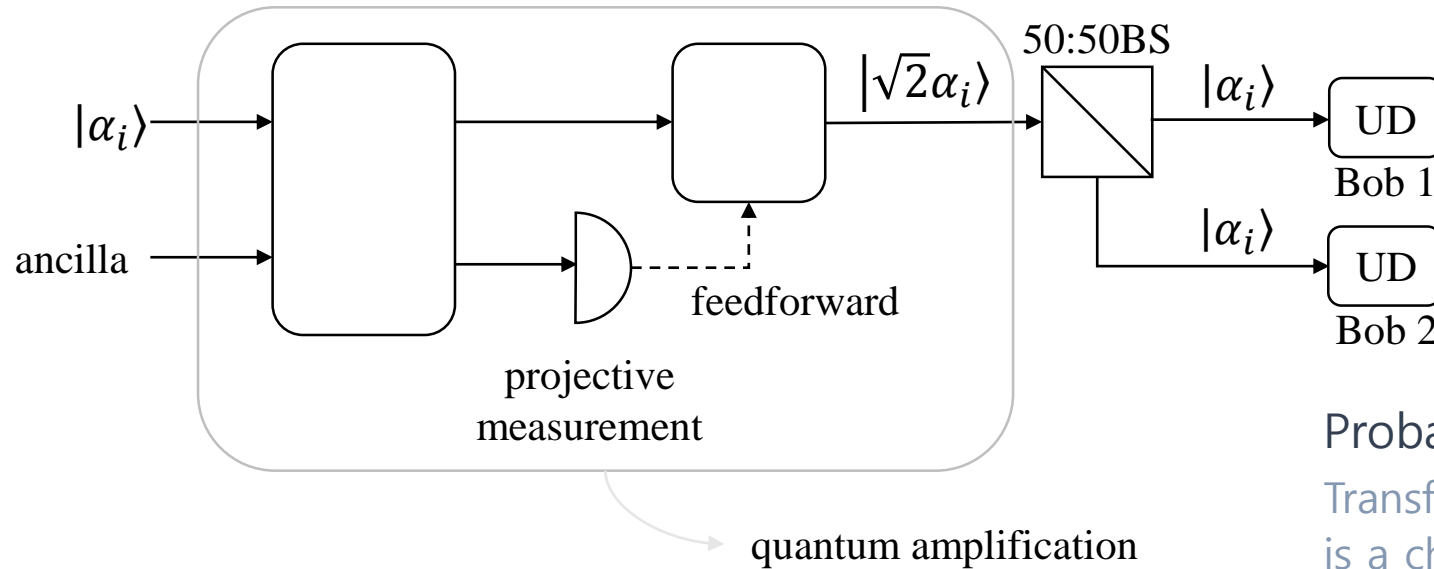
Noisy channel in Huttner-like model



In view of noisy channel:
Compensation does not decrease success probability.

In view of eavesdropping:
Eavesdropper cannot obtain any information.

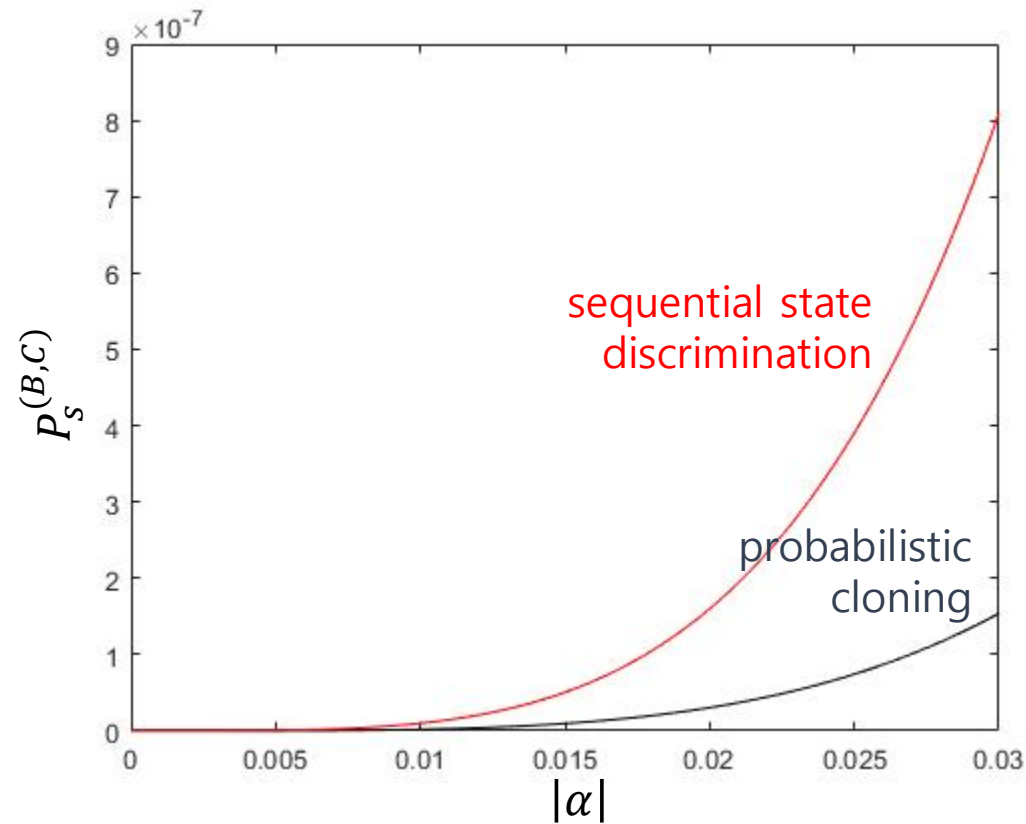




Probabilistic cloning using linear optics

Transforming an unknown coherent state $|\alpha\rangle$ into $|\sqrt{2}\alpha\rangle$ is a challenging task.

Optimal success probabilities



According to this graph, sequential state discrimination outperforms quantum probabilistic cloning strategy

○ Part I : Theory

- We express sequential state discrimination for N receivers in mathematical optimization problem.
- Sequential state discrimination of two pure states can be applied to multiparty QKD, when the number of receivers is not too many.
- Sequential state discrimination of two mixed states can be applied to multiparty QKD, even if the number of receivers is too many.
- Optimal success probability of mixed states sequential state discrimination exceeds that of quantum reproducing and quantum probabilistic broadcasting strategy. This means that sequential state discrimination is more suitable for multiparty QKD than other two strategies.

○ Part II : Application

- We propose the method to implement sequential state discrimination of two coherent states using linear optics.
- Probabilistic cloning cannot be performed using linear optics except for weak coherent states. Therefore, this strategy is difficult to be implemented technically.
- Optimal success probability of sequential state discrimination exceeds that of probabilistic cloning strategy.
- These two facts mean that sequential state discrimination of coherent states is more suitable for realistic multiparty QKD than probabilistic cloning strategy. Especially, Huttner-like model confirms security in realistic situation.

This talk is based on following works and recent results prepared in submission:

- M. Namkung and Y. Kwon, Phys. Rev. A **96**, 022318 (2017).
- M. Namkung and Y. Kwon, Scientific Reports **8**, 6515 (2018).
- M. Namkung and Y. Kwon, Scientific Reports **8**, 16915 (2018).

We plan to investigate sequential state discrimination of N quantum states. In case of three pure states, we may exploit following reference:

- D. Ha and Y. Kwon, Phys. Rev. A **91**, 062312 (2015).

Thank you for attention!

감사합니다!