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# Game Theory (CS4187)

## Lectures 19, 20, 28, 29, 30, 42

Date: 01/10/2024

03/10/2024

04/11/2024

05/11/2024

07/11/2024



42 is a video  
lecture. No date.

Instructor: Gourav Saha

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# Breakup of the lectures

This lecture slides happened over a huge span of time. So, I will give a broad idea of these lectures:

1. **Lectures 19 and 20:** These lectures happened towards the end of module 2. I covered perfect information extended form game (PIEFG). Also, I gave the broad idea of sequential equilibrium.
2. **Lectures 28, 29, and 30:** These lectures happened immediately after the fall break and just before minor 2. Before these lectures, we already covered module 3 and also started module 4. So don't get confused when going over the notes. In this lectures, we covered:
  - Subgame perfect Nash equilibrium for imperfect information extended for games (IIEFGS). This happened in lectures 28 and 29.
  - I again introduced sequential equilibrium towards the end of lecture 29. And, during lecture 30, we solved an example to compute sequential equilibrium (SE).
3. **Lectures 42:** Lecture 42 is a video lecture. Consider this as a substitute of one of the four lectures that I didn't take. In this lecture, I will again explain SE (because I am not making notes for it). Also, I will solve another example. **IMPORTANT:** I will also talk about computing mixed strategy SE which I did not do during lectures.

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# Game Theory (CS4187)

## Lectures 19, 20

Date: 01/10/2024

03/10/2024

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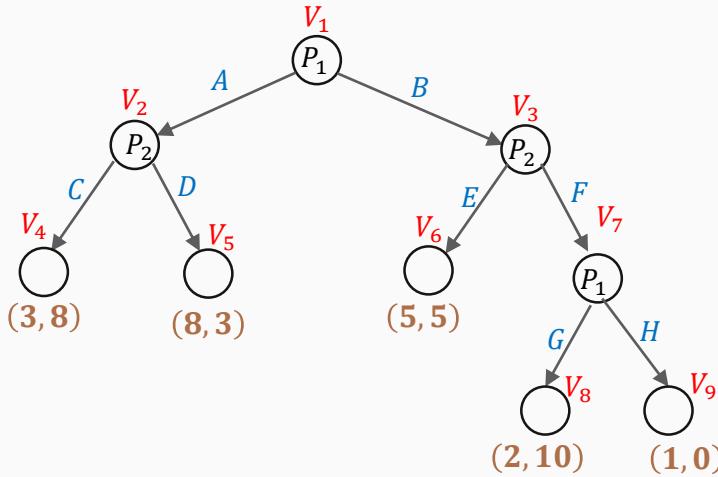
# Broad Idea of Today's Lecture

- Till now in module 2, we have discussed six solution concepts related to **strategic form game** (simultaneous move games).

The topics of this lecture is to extend few of the solution concepts of strategic form games to compute solutions for **extended form games** (sequential move games).

# Strategies for PIEFG: Examples

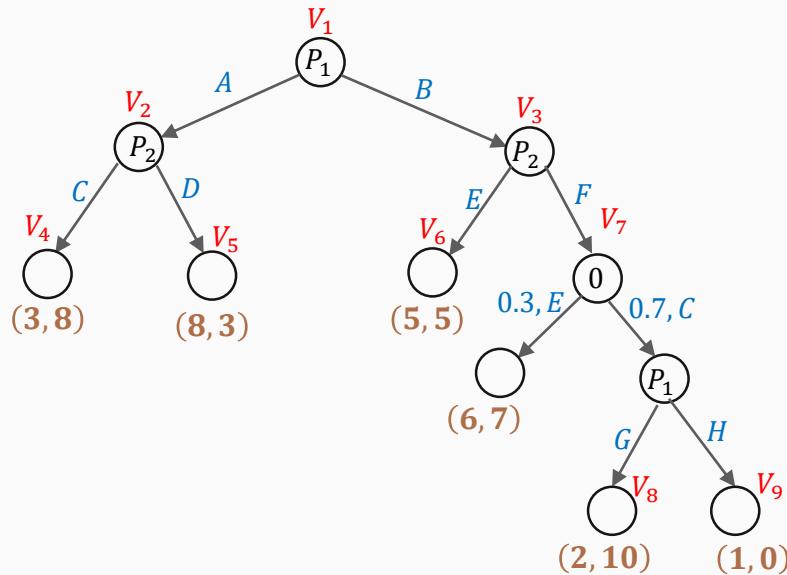
## Example 1



- Recall that EFGs are of two type: **perfect information** (PIEFG) and **imperfect information** (IIIEFG).
  - Both these types can be further sub-divided into games with **chances moves** and **without chance moves**.
- We start by discussing solution concepts of PIEFG, both with and without chance moves.
- To illustrate the concepts, we will use two examples shown in the left (for example 2, go to the next slide).

# Strategies for PIEFG: Examples

## Example 2



- Recall that EFGs are of two type: **perfect information** (PIEFG) and **imperfect information** (IIIEFG).
  - Both these types can be further sub-divided into games with **chance moves** and **without chance moves**.
- We start by discussing solution concepts of PIEFG, both with and without chance moves.
- To illustrate the concepts, we will use two examples shown in the left (for example 2, go to the next slide).

# Strategies for PIEFG

- Recall the definition of PIEFG with and without chance moves.

**Definition (PIEFG without Chance Moves):** A PIEFG without chance moves is a tuple  $\Gamma = \langle N, A, H, Z, \rho, \mathcal{X}, \sigma, u \rangle$ .

**Definition (PIEFG with Chance Moves):** A PIEFG without chance moves is a tuple  $\Gamma = \langle N, A, H, Z, \rho, \mathcal{X}, \sigma, u, (p_i)_{i \in H_0} \rangle$ .

- $N$  is the set of players,
- $A$  is the **set of actions**,
- $H$  is the **set of non-terminal vertices**,
- $Z$  is the **set of terminal vertices**,
- $\rho : H \rightarrow N$  is the **player function**,
- $\mathcal{X} : H \rightarrow 2^A$  is the **action function**,
- $\sigma : H \times A \rightarrow H \cup Z$  is the **successor function**,
- $u = (u_n)_{n \in N \setminus \{0\}}$ , where  $u_n : Z \rightarrow \mathbb{R}$  is the **utility function**,
- $H_0 = \{i \in H : \rho(i) = 0\}$  is the **set of non-terminal vertices for which the player is nature**. For every  $i \in H_0$ ,  $p_i$  is the **probability distribution** across actions in set  $\mathcal{X}(i)$ .

# Strategies for PIEFG

- A **pure strategy** for a player in PIEFG is a **complete specification** of which (deterministic) action to take at every vertices belonging to that player.
- Let,

$$H_i = \{v \in H : \rho(v) = i\}$$

is the set of non-terminal vertices where player  $i$  is playing. Let  $a_{i,j}$  denote the action that player  $i$  takes corresponding to  $H_{i,j}$  ( $H_{i,j}$  the  $j^{th}$  vertex in  $H_i$ ). Then the **pure strategy of player  $i$**  is,

$$s_i = (a_{i,1}, a_{i,2}, \dots, a_{i,|H_i|})$$

- The **set of strategies of player  $i$**  is the cartesian product of all  $\mathcal{X}(H_{i,j})$ ,

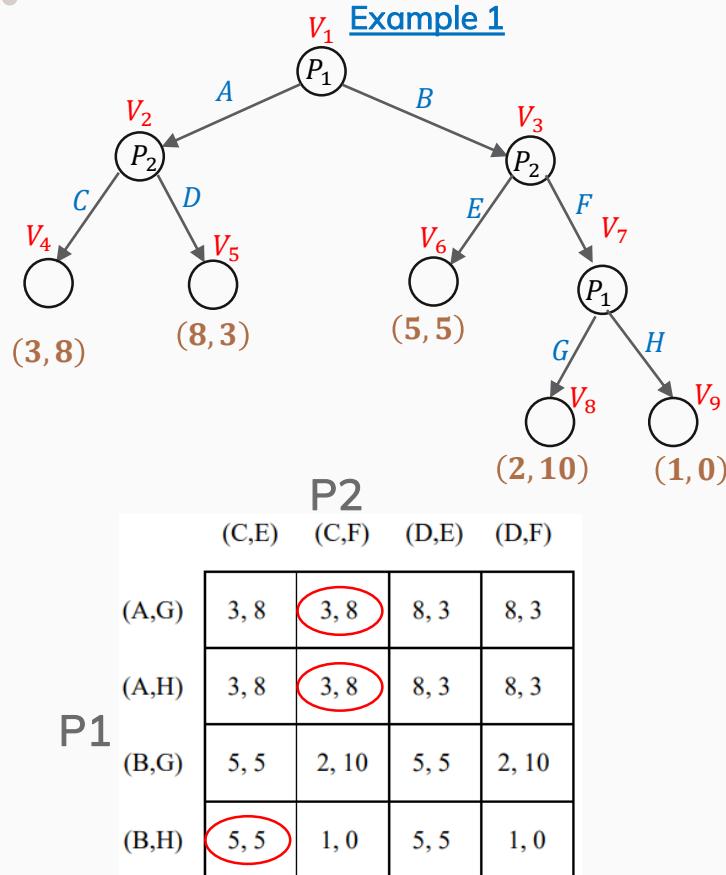
$$S_i = \mathcal{X}(H_{i,1}) \times \mathcal{X}(H_{i,2}) \times \dots \times \mathcal{X}(H_{i,|H_i|})$$

We have  $s_i \in S_i$ .

# Strategies for PIEFG

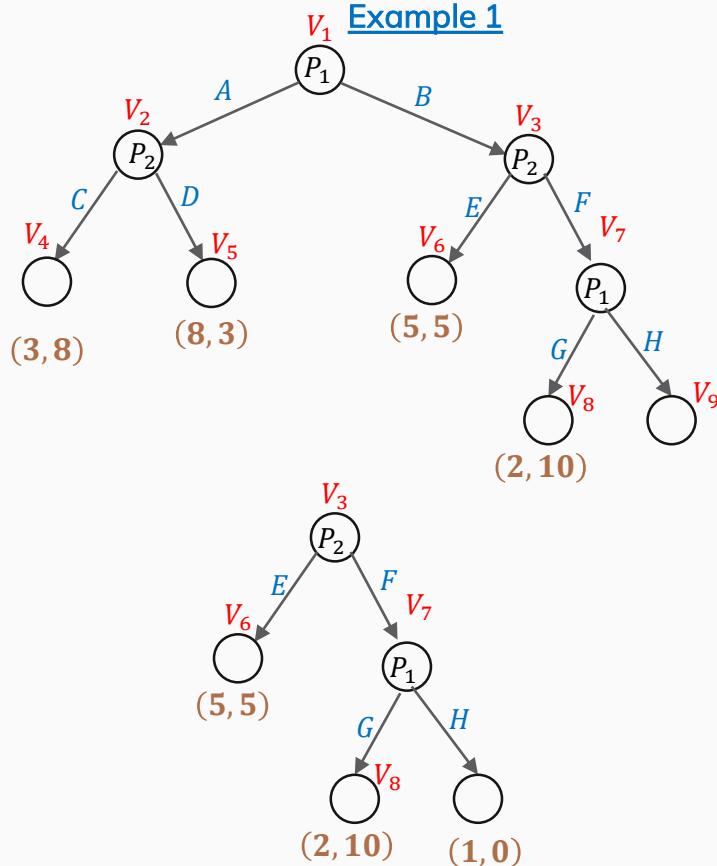
- Just like SFGs, in EFG too there is a concept of **strategy profile**. Strategy profile  $s = (s_i, s_{-i})$  is the action of all the players.
  - In EFG, each  $s_i$  is a **tuple**. This is one way in which strategy profile of SFGs and EFGs differ.
  - Even in SFGs  $s_i$  can be a tuple.
- As we know, we can convert a EFG to and SFG. Let  $\Gamma$  be the EFG. Let  $SFG(\Gamma)$  denote the SFG equivalent of  $\Gamma$ .
- Then all the solution concepts that we have learned about SFGs till now can be directly applied to  $SFG(\Gamma)$  to find the solution of EFG  $\Gamma$ .
  - The problem is that these solutions does not utilizes the sequential nature of the game and is hence may lead to illogical solutions.

# PSNE of PIEFG



- $SFG(\Gamma)$  for this game is shown in the table (directly adapted from Chapter 5 of the book Multiagent Systems by Shoham and Brown).
  - Remember, PSNE in context of EFG is related to  $SFG(\Gamma)$ , i.e. PSNE is related to the SFG equivalent of EFG.
  - PSNE of the game shown in the left is marked with the red circles. PSNEs are:
    - $((A, G), (C, F))$
    - $((A, H), (C, F))$
    - $((B, H), (C, F))$
  - PSNE  $((B, H), (C, E))$  is not logical! Why?

# Subgame



**Definition (Subgame):** Given a EFG  $\Gamma$ , the subgame of  $\Gamma$  at non-terminal vertex  $h$ , denoted by  $\Gamma(h)$ , is the EFG described by the subtree rooted at vertex  $h$ .

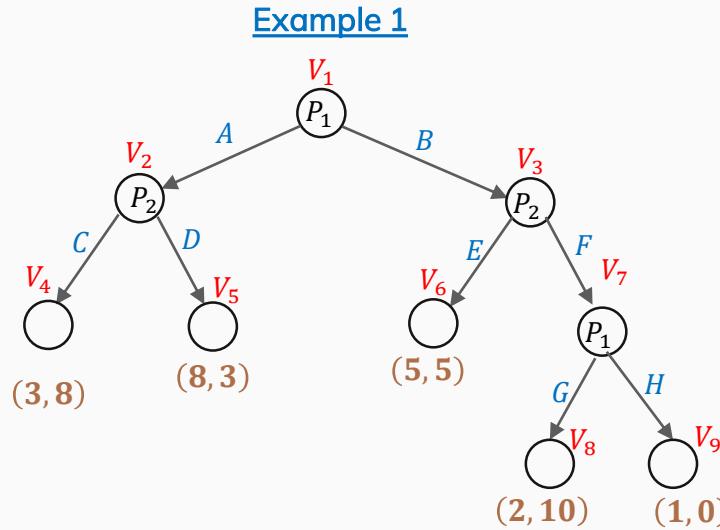
➤ Example:  $\Gamma(V_3)$  shown in the left.

# Subgame Perfect Nash Equilibrium (SPNE)

**Definition (Pure Strategy SPNE):** Pure strategy SPNE of an EFG  $\Gamma$  are all strategy profiles  $s$  such that for all  $h \in H$ , the restriction of  $s$  to  $\Gamma(h)$  is a PSNE of  $SFG(\Gamma(h))$ .

- Basically  $s$  should be a PSNE for all the subgames, then only  $s$  SPNE. **What is the logic?**
  - In order to capture the sequentially nature of the game, i.e. strategy  $s$  has to be rational after a sequence of actions was taken by other players.
- Most of the time, rather than saying “PSNE of  $SFG(\Gamma(h))$ ”, we will directly say “PSNE of  $\Gamma(h)$ ”.
- What does “restriction of  $s$ ”?

# Backward Induction

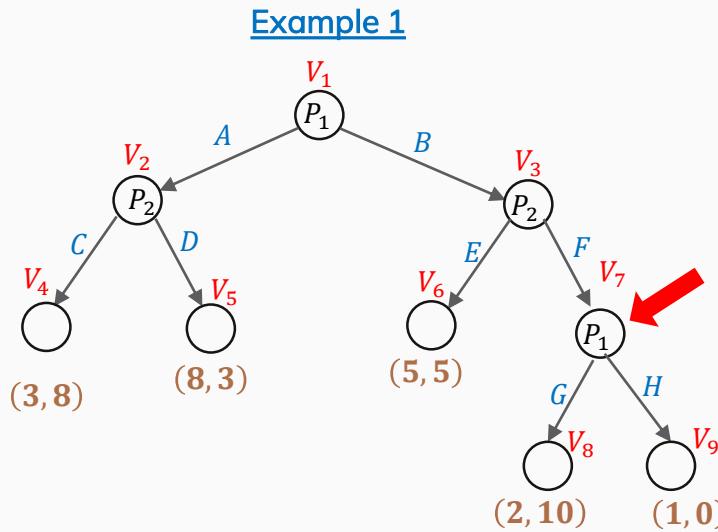


- SPNE for PIEFG can be found using Backward induction. To learn about backward induction, look into chapter 8 of the book:

Harrington, J.E., "Games, Strategies, and Decision Making", 2nd edition.

- Compute SPNE of the PIEFG shown in the left. In the next few slides, we demonstrate backward induction.

# Backward Induction



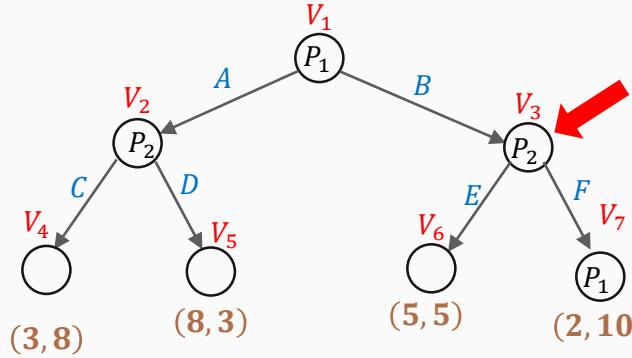
- SPNE for PIEFG can be found using Backward induction. To learn about backward induction, look into chapter 8 of the book:

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- P1 is playing in the vertex with marked with the red arrow. Action G has a payoff of 2 and H has a payoff of 1. So, P1 chooses G.

# Backward Induction

## Example 1

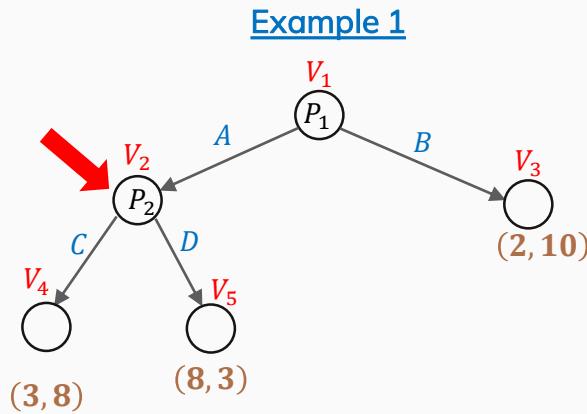


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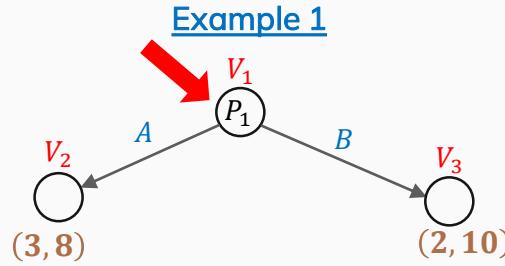
- P2 is playing in the vertex with marked with the red arrow. Action F has a payoff of 10 and E has a payoff of 5. So, P2 chooses F.

