

# Considering Uncertainty in Chess Engine Evaluation

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## 1 Abstract

We should prefer a position which we are certain is +2 to a position which we think is +3 but we are not sure. What we should aim to maximise in a chess game is not the evaluation with respect to our side, but our expected score from the game. I define the expected score as the number of points (0 for loss, 0.5 for draw, 1 for win) which we predict to obtain from this game.

Considering our uncertainty in the evaluation is important because the expected score from a game does not increase linearly with evaluation. For instance, the difference between the expected score from a position which is +0.5 and a position which is -0.5 is very significant, while the difference in expected score from a position which is +9 and one which is +10 is tiny. Therefore, the expected score from a position which is +2 ( $\pm 0.2$ ) to a position which is +3 ( $\pm 2.5$ ).

## 2 Implementation

The aim is to create a function  $W(p)$  which is inputted certain features of a position  $p$ , and outputs a weight  $w \in (0, \infty)$ .

### 2.1 Obtaining Data

Suggested in this paper are two methods of obtaining data as to the uncertainty of various positions.

#### 2.1.1 Method 1: WDL prediction

Positions in which the static evaluation is a poor predictor of the result of the game are high in uncertainty.

#### 2.1.2 Method 2: Low / High Search Depth

Positions in which the evaluation returned from a low depth search is very different to the evaluation returned from a high depth search are high in uncertainty.

## 2.2 Loss Function

Let  $WDL(x)$  be a function which maps the evaluation of a position to its expected score. This function must:

- be strictly increasing
- be S-shaped
- have range  $[0, 1]$

Let  $S(p)$  be the static evaluation of a position  $p$ .

Let  $H(p)$  be the evaluation returned by a high depth search of  $p$ .

To compute the loss of a function  $W(p)$  with method 2:

```

 $T \leftarrow 0$ 
for all positions in dataset do
   $a \leftarrow H(p)$ 
   $w \leftarrow W(p)$ 
   $b \leftarrow S(p) \cdot w$ 
   $\Delta \leftarrow WDL(b) - WDL(a)$ 
   $T = T + \Delta$ 
end for
return  $T$ 

```

If method 1 is used instead then  $H(p)$  should be replaced with the eventual result of the game which the position came from.

## 2.3 Creating $W(p)$

I will first experiment with a handcrafted approach to implementing  $W(p)$ , tuned using the loss function above. Some features which should be inputted are:

- king safety of both sides
- total material of both sides
- mobility of pieces (trapped pieces can lead to big misevaluations)
- whether or not the position is check