

# Reinforcement Learning Using Quantum Optical Projective

Team 3

A. O. Parada Flores, D. J. Arbol Guerrero, F. Saravalle, M. Patraca Gonzalez, I. Tsoni

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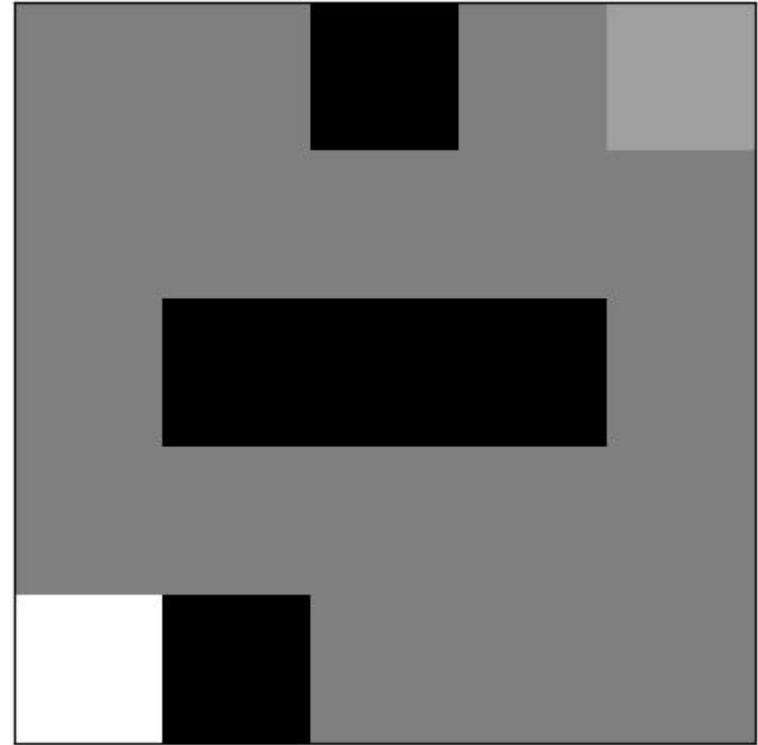
## Problem statement

### Legend

- white: starting position
- light grey: ending position
- black: wall(not accessible)

## Goal

Finding the paths going from white box to the light grey one, without passing through the walls.