# Backward Induction



- SPNE for PIEFG can be found using Backward induction. To learn about backward induction, look into chapter 8 of the book:  
**Harrington, J.E., "Games, Strategies, and Decision Making", 2nd edition.**
- P2 is playing in the vertex with marked with the **red arrow**. Action C has a payoff of 8 and C has a payoff of 3. So, P2 chooses C.

# Backward Induction



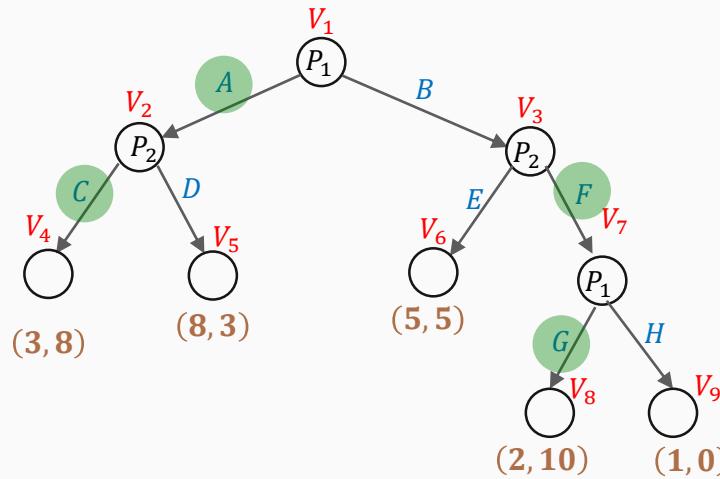
- SPNE for PIEFG can be found using Backward induction. To learn about backward induction, look into chapter 8 of the book:

Harrington, J.E., "Games, Strategies, and Decision Making", 2nd edition.

- P1 is playing in the vertex with marked with the red arrow. Action A has a payoff of 3 and B has a payoff of 2. So, P1 chooses A.

# Backward Induction

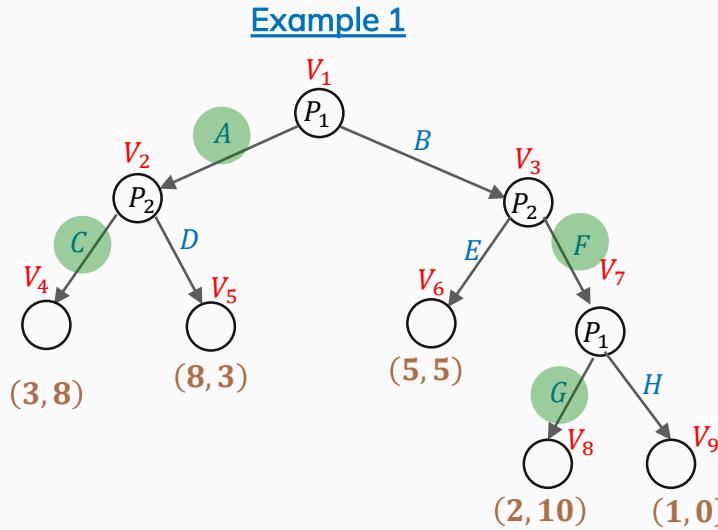
## Example 1



➤ So, the SPNE is,

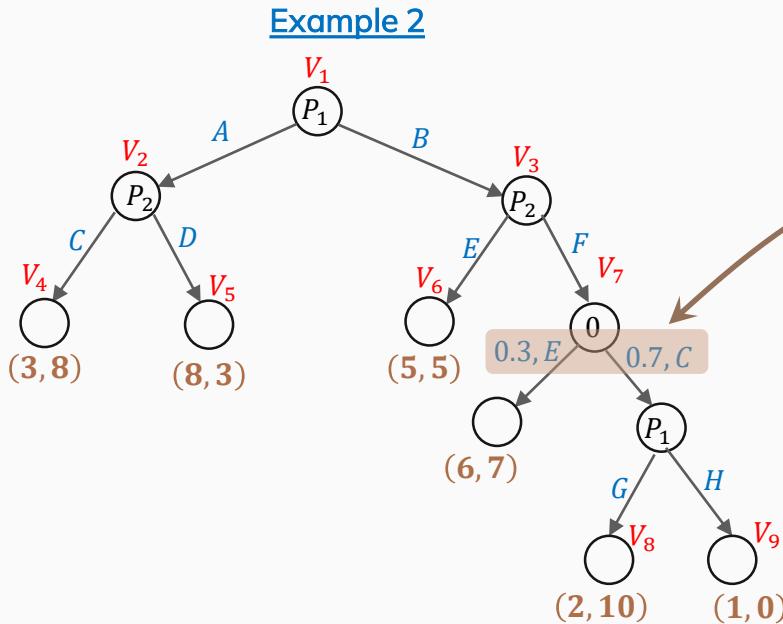
$$((A, G), (C, F))$$

# Backward Induction



- So, the SPNE is,  
 $((A, G), (C, F))$
- Is the computed SPNE one of the NEs that was discussed in [this slide \(click here\)](#)?
- Is it always true that an SPNE of an EFG  $\Gamma$  will be an NE of the induced SFG,  $SFG(\Gamma)$ ?

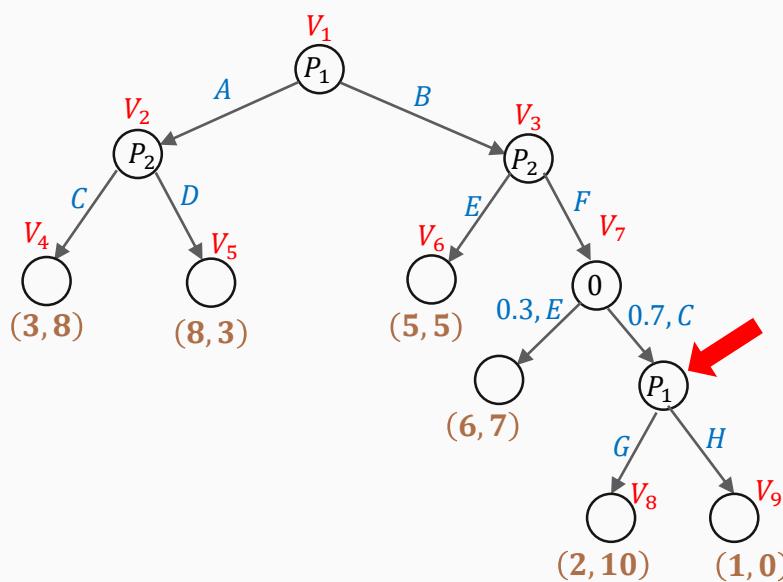
# Backward Induction



➤ Compute SPNE of the PIEFG **with chance moves** shown in the left. In the next few slides, we demonstrate backward induction.

# Backward Induction

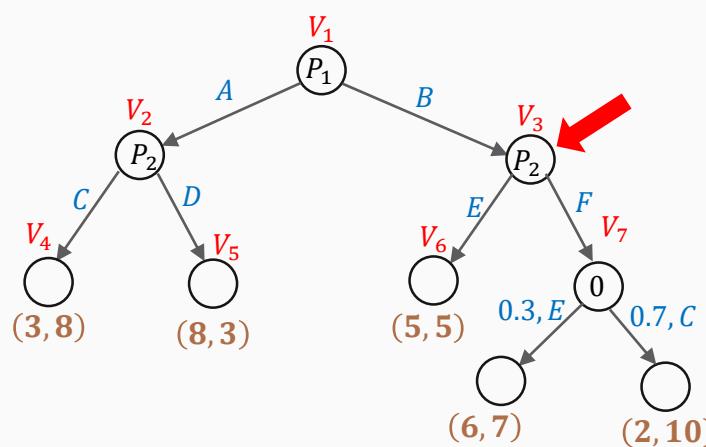
Example 2



- Compute SPNE of the PIEFG **with chance moves** shown in the left. In the next few slides, we demonstrate backward induction.
- P1 is playing in the vertex with marked with the **red arrow**. Action G has a payoff of 2 and H has a payoff of 1. So, P1 chooses G.

# Backward Induction

Example 2



➤ Compute SPNE of the PIEFG **with chance moves** shown in the left. In the next few slides, we demonstrate backward induction.

➤ P2 is playing in the vertex with marked with the red arrow. Action E has a payoff of 5 and F has a payoff of:

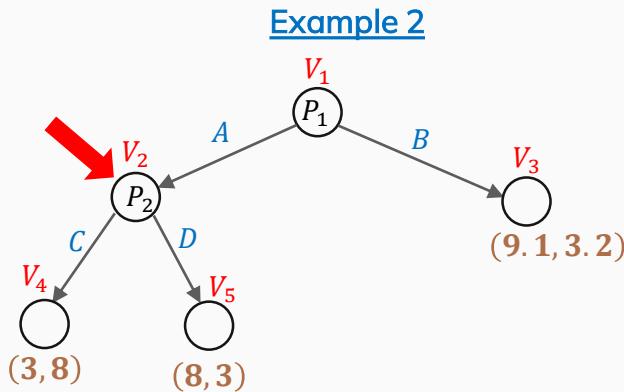
$$0.3 * 7 + 0.7 * 10 = 9.1$$

The above payoff is for P2. The payoff of P1 for action F is,

$$0.3 * 6 + 0.7 * 2 = 3.2$$

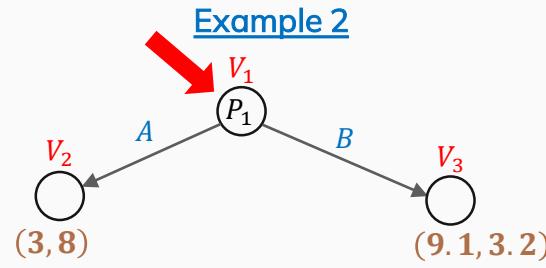
P2 chooses F because 9.1 is greater than 5.

# Backward Induction



- Compute SPNE of the PIEFG **with chance moves** shown in the left. In the next few slides, we demonstrate backward induction.
- P2 is playing in the vertex with marked with the **red arrow**. Action C has a payoff of 8 and C has a payoff of 3. So, P2 chooses C.

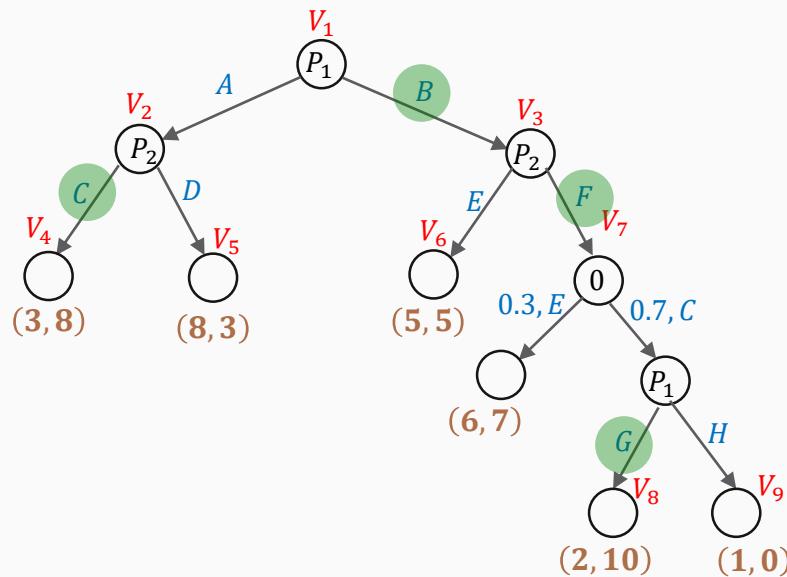
# Backward Induction



- Compute SPNE of the PIEFG **with chance moves** shown in the left. In the next few slides, we demonstrate backward induction.
- P1 is playing in the vertex with marked with the **red arrow**. Action B has a payoff of 9.1 and A has a payoff of 3. So, P1 chooses B.

# Backward Induction

Example 2



➤ So, the SPNE is,

$$((B, G), (C, F))$$

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# Game Theory (CS4187)

## Lectures 28, 29

Date: 04/11/2024

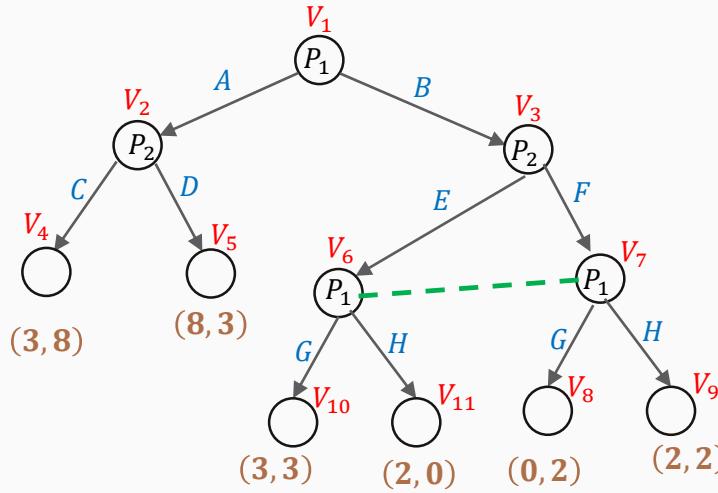
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# Solution Concepts for IIEFG

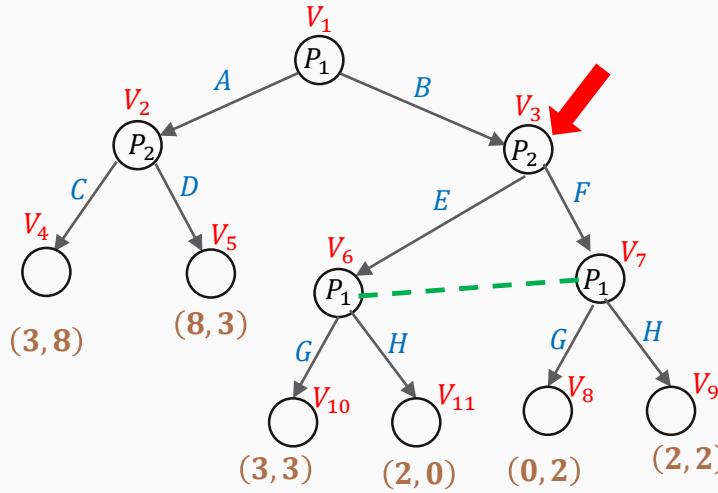
- For PIEFG, a pure strategy SPNEs/PSNEs always exists. This is a natural consequence of backward induction.
- But for IIEFG, it is not guaranteed that pure strategy SPNEs will exist. How?
  - All SFGs can be expressed as EFGs.
  - We know that there are SFGs for which PSNEs don't exist.
  - Since SPNE is a subset of PSNE, SPNE may not exist.
- We will talk about the following solution concepts of IIEFGs:
  - SPNEs.
  - Sequential equilibrium (not coming for minor 2).

# SPNE for IIEFGs



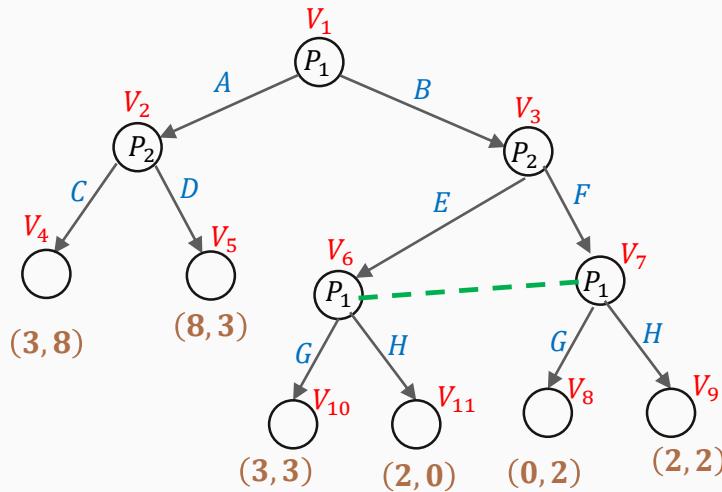
- We will demonstrate the concept using the example shown in the left.
- We can't solve imperfect information EFGs directly using backward induction. That said the overall idea of backward induction is still useful of IIEFGs.
- Backward induction can't be directly used because two nodes in the same information set can't have two different strategies as it is not possible to distinguish between the nodes in an information set. Example: P1 can take action G at vertex V6 and H at vertex V7 just because those actions gives better payoff in the corresponding vertices.

# SPNE for IIEFGs



- We can see however that the subgame starting at vertex  $V_3$  (marked with red arrow) can definitely be solved. Infact, the subgame at  $V_3$  is a **simultaneous move game** between P1 and P2.
- Since, it is a simultaneous move game we can find its NEs as we did for SFGs. This is shown in the next slide.

# SPNE for IIEFGs



P2

	E	F
G	(3, 3)	(0, 2)
H	(2, 0)	(2, 2)

P1

➤ NEs are:

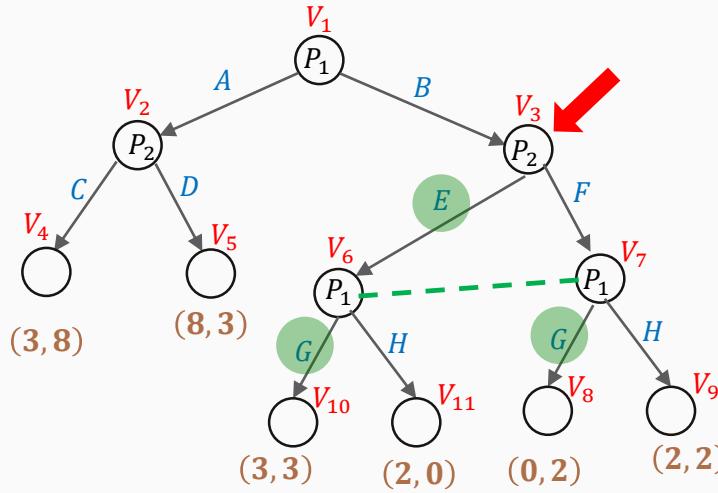
(G, E)  
(H, F)

$\left( \left( \frac{2}{3} (G), \frac{1}{3} (H) \right), \left( \frac{2}{3} (E), \frac{1}{3} (F) \right) \right)$

} PSNEs  
} MSNE

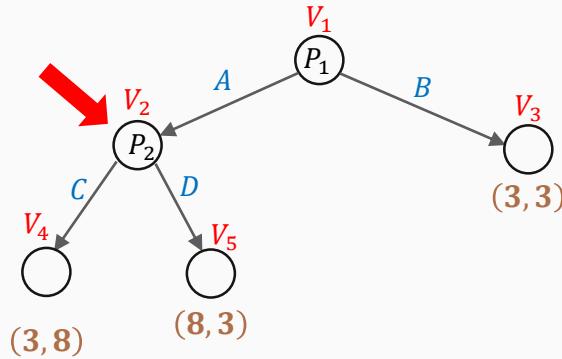
➤ For each of these NEs, we can do backward induction as before. We show it for one PSNE and one MSNE in the next few slides.

