

## Evaluation Function

How good is this state

→ What is it?

An evaluation function is used to estimate the value or "goodness" of a game state when the search hasn't reached the end of the game.

In simple words:

"If the game stopped right now, how likely am I to win from this position?"

Why is it needed?

Searching all the way to the end of the game takes a lot of time (especially in games like chess). So we cut off the search early, and use an evaluation function to estimate how good that state is.

## Example (Tic Tac Toe)

You might give score like this:

- +10 if the AI wins
- -10 if the opponent wins
- 0 if it's a draw
- otherwise, count how many 2-in-a-rows the AI has.

## Cutting-off search

When do we stop looking further?

What is it?

Instead of going all the way to the terminal state (like win/loss), we cut off the search at a certain depth or time.

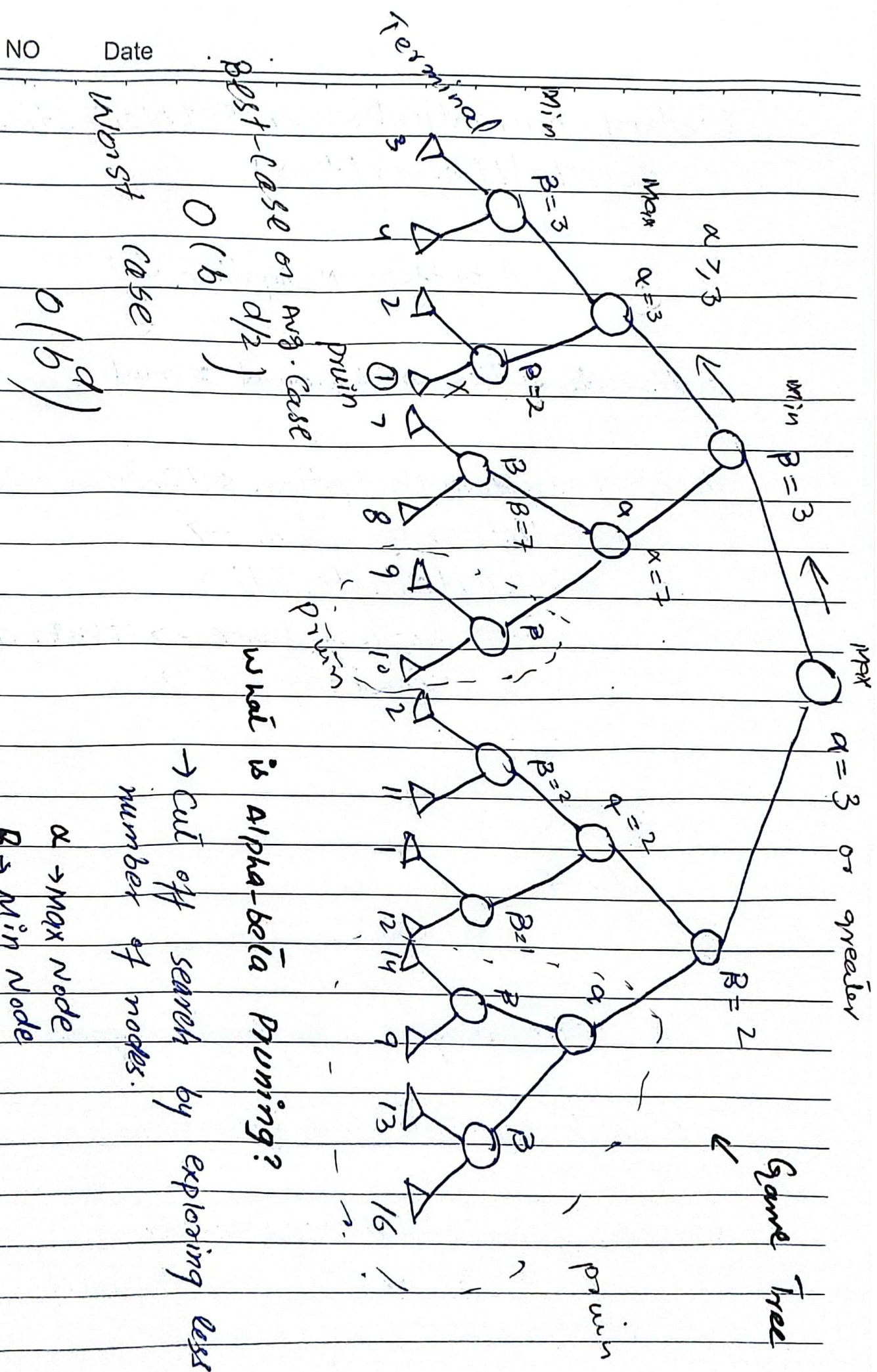
In simple words:

"Stop search after 4 moves ahead.  
Evaluate the state using evaluation function."

Why do we cut off?

Deep search is slow and expensive.  
So, we stop at a certain depth level or time limit and then evaluate the state.





What is Alpha-Beta Pruning?

→ Cut off search by exploring less number of nodes.

$\alpha \rightarrow$  Max Node  
 $\beta \rightarrow$  Min Node

# Why Alpha-Beta Pruning is Effective?

## Benefit

- Cut-off bad branches
- Save computation
- Can search deeper
- Same result as full minimax
- $O(b^d)$  to  $O(b^{d/2})$

## Result

- Fewer nodes to evaluate
- Runs faster
- Better game decisions
- But much more efficient.
- Reduce time complexity