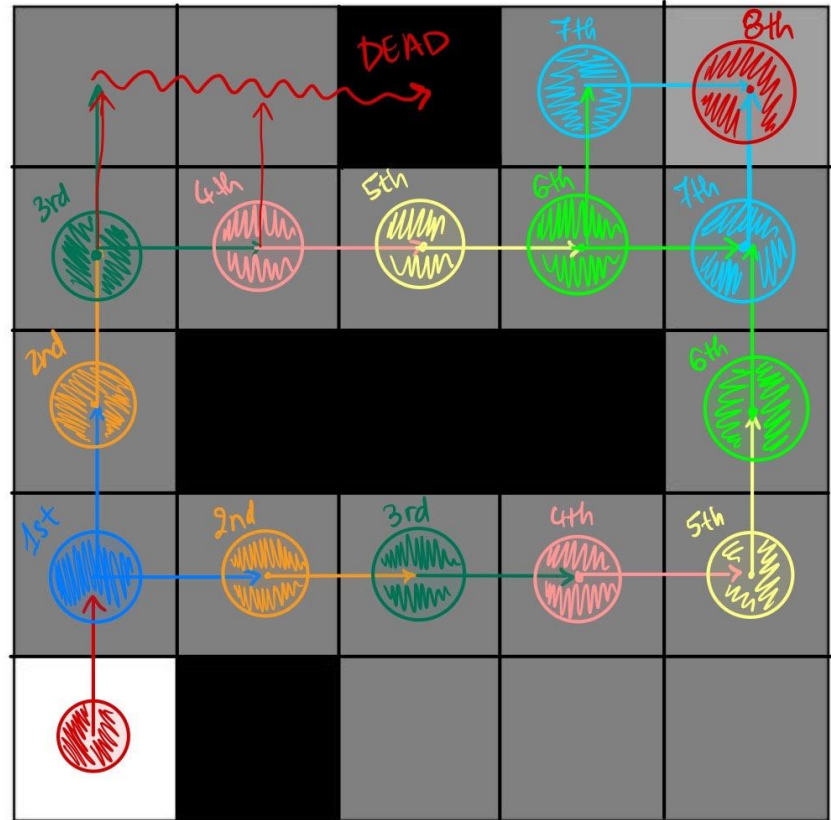


# Possible Path

## Simplification

Possible moves:  
right and up.

In the picture  
all possible solutions  
to the problem.



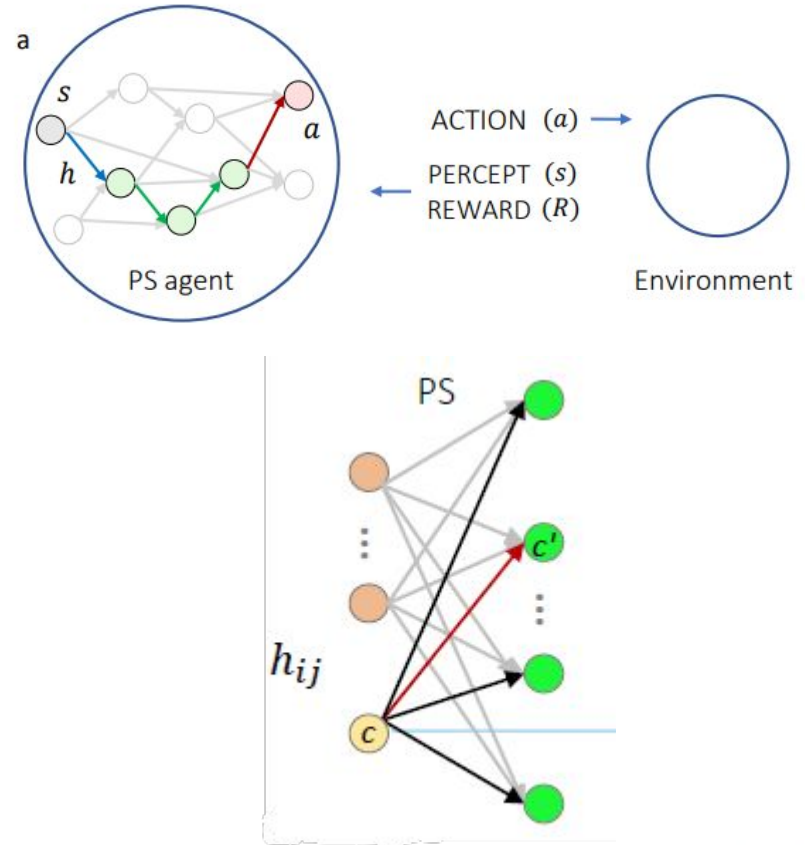
# Classic Projective Simulation

An “agent” navigates through an environment, doing a random walk, by projective his current state into a set of future states and decide which action to take.

$$p_{ij} = \frac{h_{ij}}{\sum_{j'} h_{ij'}}$$

Execute the action and evaluate with a reward system, which depends on the problem. Iterate to find the best paths.