# SPNE for IIEFGs



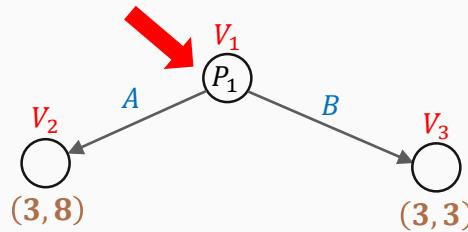
- Backward induction for NE  $(G, E)$ .
- The actions are highlighted using green circles. The corresponding payoff is  $(3, 3)$ .

# SPNE for IIEFGs



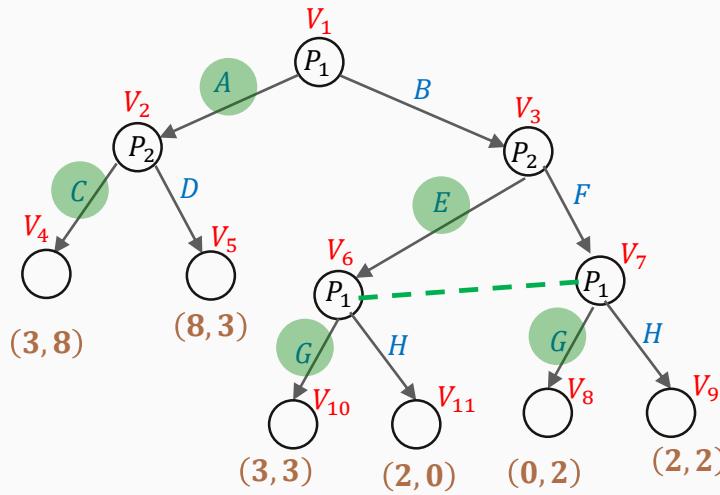
- Backward induction for NE  $(G, E)$ .
- P2 is playing in the vertex with marked with the red arrow. Action C has a payoff of 8 and C has a payoff of 3. So, P2 chooses C.

# SPNE for IIEFGs



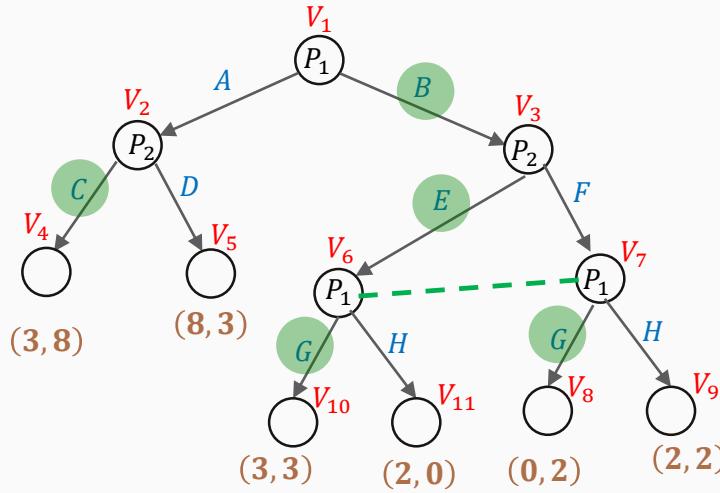
- Backward induction for NE  $(G, E)$ .
- P2 is playing in the vertex with marked with the red arrow. Both actions A and B has the same payoffs. So both are possible. In the next two slides, we will show the SPNE for actions A and B separately.

# SPNE for IIEFGs



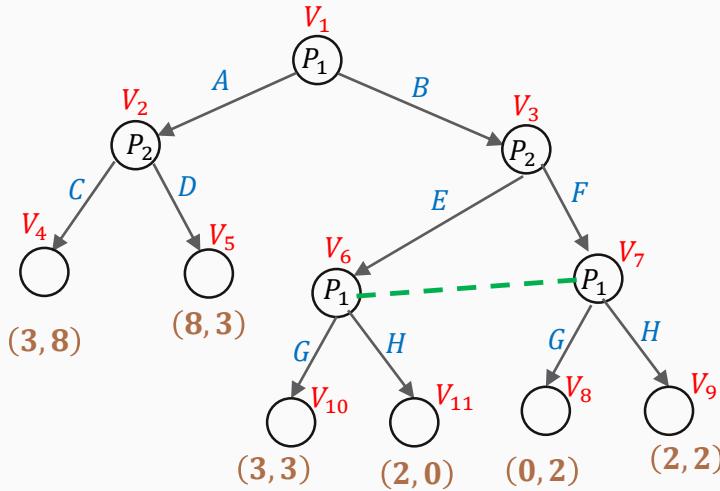
- Backward induction for NE  $(G, E)$ .
- Let the action taken by P1 at vertex V1 be A.
- SPNE is:  
 $((A, G), (C, E))$

# SPNE for IIEFGs



- Backward induction for NE  $(G, E)$ .
- Let the action taken by P1 at vertex V1 be B.
- SPNE is:  
 $((B, G), (C, E))$
- So essentially, there are two SPNEs when the subgame starting at node V3 has NE  $(G, E)$ . They are:  
 $((A, G), (C, E))$   
 $((B, G), (C, E))$

# SPNE for IIEFGs



➤ Backward induction for NE

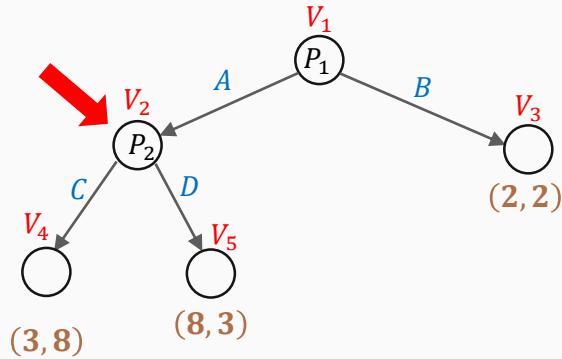
$$\left( \left( \frac{2}{3} (G), \frac{1}{3} (H) \right), \left( \frac{2}{3} (E), \frac{1}{3} (F) \right) \right)$$

➤ The payoff of the P1 and P2 for the above NE for the subgame at vertex V3 is

$$P1: \frac{2}{3} * \frac{2}{3} * 3 + \frac{2}{3} * \frac{1}{3} * 2 + \frac{1}{3} * \frac{2}{3} * 0 + \frac{1}{3} * \frac{1}{3} * 2 = 2$$

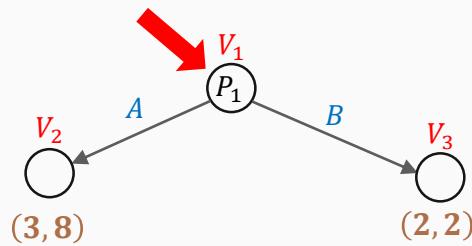
$$P2: \frac{2}{3} * \frac{2}{3} * 3 + \frac{2}{3} * \frac{1}{3} * 0 + \frac{1}{3} * \frac{2}{3} * 2 + \frac{1}{3} * \frac{1}{3} * 2 = 2$$

# SPNE for IIEFGs



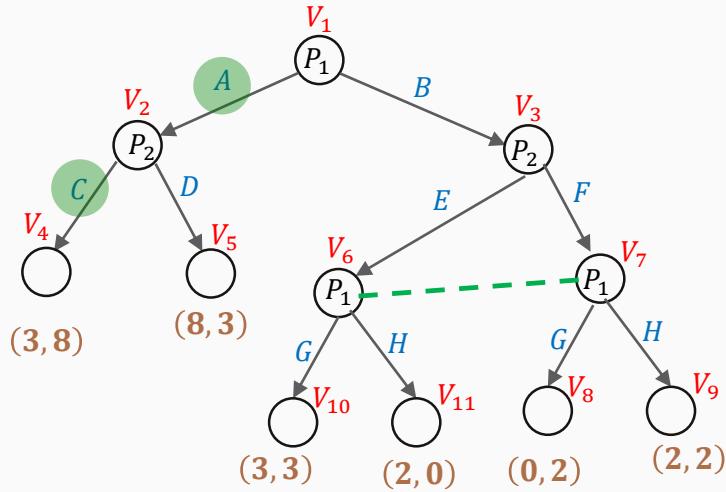
- Backward induction for NE $\left(\left(\frac{2}{3} (G), \frac{1}{3} (H)\right), \left(\frac{2}{3} (E), \frac{1}{3} (F)\right)\right)$
- P2 is playing in the vertex with marked with the red arrow. Action C has a payoff of 8 and C has a payoff of 3. So, P2 chooses C.

# SPNE for IIEFGs



- Backward induction for NE
$$\left(\left(\frac{2}{3} (G), \frac{1}{3} (H)\right), \left(\frac{2}{3} (E), \frac{1}{3} (F)\right)\right)$$
- P1 is playing in the vertex with marked with the red arrow. Action A has a payoff of 3 and action B has a payoff of 2. P1 choose A.

# SPNE for IIEFGs



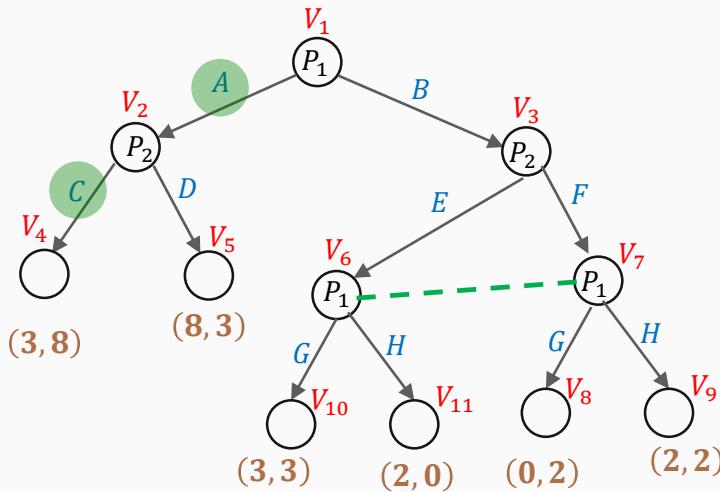
➤ Backward induction for NE

$$\left( \left( \frac{2}{3} (G), \frac{1}{3} (H) \right), \left( \frac{2}{3} (E), \frac{1}{3} (F) \right) \right)$$

➤ SPNE is,

$$\left( \left( A, \left( \frac{2}{3} (G), \frac{1}{3} (H) \right) \right), \left( C, \left( \frac{2}{3} (E), \frac{1}{3} (F) \right) \right) \right)$$

# SPNE for IIEFGs



➤ So the SPNEs that we found till now are:

$$(A, G), (C, E)$$

$$(B, G), (C, E)$$

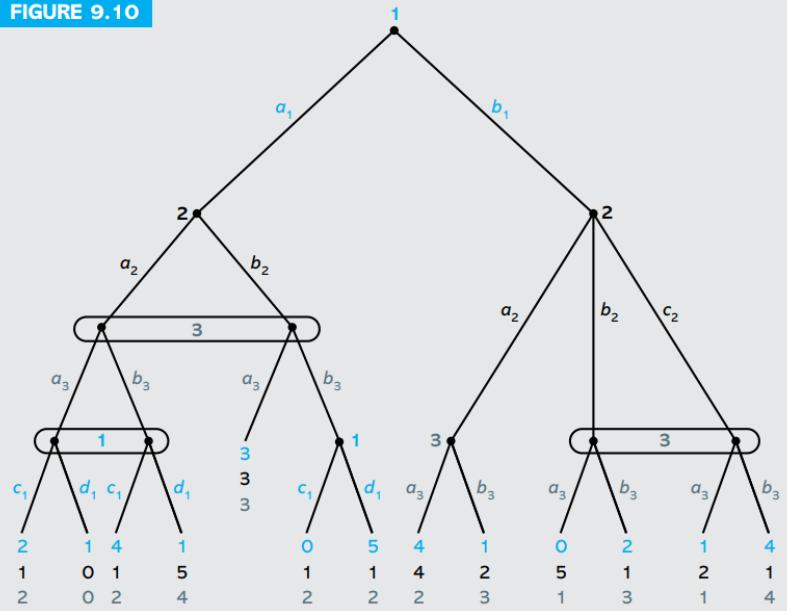
$$\left( \left( A, \left( \frac{2}{3} (G), \frac{1}{3} (H) \right) \right), \left( C, \left( \frac{2}{3} (E), \frac{1}{3} (F) \right) \right) \right)$$

➤ Find the SPNE when the NE for the subgame at vertex V3 is,

$$(H, F)$$

# Solution Concepts for IIEFG

FIGURE 9.10

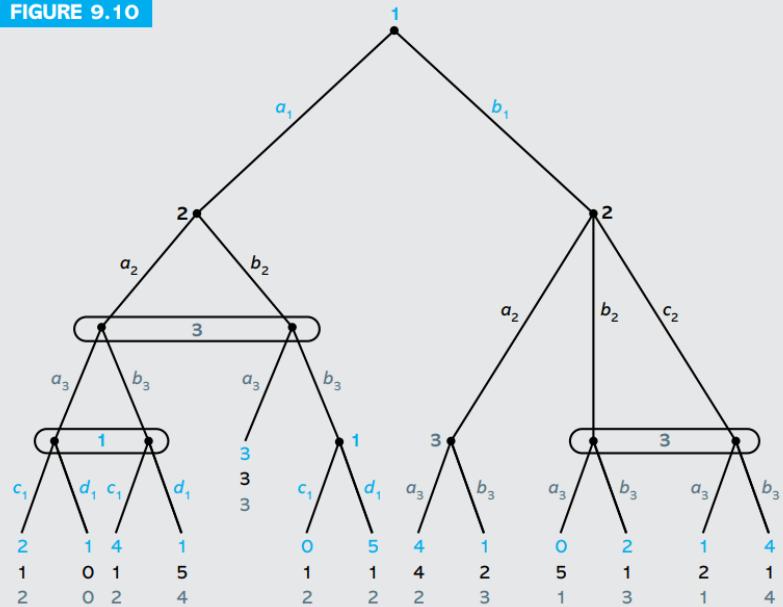


## The Need for a Systematic Approach to Find Sub-Games

- For the problem that we discussed in the last few slides, the sub-game was easy to identify. This is always not so in case of imperfect information game.
- Consider the game shown in the left (directly adapted from the book by Harrington). What are the subgames?

# Solution Concepts for IIEFG

FIGURE 9.10



The Need for a Systematic Approach to Find Sub-Games

**Definition (Subtree):** A sub-tree of  $T$  is a tree  $T'$  consisting of a node in  $T$  (called the **root of the sub-tree**) and **ALL of its descendants in  $T$** , along with ALL the edges that connect these nodes.

**Definition (Regular Subtree):** A regular sub-tree is a subtree that contains all information sets that have at least one node in the subtree.

In other words, to be a regular subtree, a subtree that includes one node of an information set must include all nodes of that information set.

# Game Theory (CS4187)

## Lectures 29

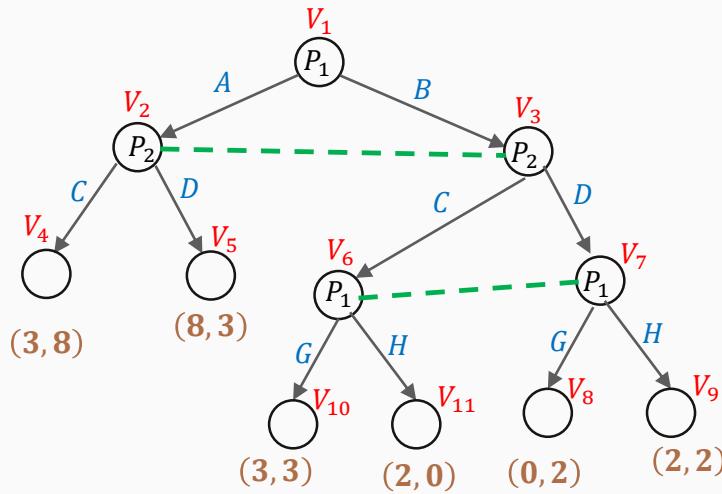
Date: 05/11/2024

**YouTube video:**

<https://youtu.be/huKiQ5mVlis>

The video says that it is part 1 of lecture 42. This is because I recorded it along with lecture 42. But the topic covered in this video were the topics that we discussed in lecture 29. I have recorded the video because the lecture slides are not sufficient.

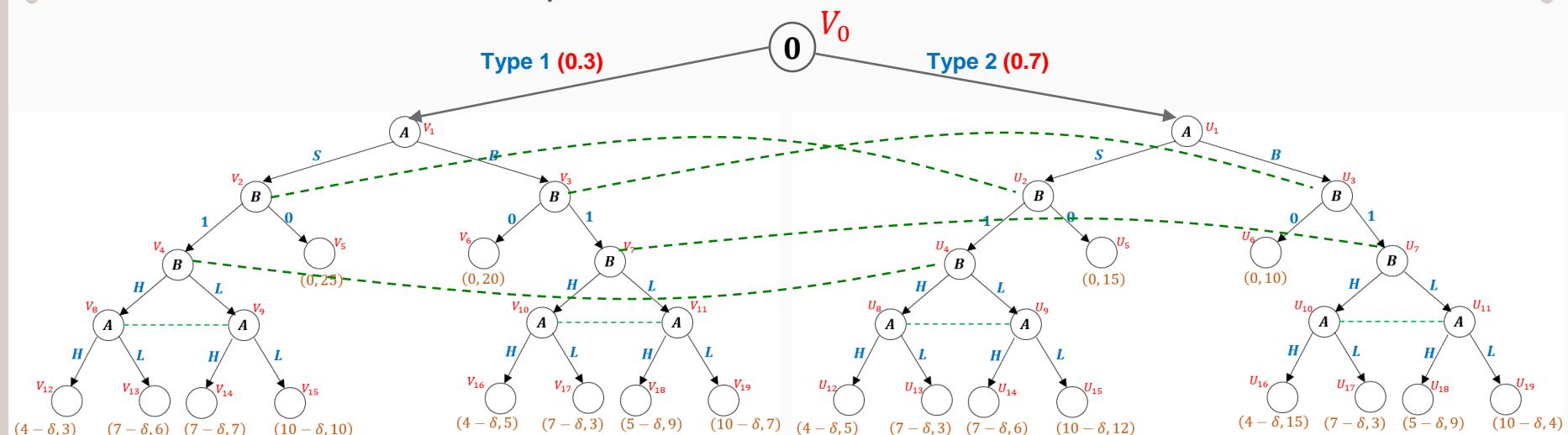
# Solution Concepts for IIEFG



## Sequential Equilibrium

- What is the need to come up with another equilibrium concept?
- Consider the EFG in the left. Player P1 in the information set  $\{V6, V7\}$  definitely knows that P1 played B in vertex V1.
  - One may argue that a player (P1 here) will definitely remember its move before but the point conveyed above remains true if another player (say P3) was playing in information set  $\{V6, V7\}$ .

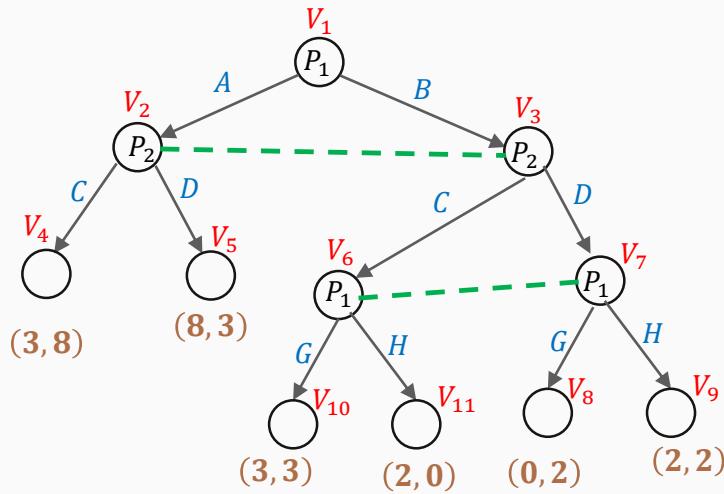
# Solution Concepts for IIEFG



## Sequential Equilibrium

- What is the need to come up with another equilibrium concept?
- For Bayesian sequential move games, for most case we will not have any sub-games!

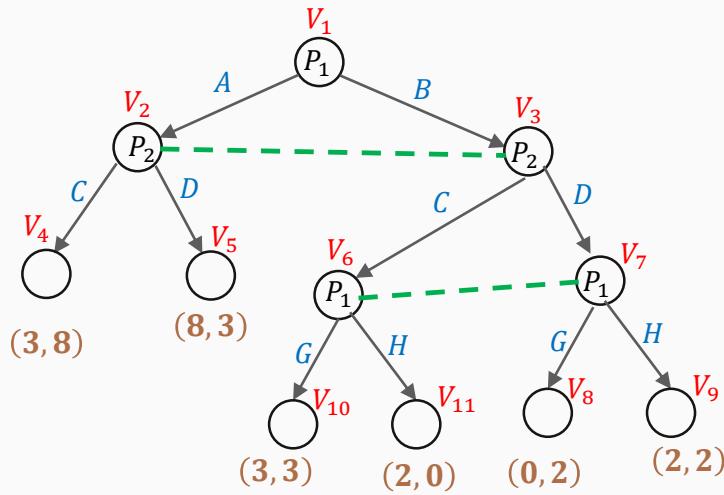
# Solution Concepts for IIEFG



## Sequential Equilibrium

- A sequential equilibrium (SE)  $(s, \mu)$  consists of:
  - $s$  is the strategy (pure or mixed) at every information set.
  - $\mu$  is the belief. It containing belief vectors for every information set. Belief vector of an information set contains the belief the player corresponding to that information set holds about its probability of being in individual nodes of that information set.

# Solution Concepts for IIEFG



## Sequential Equilibrium

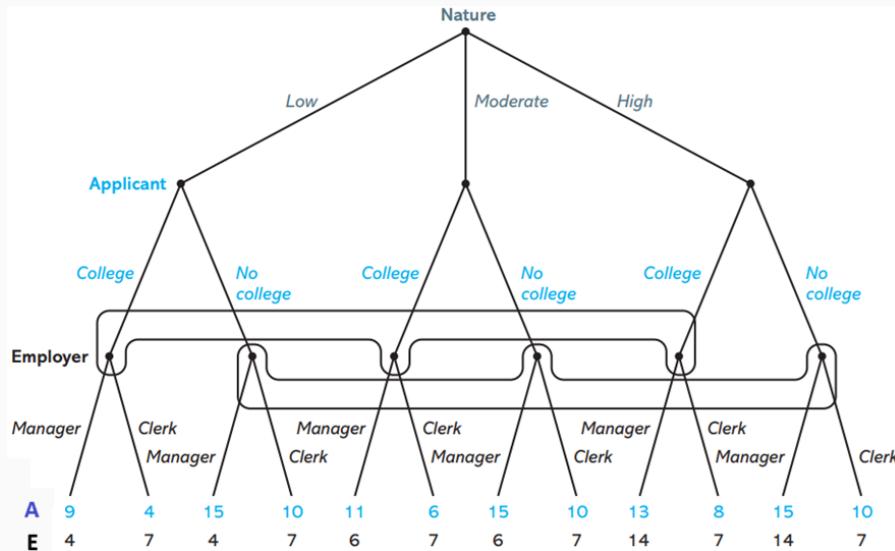
- A sequential equilibrium (SE)  $(s, \mu)$  must satisfy the following criteria:

**Sequential rationality:** Strategy  $s$  must be sequentially rationally given the belief  $\mu$ .

**Consistent belief:** Belief  $\mu$  must be consistent given the strategy  $s$ .

- Use Bayes' rule to compute  $\mu$ .

# Computing Sequential Equilibrium (Pure Strategy)



➤ This was an example that was done during lecture 30. I am not going to solve it again here. But I will solve another example in the following slides.

# Game Theory (CS4187)

## Lectures 30

Date: 07/11/2024

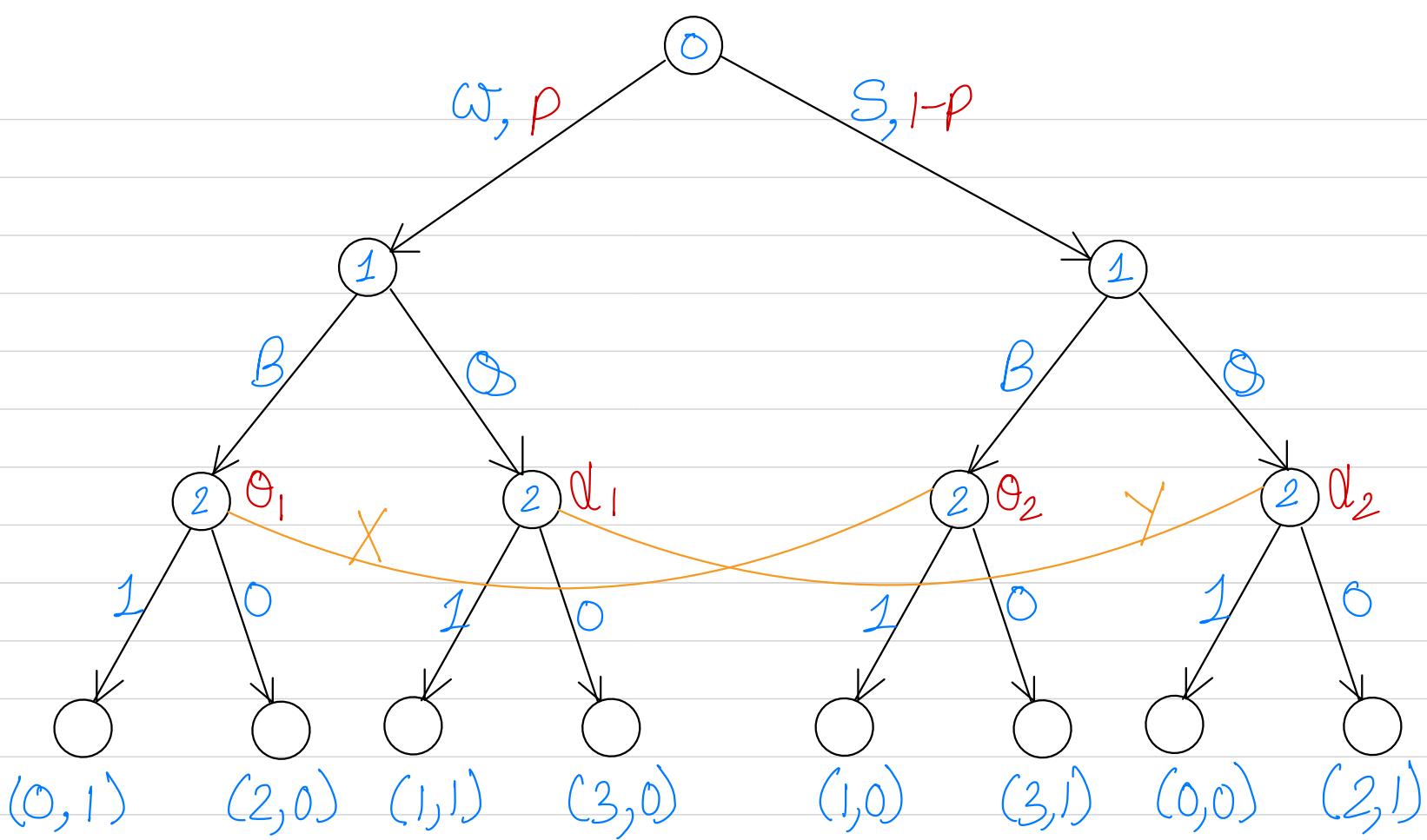
**YouTube video:**

<https://youtu.be/XzOW04VuZJ4>

The video says that it is part 2 of lecture 42. This is because I recorded it along with lecture 42. But the topic covered in this video were the topics that we discussed in lecture 30. I have recorded the video because the lecture slides are not sufficient.

Computing Pure Strategy

Sequential Equilibrium



$$E[U_2^X(1)] = \theta_1$$

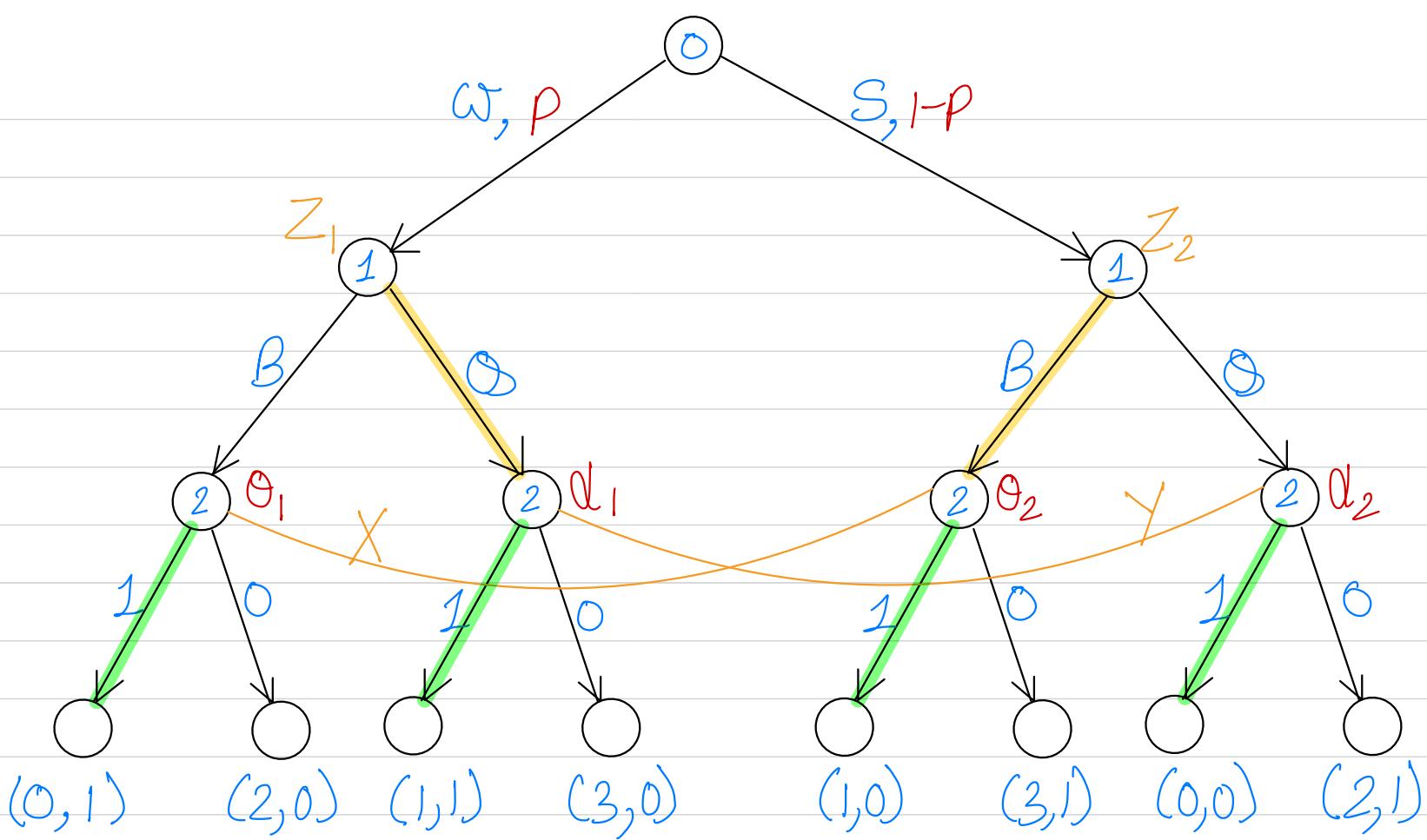
$$E[U_2^X(0)] = \theta_2$$

$$S_X = \begin{cases} 1 ; & \theta_1 > \theta_2 \\ 0 ; & \theta_1 \leq \theta_2 \end{cases}$$

$$E[U_2^Y(1)] = \alpha_1$$

$$E[U_2^Y(0)] = \alpha_2$$

$$S_Y = \begin{cases} 1 ; & \alpha_1 > \alpha_2 \\ 0 ; & \alpha_1 \leq \alpha_2 \end{cases}$$



Case -1

$$S_X = 1; \theta_1 > \theta_2$$

$$S_Y = 1; \alpha_1 > \alpha_2$$

$$S_{Z_1} = \emptyset$$

$$S_{Z_2} = B$$

$$\alpha_1 = \frac{P \cdot 1}{P \cdot 1 + (\neg P) \cdot 0} = 1$$

$$\alpha_2 = 0$$

$$\theta_2 = \frac{(\neg P) \cdot 1}{(1-P) \cdot 1 + P \cdot 0} = 1$$

$$\theta_1 = 0$$

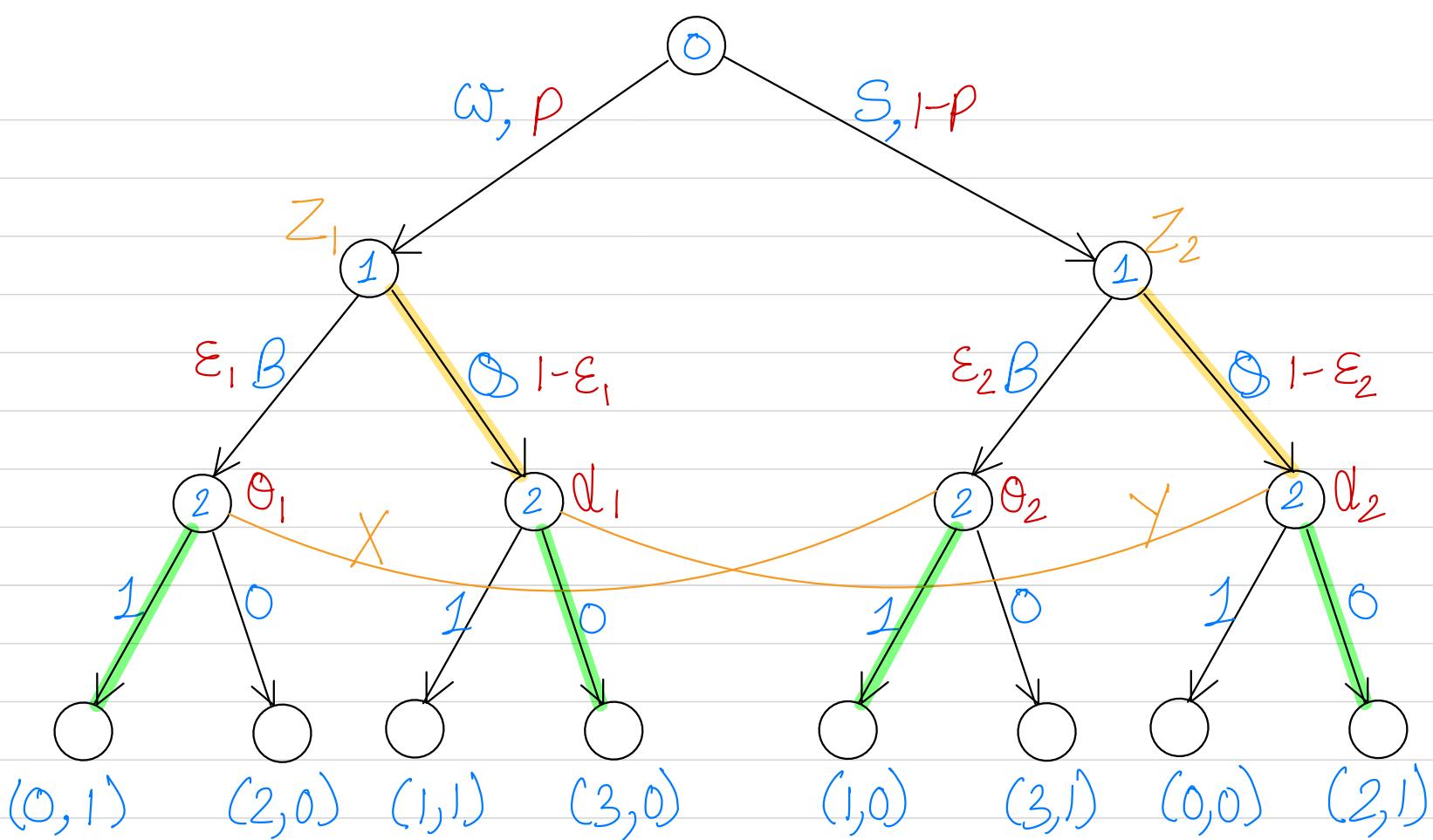
IS  $\theta_1 > \theta_2$ ? NO.

IS  $\alpha_1 > \alpha_2$ ? YES.