$$h_{ij}^{(t+1)} = 1 + (1 - \gamma)(h_{ij}^{(t)} - 1) + g_{ij}^{(t+1)} R$$

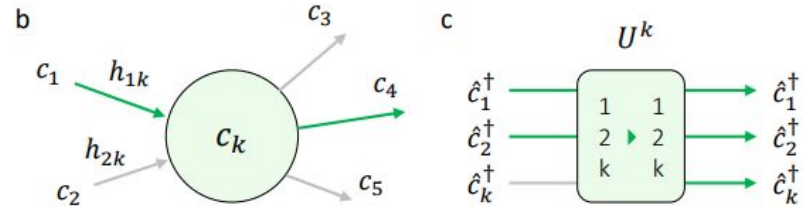


# Quantum Projective Simulation

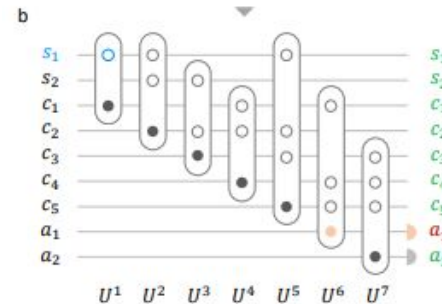
Decision-making is realized as a single-photon evolution in a mesh of tunable beam splitters with phase shifters.

The configuration is stored electronically in the phase shifters.

Each phase-shifter  $\theta_{kl}$  at node  $(k,l)$  is set to implement the transition probabilities for the corresponding clip-to-clip connections.

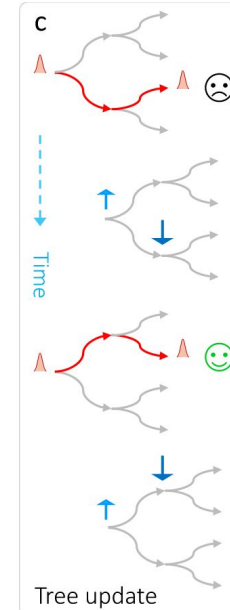
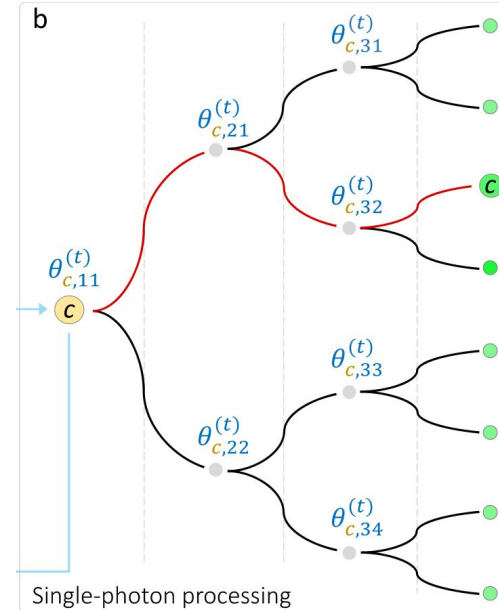
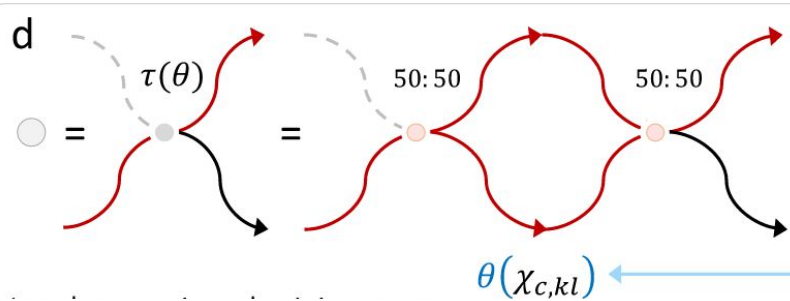


$$\hat{c}_j^\dagger(\tau_n) = \sum_{i=1}^{|C|} U_{ji}^{ECM} \hat{c}_i^\dagger(\tau_0); \quad U^{ECM}(\vec{\theta}) = \prod_{k=1}^n U^k(\vec{\theta})$$



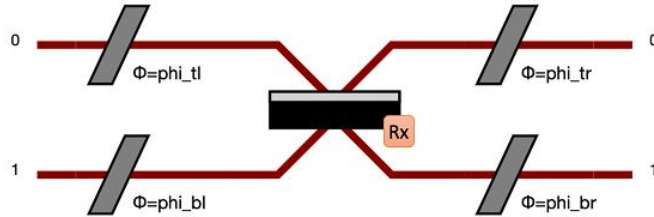
# Circuit Building

Arbitrary probabilistic transitions can be implemented on a photonic platform with a cascade of beamsplitters, whose transmissivities reproduce the distribution probability.