Following SE is NOT possible

$$(\emptyset, B, 1, 1, (0, 1), (1, 0))$$

$\xleftarrow{\text{strategy}}$   $\xleftarrow{\text{Belief}}$



Case - 2

$$S_X = 1 ; \theta_1 > \theta_2$$

$$S_Y = 0 ; d_1 \leq d_2$$

$$S_{Z_1} = \emptyset$$

$$S_{Z_2} = \emptyset$$

$$d_1 = \frac{P \cdot 1}{P \cdot 1 + (1-P) \cdot 1} = P$$

$$d_2 = 1 - P$$

$$\theta_1 = \frac{P \cdot \varepsilon_1}{P \cdot \varepsilon_1 + (1-P) \cdot \varepsilon_2} \in [0, 1]$$

$$\theta_2 \in [0, 1]$$

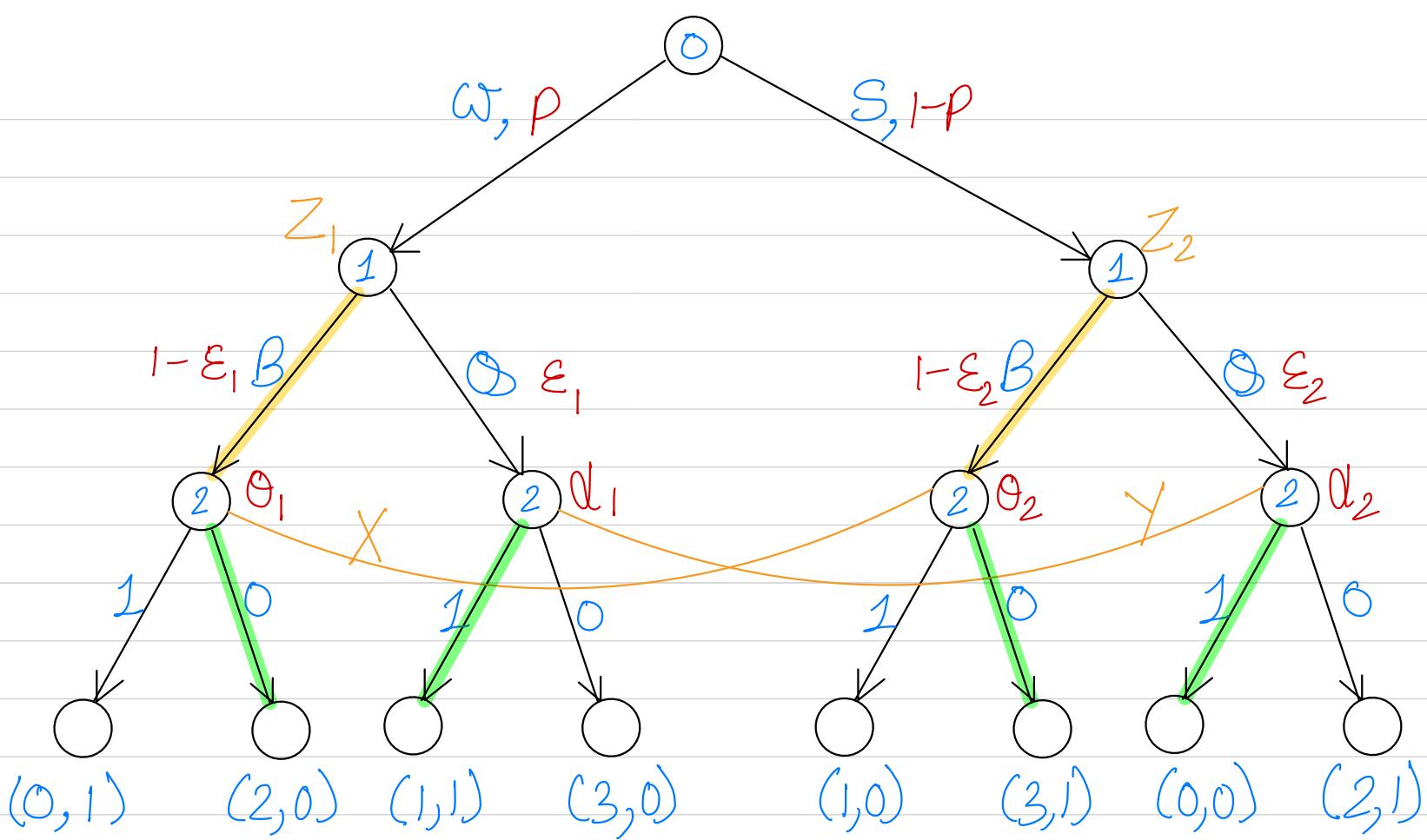
$$\theta_1 + \theta_2 = 1$$

Is  $\theta_1 > \theta_2$ ?  $\theta_1 > 0.5$

Is  $d_1 \leq d_2$ ?  $P \leq 1-P$   
 $P \leq 0.5$

Following SE is possible if  
 $P \leq 0.5$ :

$(\theta_1, \theta_2, 1, 0, (\theta_1, 1 - \theta_1), (P, 1 - P))$   
for all  $\theta_1 > 0.5$ .



Case-3

$$S_X = 0; \theta_1 \leq \theta_2$$

$$S_Y = 1; \alpha_1 > \alpha_2$$

$$S_{Z_1} = B$$

$$\alpha_1 = \frac{P \cdot \epsilon_1}{P \cdot \epsilon_1 + (1-P) \cdot \epsilon_2} \in [0, 1]$$

$$S_{Z_2} = B$$

$$\alpha_2 \in [0, 1] \quad \alpha_1 + \alpha_2 = 1$$

$$\theta_1 = \frac{P \cdot 1}{P \cdot 1 + (1-P) \cdot 1} = P$$

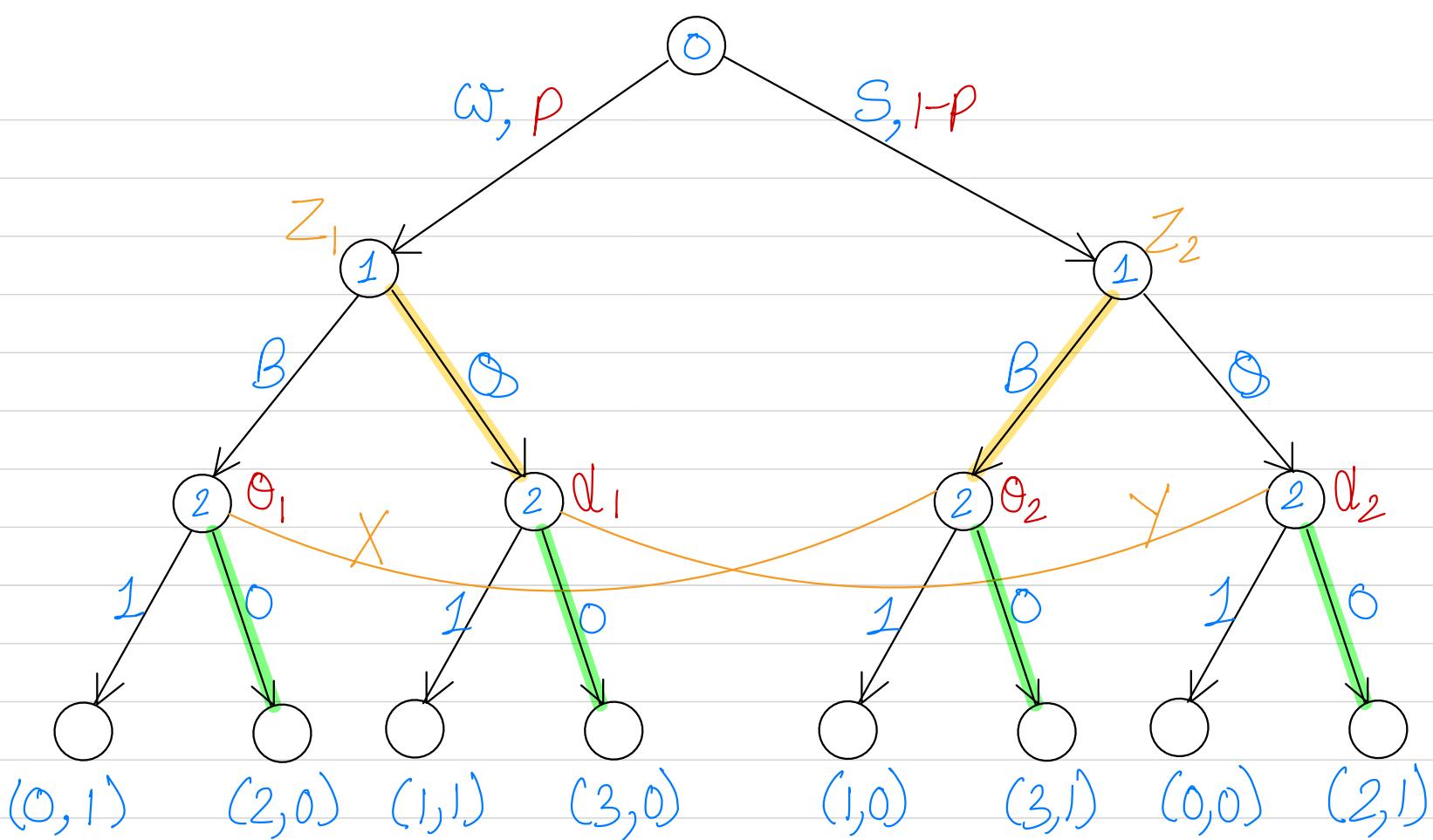
$$\theta_2 = 1 - P$$

Is  $\theta_1 \leq \theta_2$ ?  $P \leq 1-P$   
 $P \leq 0.5$

Is  $\alpha_1 > \alpha_2$ ?  $\alpha_1 > 0.5$

Following SE is possible if  
 $P \leq 0.5$ :

$(B, B, 0, 1, (P, 1-P), (\alpha_1, 1-\alpha_1))$   
 for all  $\alpha_1 > 0.5$ .



Case - 4

$$S_X = 0; \theta_1 \leq \theta_2$$

$$S_Y = 0; d_1 \leq d_2$$

$$S_{Z_1} = \emptyset$$

$$S_{Z_2} = B$$

$$d_1 = \frac{P \cdot 1}{P \cdot 1 + (1-P) \cdot 0} = 1$$

$$d_2 = 0$$

$$\theta_2 = \frac{(1-P) \cdot 1}{(1-P) \cdot 1 + P \cdot 0} = 1$$

$$\theta_1 = 0$$

➢ IS  $\theta_1 \leq \theta_2$ ? YES.

➢ Is  $d_1 \leq d_2$ ? NO.

Following SE is NOT possible

$(\emptyset, B, 0, 0, (0,1), (1,0))$

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# Game Theory (CS4187)

## Lectures 42

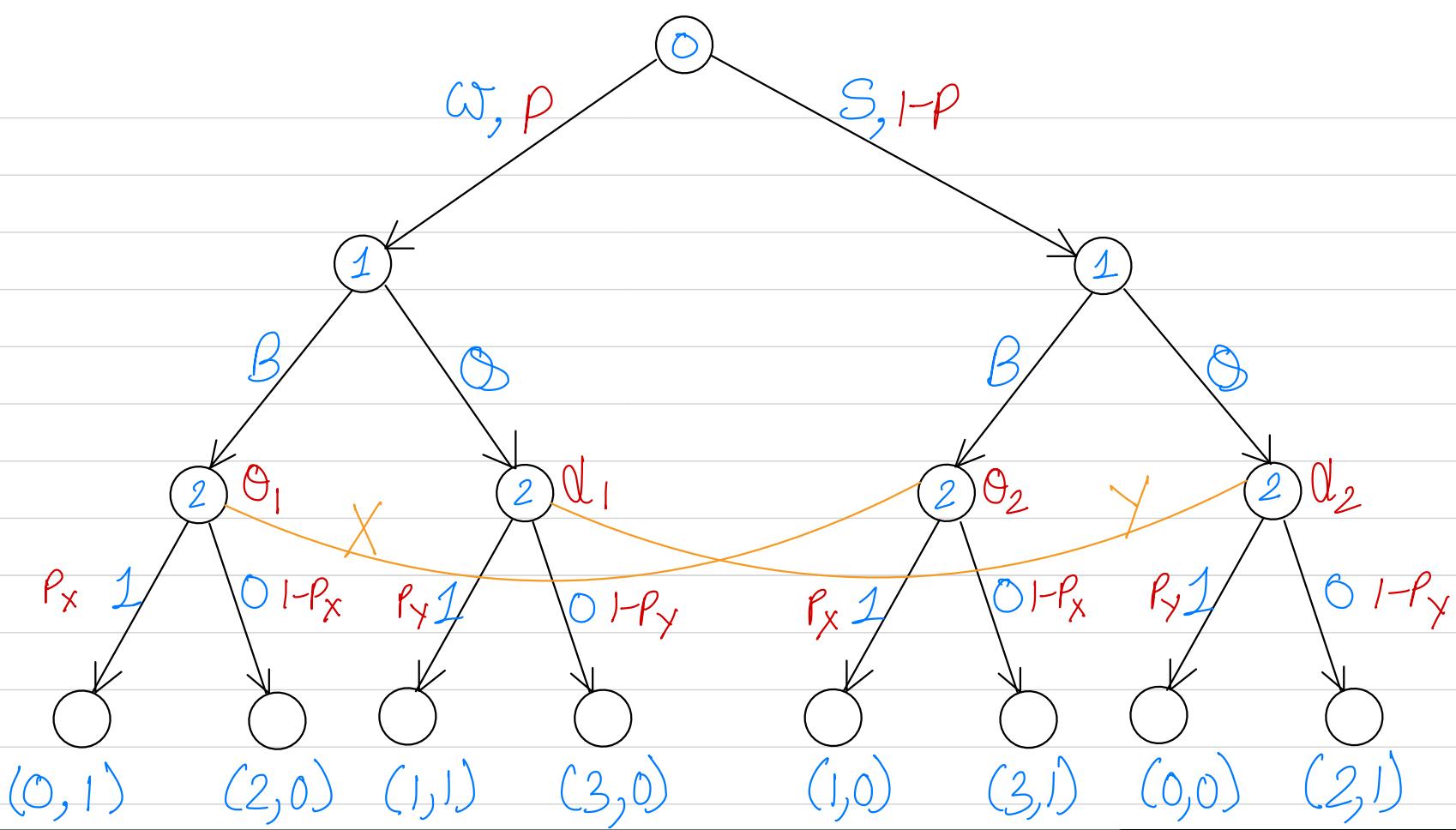
**YouTube video:**

<https://youtu.be/xR3C82laibg>

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Computing Mixed Strategy

Sequential Equilibrium

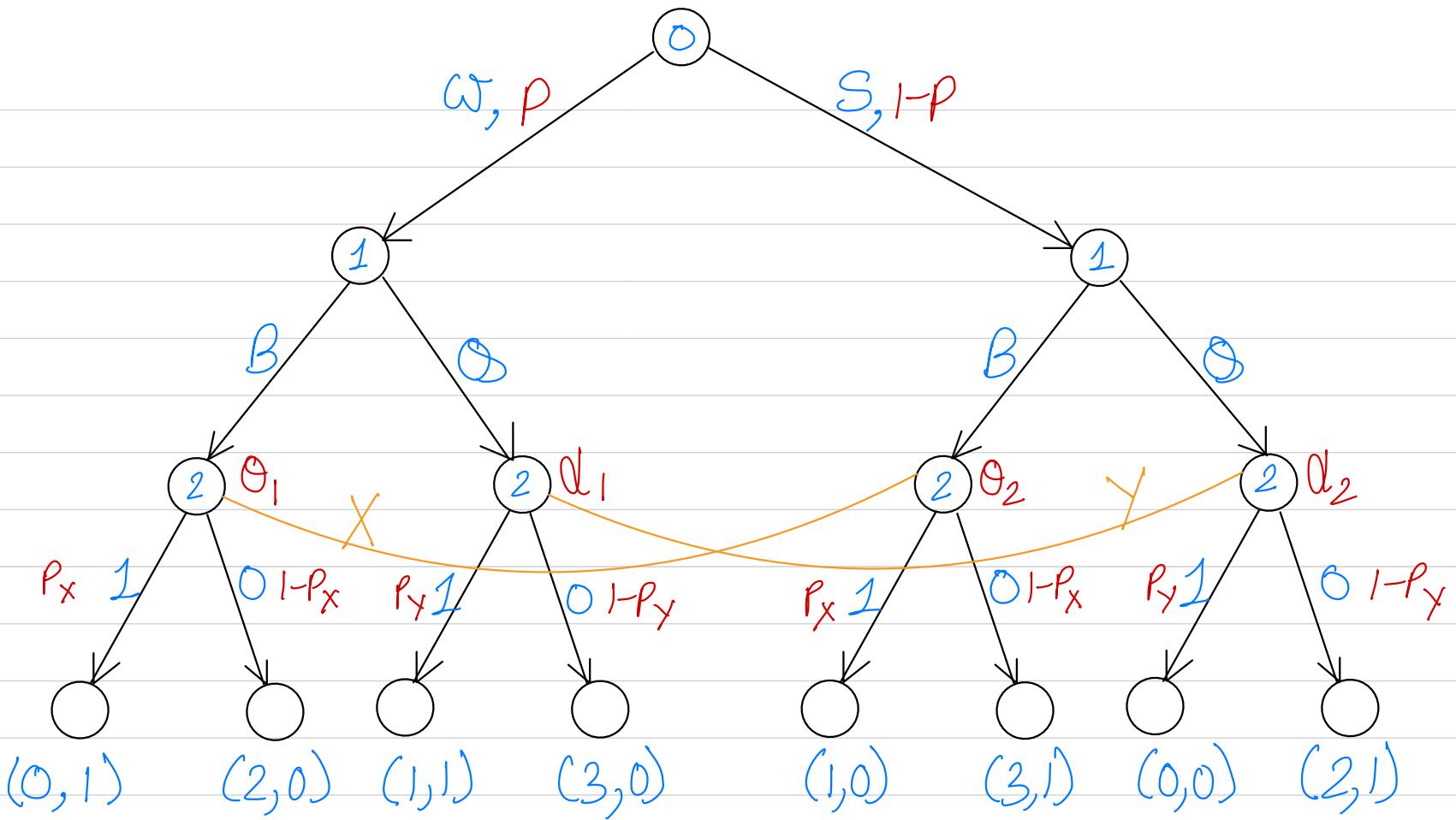


$$E[U_2^X((P_X(1), 1 - P_X(0)))] = \theta_1 \cdot (P_X \cdot 1 + (1 - P_X) \cdot 0) + \theta_2 \cdot (P_X \cdot 0 + (1 - P_X) \cdot 1) = \theta_1 P_X + \theta_2 (1 - P_X)$$

$P_X$ : Prob. that player 2 will play 1 in information set X.