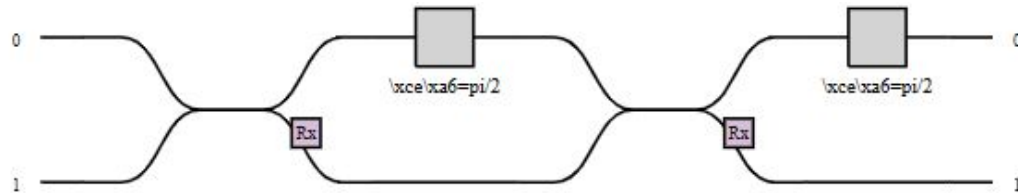
# Circuit Building

**Principal components:** Beam Splitter (BS), Phase Shifters (PS).



$$\begin{pmatrix} e^{i(\phi_{tl}+\phi_{tr})} \cos(\frac{\theta}{2}) & ie^{i(\phi_{tr}+\phi_{bl})} \sin(\frac{\theta}{2}) \\ ie^{i(\phi_{tl}+\phi_{br})} \sin(\frac{\theta}{2}) & e^{i(\phi_{br}+\phi_{bl})} \cos(\frac{\theta}{2}) \end{pmatrix}$$

## Mach-Zehnder Interferometer





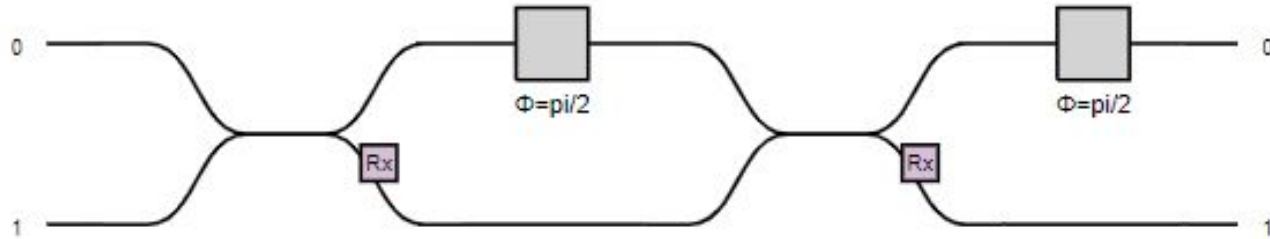
# Implementation

- Grid: 2D array  $n \times n$ .
- Reward: 0 for each step, 10 for reaching the end and -20 for crossing a wall.
- Creation of the optical circuit.
- Evaluation of the loss function to determine the best path.

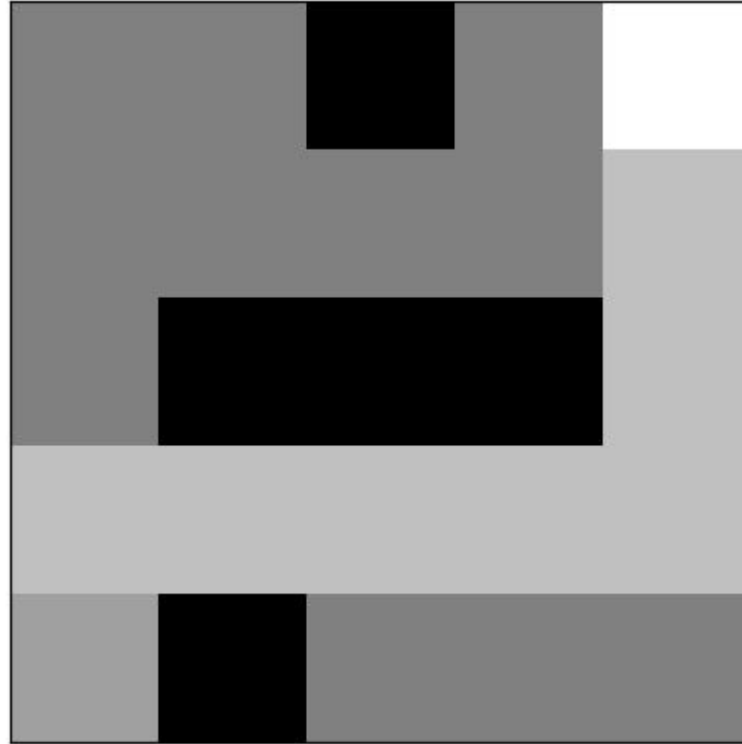
Same implementation with noise, emission probability 0.3