$$E[U_2^Y((P_Y(1), 1 - P_Y(0)))] = \alpha_1 \cdot (P_Y \cdot 1 + (1 - P_Y) \cdot 0) + \alpha_2 \cdot (P_Y \cdot 0 + (1 - P_Y) \cdot 1) = \alpha_1 P_Y + \alpha_2 (1 - P_Y)$$

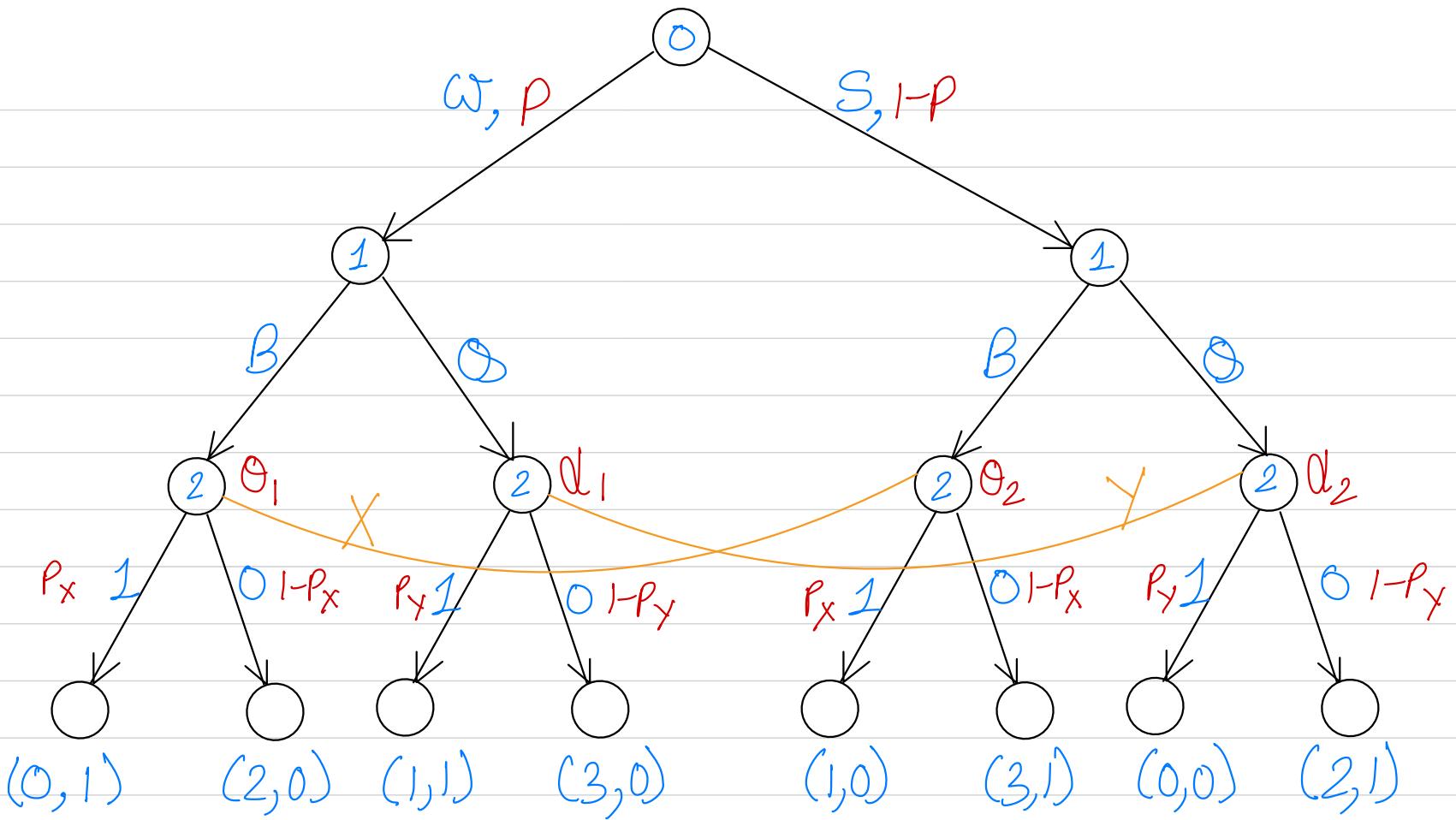
$P_Y$ : Prob. that player 2 will play 1 in information set Y.



$$P_X^* = \underset{P_X}{\operatorname{argmax}} E[U_2^X((P_X(1), 1-P_X(0)))]$$

$$\underset{P_X}{\operatorname{argmax}} \theta_1 P_X + \theta_2 (1-P_X) = \begin{cases} 1 & ; \theta_1 > \theta_2 \\ 0 & ; \theta_1 < \theta_2 \\ P_X & ; \theta_1 = \theta_2 \end{cases} = \begin{cases} 1 & ; \theta_1 > 0.5 \\ 0 & ; \theta_1 < 0.5 \\ P_X & ; \theta_1 = 0.5 \end{cases}$$

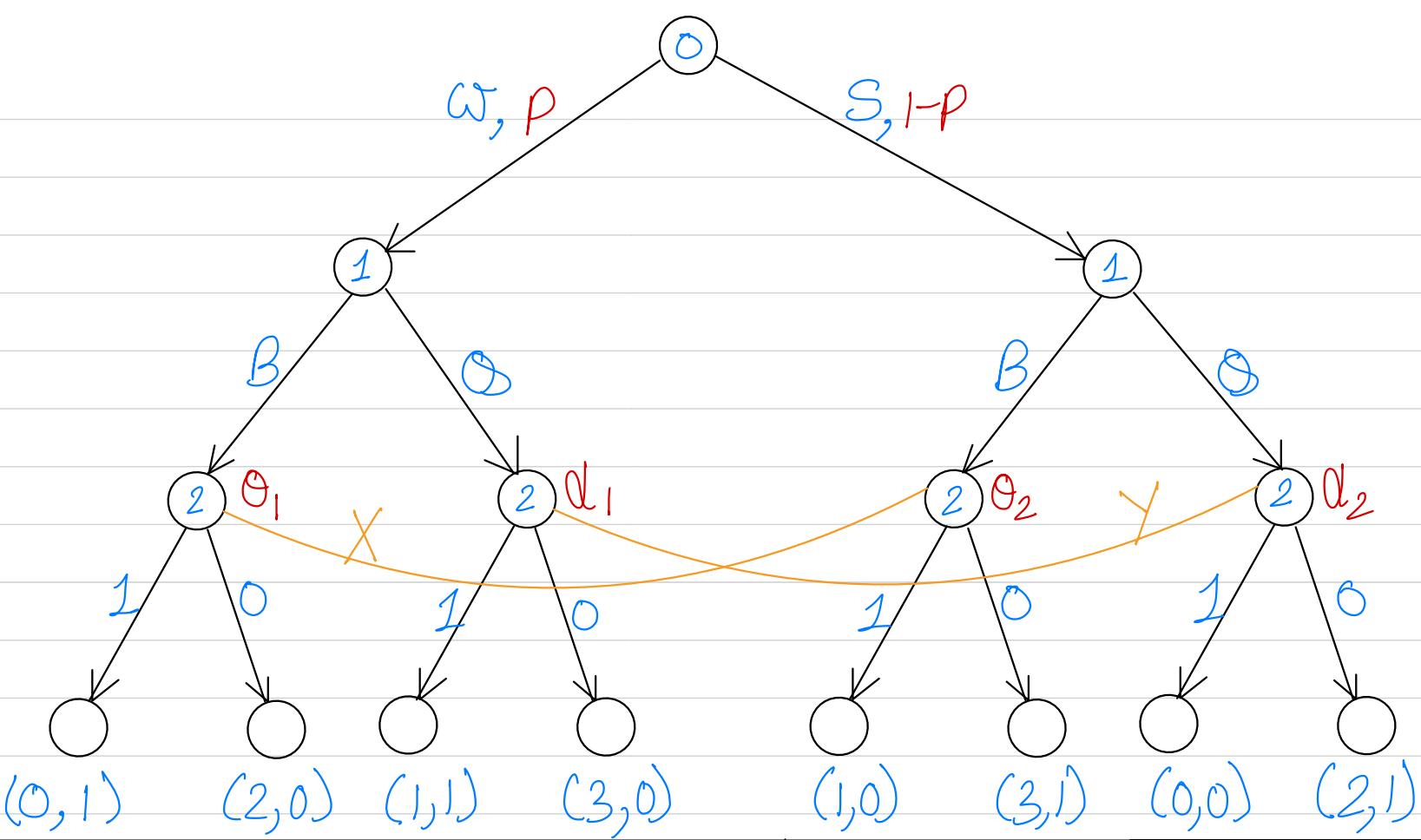
Indifference Same  
as def. 6 of MSNE.



$$P_Y^* = \underset{P_Y}{\operatorname{argmax}} E[U_2^Y((P_Y(1), 1-P_Y(0)))]$$

$$\underset{P_Y}{\operatorname{argmax}} \theta_1 P_Y + \theta_2 (1-P_Y) = \begin{cases} 1 & ; \alpha_1 > \alpha_2 \\ 0 & ; \alpha_1 < \alpha_2 \\ P_Y & ; \alpha_1 = \alpha_2 \end{cases} = \begin{cases} 1 & ; \alpha_1 > 0.5 \\ 0 & ; \alpha_1 < 0.5 \\ P_Y & ; \alpha_1 = 0.5 \end{cases}$$

Indifference Same  
as def. 6 of MSNE.

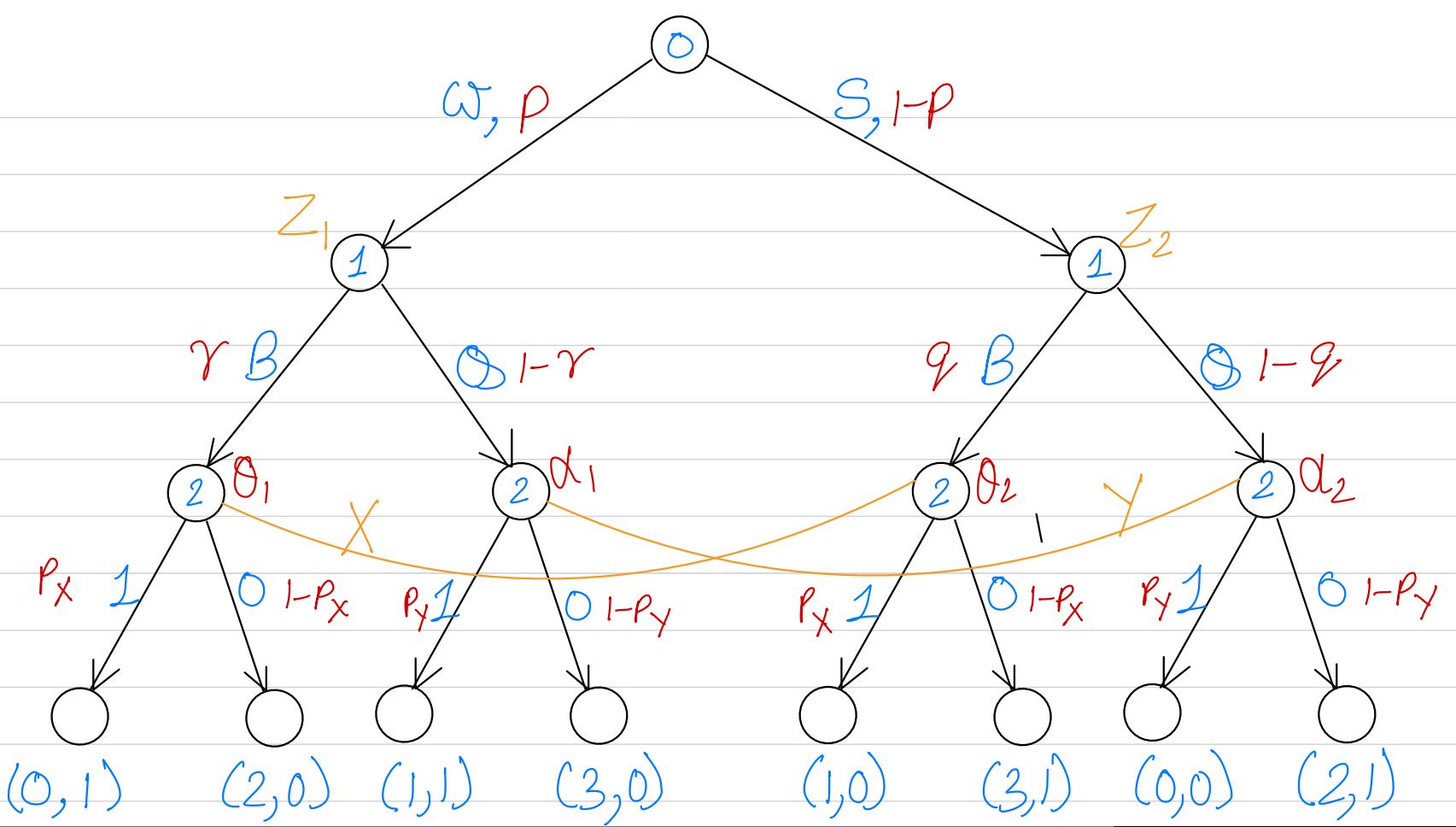


$$S_X = \begin{cases} 1 & ; \theta_1 > 0.5 \\ 0 & ; \theta_1 < 0.5 \end{cases}$$

$(P_X(1), 1-P_X(0)) ; \theta_1 = 0.5$

$$S_Y = \begin{cases} 1 & ; \alpha_1 > 0.5 \\ 0 & ; \alpha_1 < 0.5 \end{cases}$$

$(P_Y(1), 1-P_Y(0)) ; \alpha_1 = 0.5$



Case - 5

$$S_X = (P_X(1), 1 - P_X(0))$$

$$\theta_1 = \theta_2 = 0.5$$

$$S_Y = (P_Y(1), 1 - P_Y(0))$$

$$\alpha_1 = \alpha_2 = 0.5$$

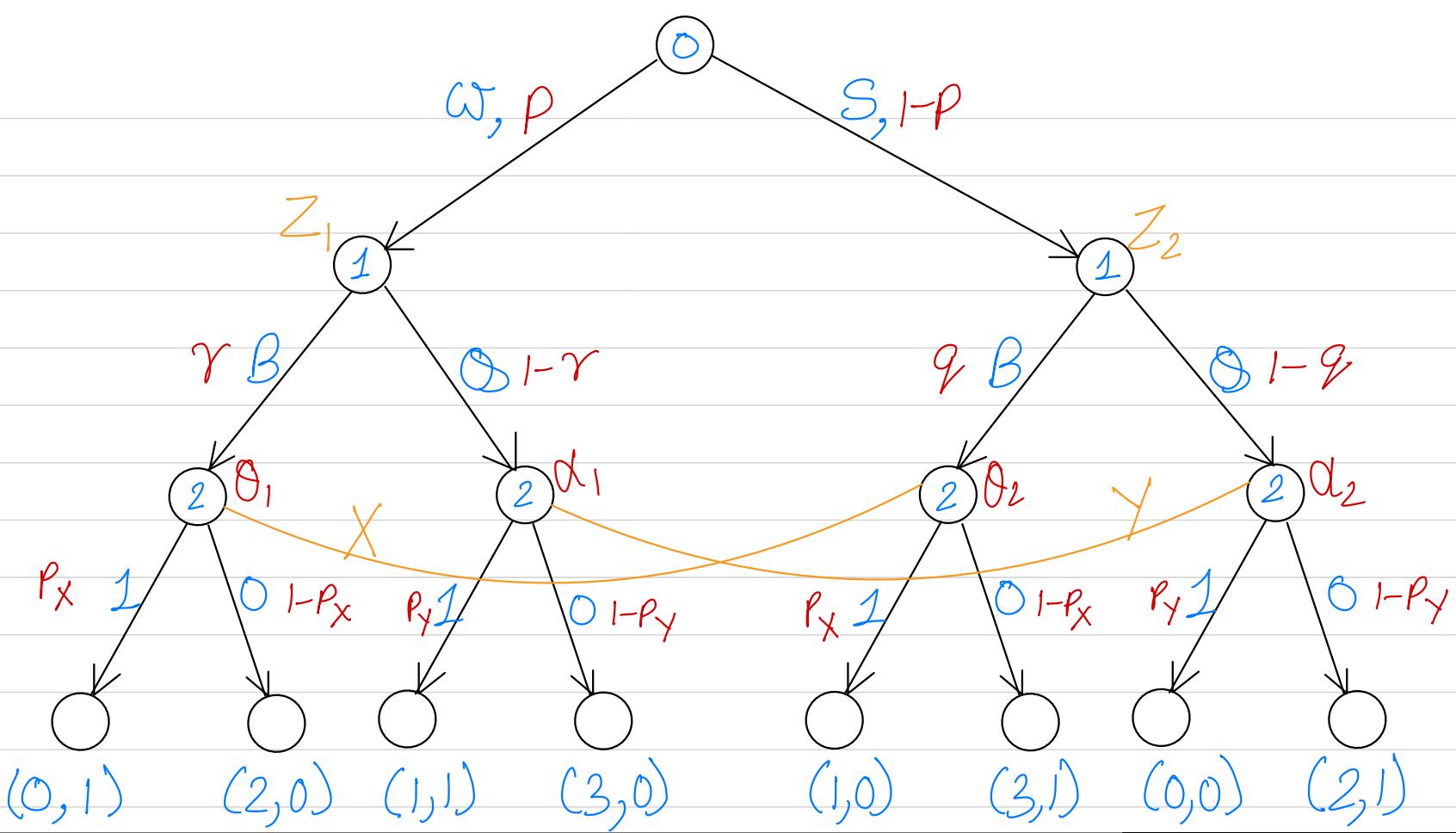
$$E\{U_1^{Z_1}((r(B), 1-r(S))]\}$$

$$= r \cdot P_X \cdot 0 + r \cdot (1-P_X) \cdot 2$$

$$+ (1-r) \cdot P_Y \cdot 1 + (1-r) \cdot (1-P_Y) \cdot 3$$

$$= 2r(1-P_X) + (1-r)(3-2P_Y)$$

$$S_{Z_1} = \begin{cases} B & ; 2(1-P_X) > 3-2P_Y \\ \emptyset & ; 2(1-P_X) \leq 3-2P_Y \\ (r(B), 1-r(S)) ; 2(1-P_X) = 3-2P_Y \end{cases}$$