# Up-Right optimization

## Optical circuit



# Up-Right after optimization

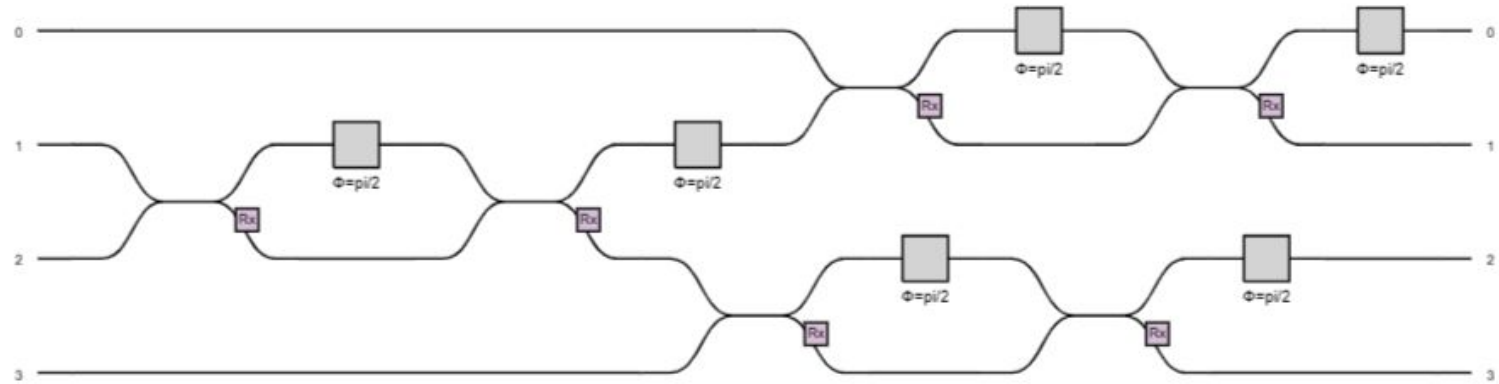


With or without noise we nearly get the same result



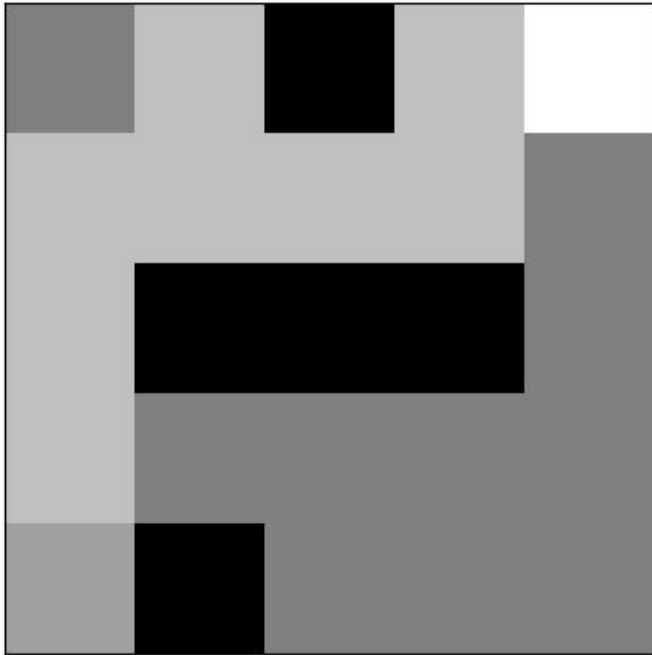
# Generalization

## Optical circuit

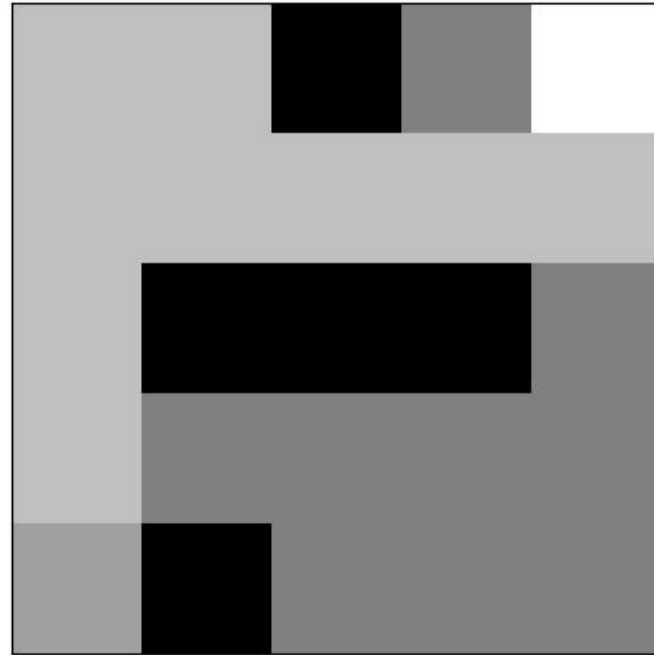