Case - 5

$$S_X = (P_X(1), 1 - P_X(0))$$

$$\theta_1 = \theta_2 = 0.5$$

$$S_Y = (P_Y(1), 1 - P_Y(0))$$

$$\alpha_1 = \alpha_2 = 0.5$$

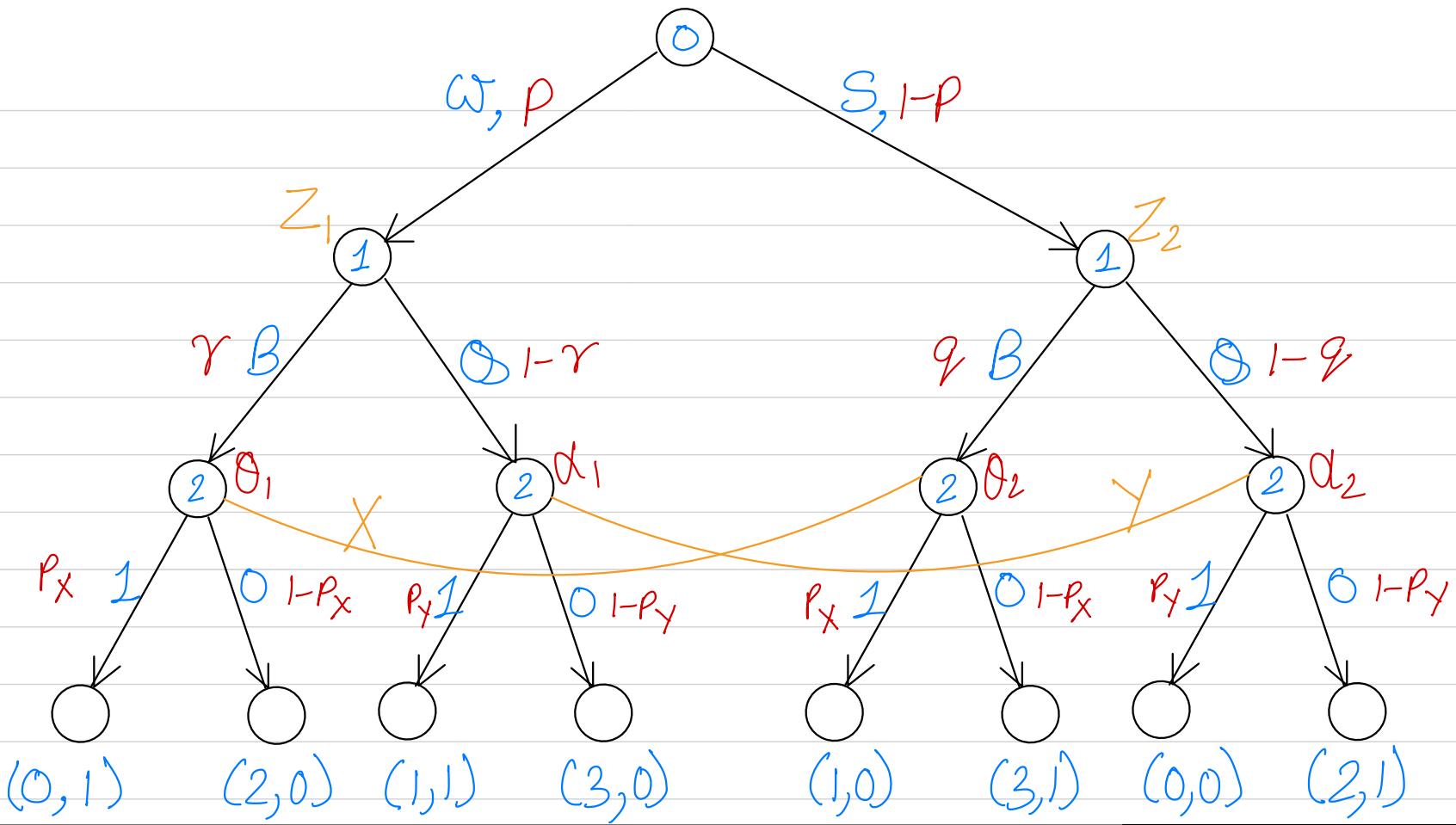
$$E\left[U_{\gamma}^{Z_2}((\varphi(B), 1-\varphi(\bar{B}))\right]$$

$$= \varphi \cdot P_X \cdot 1 + \varphi \cdot (1 - P_X) \cdot 3$$

$$+ (1 - \varphi) \cdot P_Y \cdot 0 + (1 - \varphi) \cdot (1 - P_Y) \cdot 2$$

$$= \varphi \cdot (3 - 2P_X) + 2 \cdot (1 - \varphi) \cdot (1 - P_Y)$$

$$S_{Z_2} = \begin{cases} B & ; (3 - 2P_X) > 2(1 - P_Y) \\ \emptyset & ; (3 - 2P_X) \leq 2(1 - P_Y) \\ (\varphi(B), 1 - \varphi(\bar{B})); (3 - 2P_X) = 2(1 - P_Y) \end{cases}$$



Case - 5

$$\theta_1 = \frac{p\gamma}{p\gamma + (1-p)q} = 0.5 \Rightarrow p\gamma = (1-p)q \quad \text{--- (1)}$$

$$S_X = (p_X(1), 1-p_X(0))$$

$$\theta_1 = \theta_2 = 0.5$$

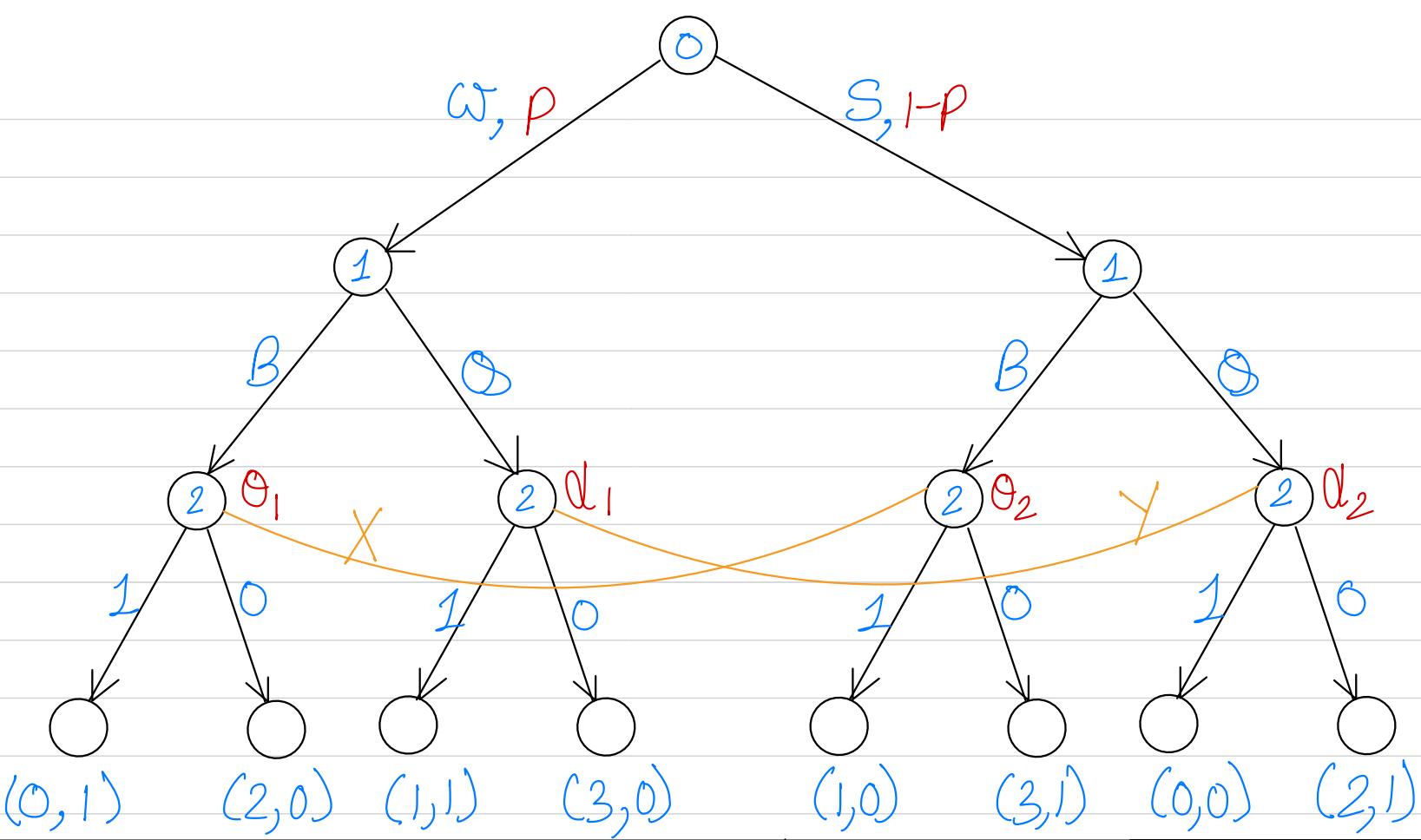
$$S_Y = (p_Y(1), 1-p_Y(0))$$

$$d_1 = d_2 = 0.5$$

$$d_1 = \frac{p(1-\gamma)}{p(1-\gamma) + (1-p)(1-q)} = 0.5 \Rightarrow p(1-\gamma) = (1-p)(1-q) \quad \text{--- (2)}$$

Substitute  $\gamma = \frac{1-p}{p}q$  from (1) in (2)  
we get,  $p=0.5$ .

Shortcut!

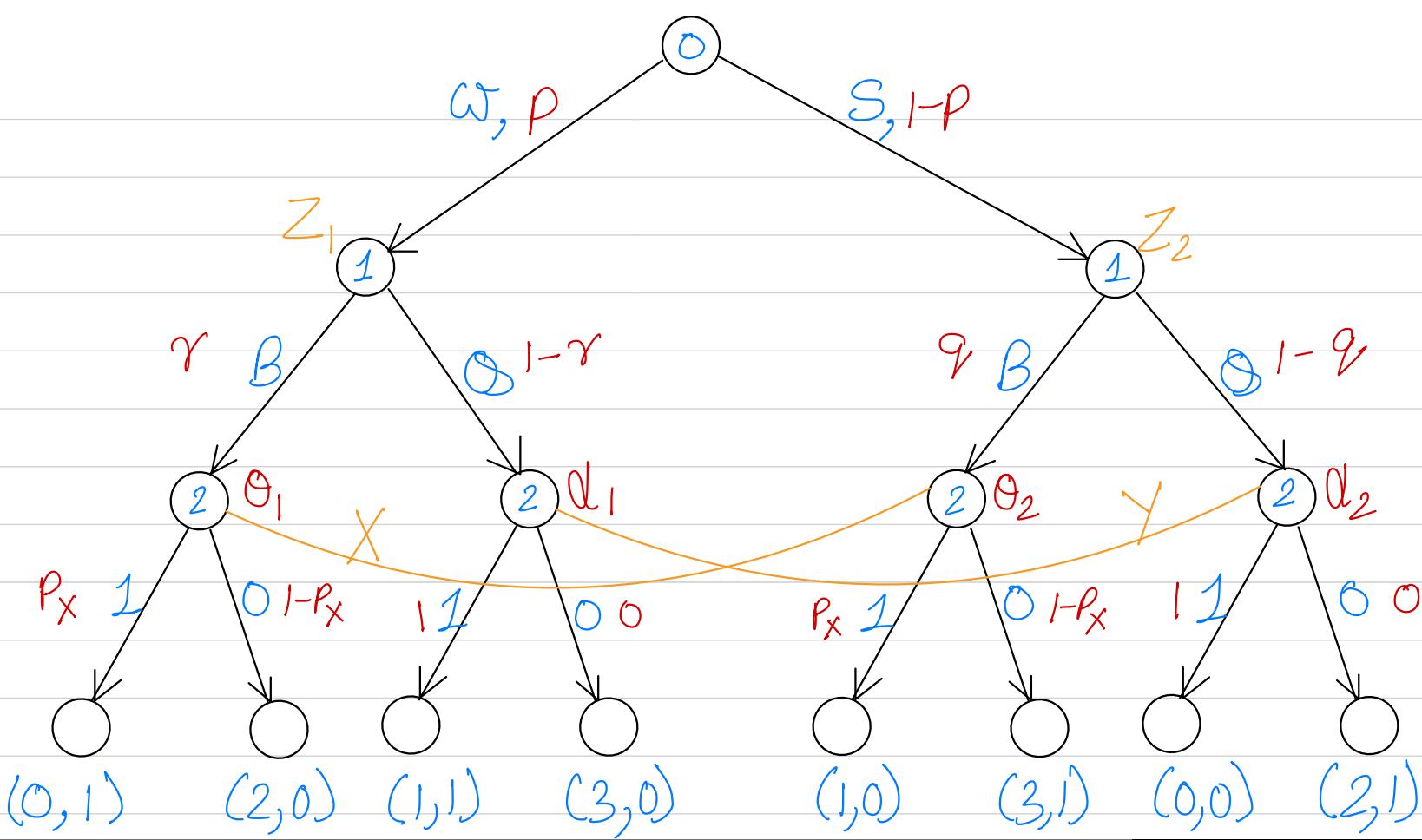


$$S_X = \begin{cases} 1 & ; \theta_1 > 0.5 \\ 0 & ; \theta_1 < 0.5 \end{cases}$$

$(P_X(1), 1-P_X(0)) ; \theta_1 = 0.5$

$$S_Y = \begin{cases} 1 & ; \alpha_1 > 0.5 \\ 0 & ; \alpha_1 < 0.5 \end{cases}$$

$(P_Y(1), 1-P_Y(0)) ; \alpha_1 = 0.5$



Case - 6

$$S_X = (P_X(1), 1 - P_X(0))$$

$$\theta_1 = \theta_2 = 0.5$$

$$S_Y = (1(1), 0(0))$$

$$\begin{aligned} &= 1 \\ &\alpha_1 > 0.5 \end{aligned}$$

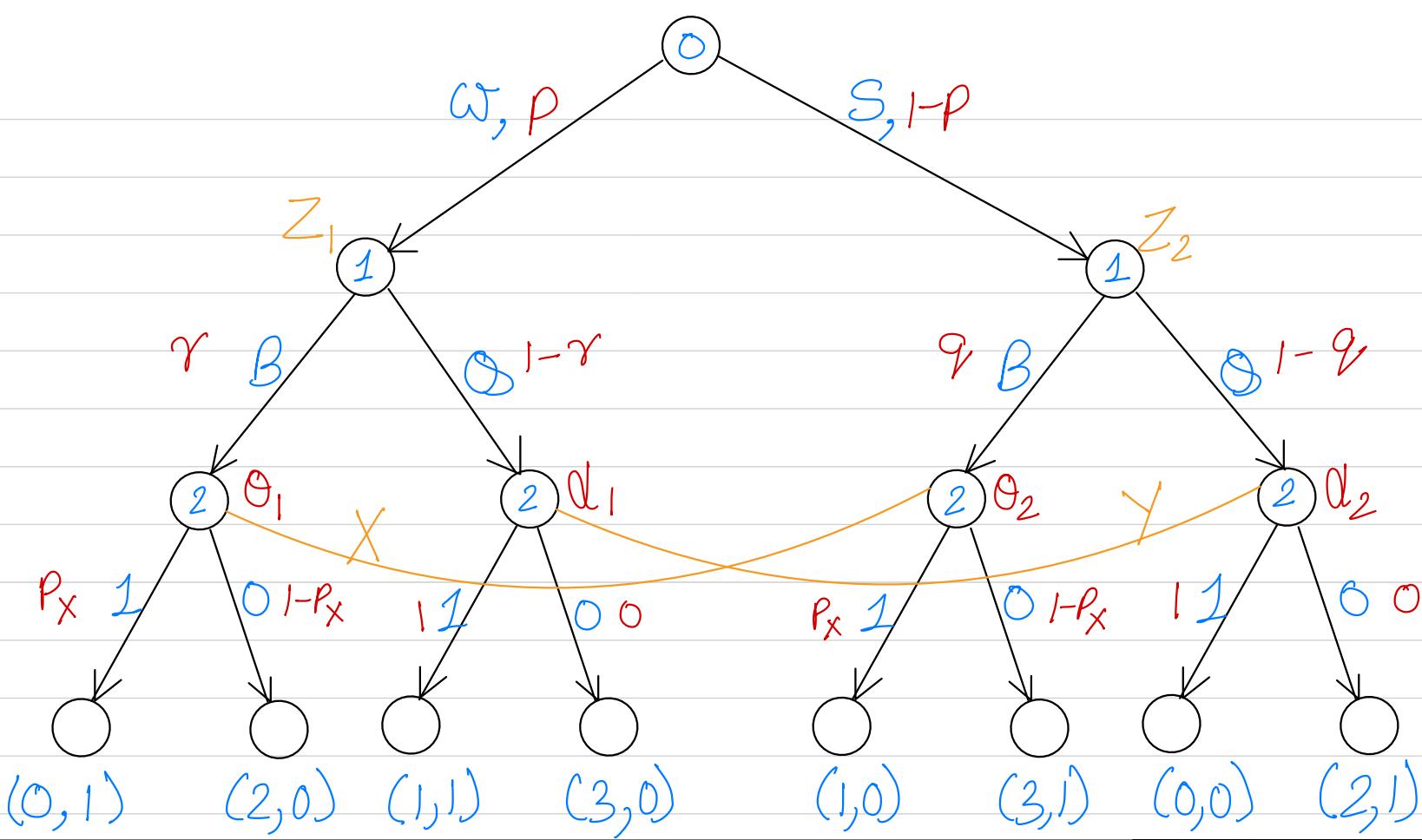
$$E[U_{\gamma}^{\bar{F}_1}((r(B), 1-r(Q)))]$$

$$= r \cdot P_X \cdot 0 + r \cdot (1 - P_X) \cdot 2$$

$$+ (1-r) \cdot 1 \cdot 1 + (1-r) \cdot 0 \cdot 3$$

$$= 2r(1 - P_X) + (1 - r) \cdot 1$$

$$S_{Z_1} = \begin{cases} B & ; 2(1 - P_X) > 1 \\ Q & ; 2(1 - P_X) < 1 \\ (r(B), 1 - r(Q)) & ; 2(1 - P_X) = 1 \end{cases}$$



Case - 6

$$E[U_i^{Z_2}((\gamma(B), 1-\gamma(\emptyset)))]$$

$$S_X = (P_X(1), 1-P_X(0)) = \gamma \cdot P_X \cdot 1 + \gamma \cdot (1-P_X) \cdot 3$$

$$S_{Z_2} = B$$

$$\theta_1 = \theta_2 = 0.5$$

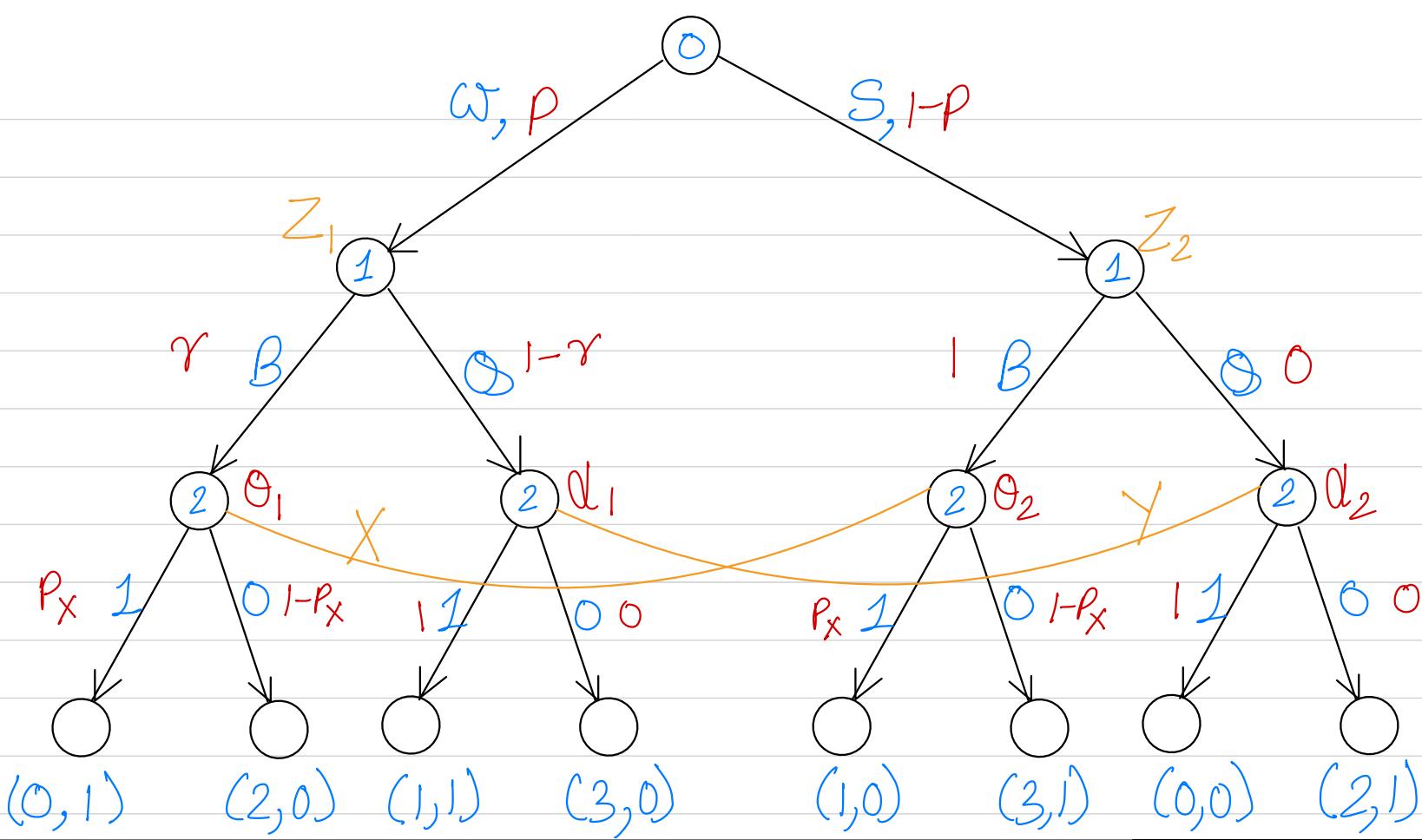
$$+ (1-\gamma) \cdot 1 \cdot 0 + (1-\gamma) \cdot 0 \cdot 2$$

$$S_Y = (1(1), 0(0))$$

$$= 1$$

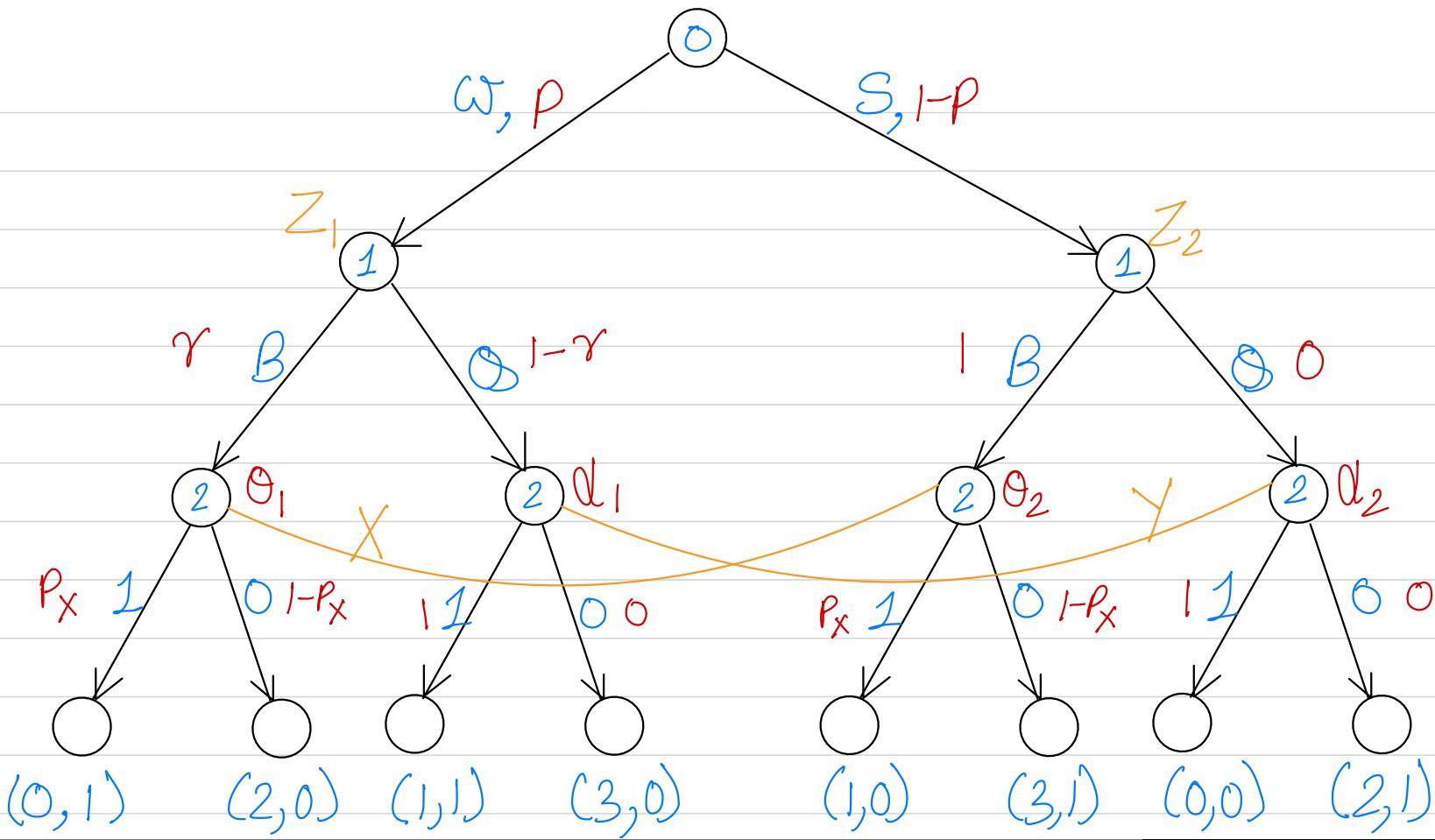
$$\alpha_1 > 0.5$$

$$= 3\gamma$$



<u>Case - 6</u> $S_X = (P_x(1), 1-P_x(0))$ $\theta_1 = \theta_2 = 0.5$ $S_Y = (1(1), 0(0))$ $\alpha_1 = 1$ $\alpha_1 > 0.5$	$S_{Z_1} = \begin{cases} B & ; 2(1-P_x) > 1 \\ 0 & ; 2(1-P_x) < 1 \end{cases}$ $(\gamma(B), 1-\gamma(0)) ; 2(1-P_x) = 1$	<u>Case - A (<math>S_{Z_1} = B</math>)</u> $\Rightarrow \gamma = 1$ $P_x < 0.5$ $\theta_1 = \frac{P \cdot 1}{P \cdot 1 + (1-P) \cdot 1} = P$ $\alpha_1 = \frac{P \cdot \varepsilon_1}{P \cdot \varepsilon_1 + (1-P) \cdot \varepsilon_2} \in [0, 1]$
--------------------------------------------------------------------------------------------------------------------------------------------	-----------------------------------------------------------------------------------------------------------------------------	---------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------

IS  $\theta_1 = 0.5$ ? IF  $P = 0.5$   
IS  $\alpha_1 > 0.5$ ? Yes. Possible.



Case - 6

$$S_X = (P_x(1), 1 - P_x(0))$$

$$\theta_1 = \theta_2 = 0.5$$

$$S_Y = (1(1), 0(0))$$

$$\begin{aligned} &= 1 \\ \alpha_1 &> 0.5 \end{aligned}$$

$$S_{Z_1} = \begin{cases} B & ; 2(1 - P_x) > 1 \\ \emptyset & ; 2(1 - P_x) < 1 \end{cases}$$

$$(r(B), 1 - r(\emptyset)); 2(1 - P_x) = 1$$

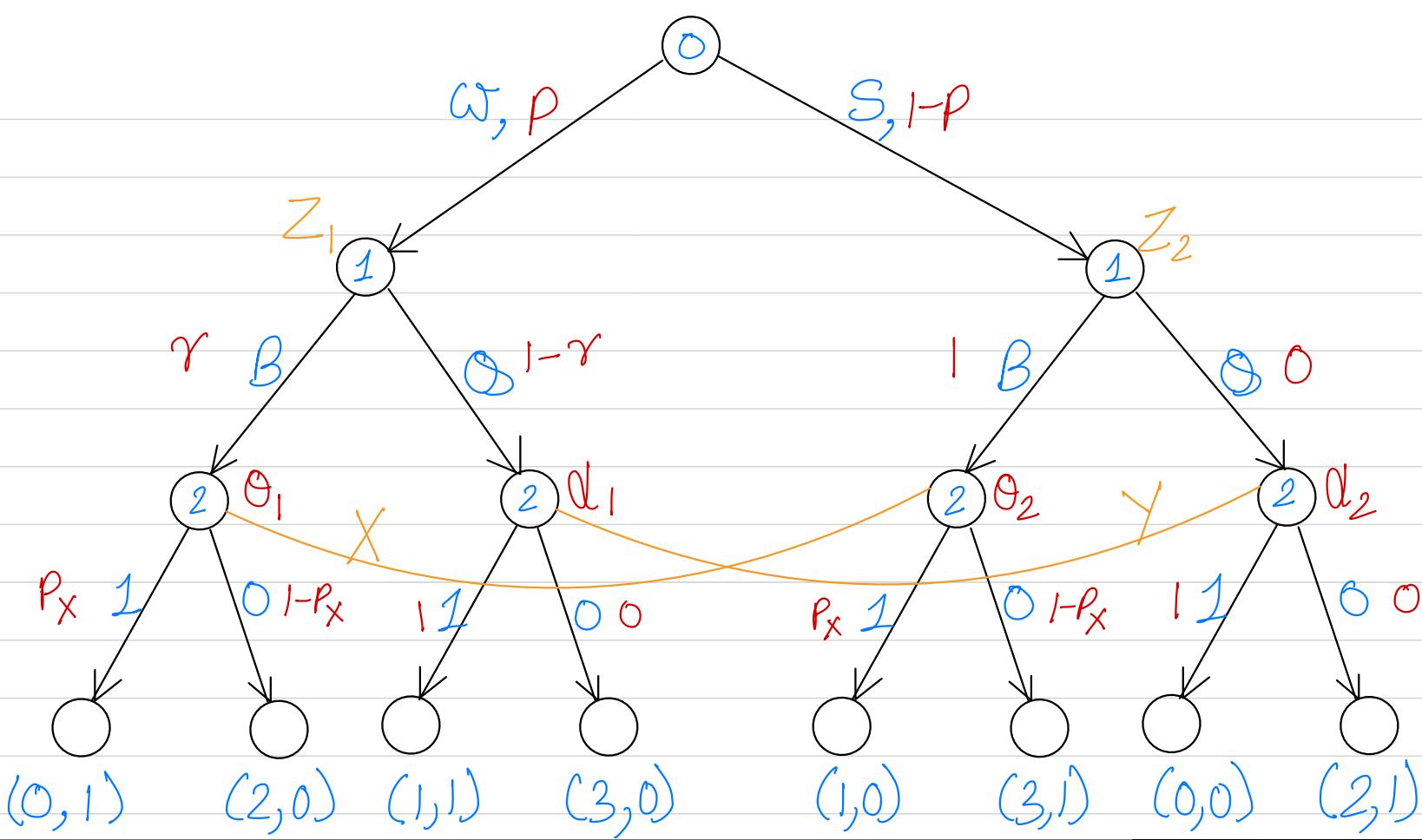
Case - B ( $S_{Z_1} = \emptyset$ )  $\Rightarrow \gamma = 0$   
 $P_x > 0.5$

$$\theta_1 = \frac{P \cdot 0}{P \cdot 0 + (1 - P) \cdot 1} = 0$$

$$\alpha_1 = \frac{P \cdot 1}{P \cdot 1 + (1 - P) \cdot 0} = 1$$

IS  $\theta_1 = 0.5$ ? NO!

IS  $\alpha_1 > 0.5$ ? Yes.



Case - 6

$$S_X = (P_X(1), 1-P_X(0))$$

$$\theta_1 = \theta_2 = 0.5$$

$$S_Y = (1(1), 0(0))$$

$$= 1$$

$$\alpha_1 > 0.5$$

$$S_{Z_1} = \begin{cases} B & ; 2(1-P_X) > 1 \\ S & ; 2(1-P_X) < 1 \\ (r(B), 1-r(S)) & ; 2(1-P_X) = 1 \end{cases}$$

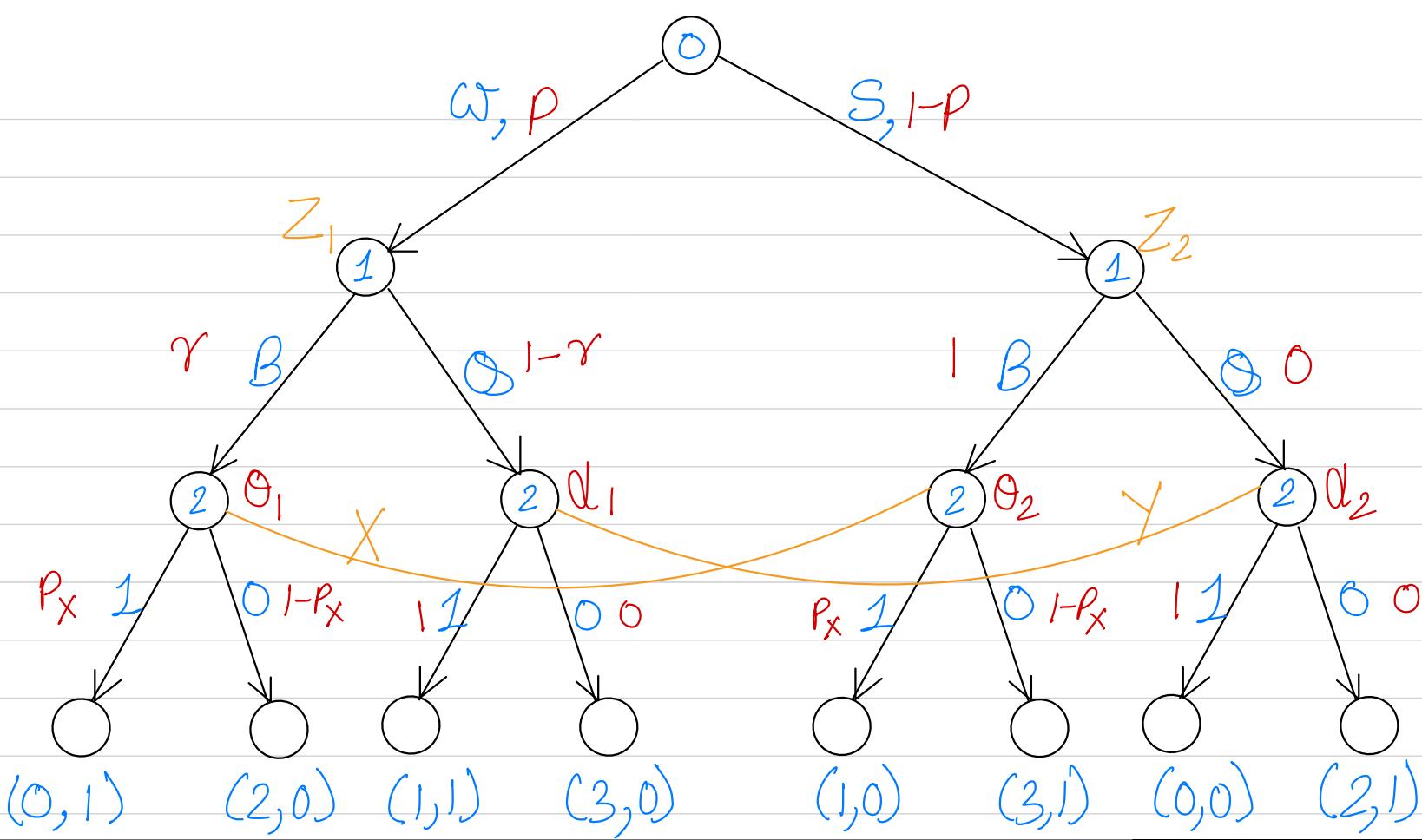
Case-C ( $S_{Z_1} = S$ )  $P_X = 0.5$

$$\theta_1 = \frac{P \cdot r}{P \cdot r + (1-P) \cdot 1}$$

$$\alpha_1 = \frac{P \cdot (1-r)}{P \cdot (1-r) + (1-P) \cdot 0} = 1$$

IS  $\theta_1 = 0.5$ ? Set to 0.5.  $r = \frac{1-P}{P}$

IS  $\alpha_1 > 0.5$ ? Yes.



Case - 6

$$S_X = (P_X(1), 1 - P_X(0))$$

$$\theta_1 = \theta_2 = 0.5$$

$$S_Y = (I(1), 0(0))$$

$$\alpha_1 = 1 \\ \alpha_1 > 0.5$$

Following SE is possible for  $P \geq 0.5$ :

$$\left( \left( \frac{1-P}{P}(B), \frac{2P-1}{P}(0) \right), B, (0.5(1), 0.5(0)), I, (1, 1), (1, 1), (0.5, 0.5), (1, 0) \right)$$

$S$

$P$



# Thank You!