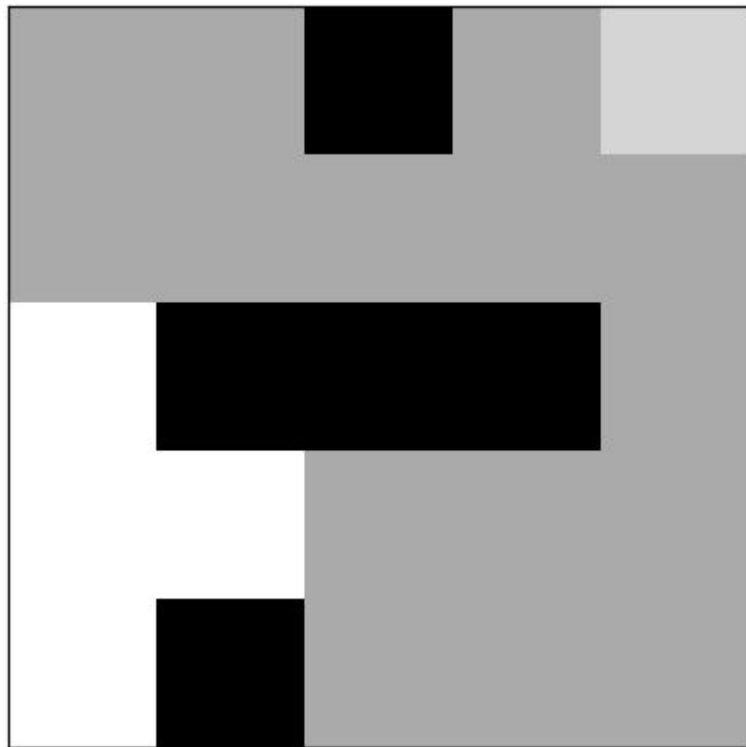
# All possible directions - after optimization

With noise



Without noise





$(0, 0)$  {0: 0.03619361979906159, 1: 0.962371789997033, 2: 0.0012302937247732286, 3: 0.00020429647913222087}

**Thank you!**