

## 2.2 Game tree evaluation: randomized algorithms

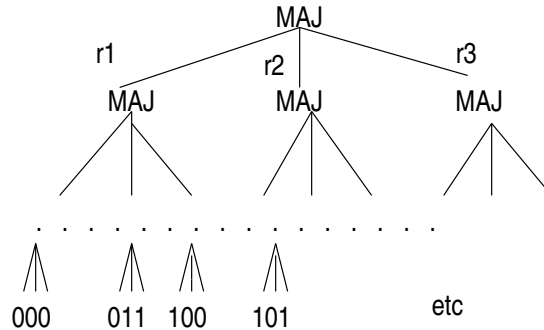
Let us now focus on game trees with a uniform structure. The most canonical example is alternating levels of binary AND and OR gates. This can be simplified with the identity  $\text{OR}(x, y) = \neg\text{AND}(\neg x, \neg y)$ . So we can reduce to the problem of evaluating a complete binary tree of NAND gates. (In the conversion, it may be necessary to flip the input gates or the output gate or both; but this does not affect the query complexity of the problem.)

NAND	0	1
0	1	1
1	1	0

The key to savings in the randomized case is that when either input to a NAND gate is found to be 0, the output is bound to be 1 and therefore we don't need to look at the other input. This suggests a *Random DFS* evaluation—randomly, uniformly pick which child to recurse on—should be efficient. Of course, the potential savings make sense for deterministic evaluation too, but if we go in a prescribed order, the adversary will always make us see a 1 first. (This is exactly what we exploited in the lower bound in the previous section.) A randomized algorithm has the advantage that if either child equals 0, we have half probability of looking at that child first and avoiding the other evaluation.

### Simpler question: MAJ3 trees of depth $n$

There's a small technical challenge in analyzing the NAND tree so let's look at an easier question: The complete ternary tree of depth  $n$  in which every non-leaf vertex is a Majority gate. (A singleton node is depth 0.) Let  $S_n =$  worst-case expected number of leaves evaluated by Random DFS.



Whatever the values of the three children of a node, we only have to evaluate the last child if the first two disagreed. At least one of the three pairs is always in agreement, so the probability we first find a disagreeing pair is at most  $2/3$ . So

$$S_n \leq (2/3) \cdot 3 \cdot S_{n-1} + (1/3) \cdot 2 \cdot S_{n-1} = (8/3)S_{n-1}$$

Hence (keeping in mind  $S_0 = 1$ ),  $S_n \leq (8/3)^n = N^{\log_3 8/3} \cong N^{0.893}$  where  $N = 3^n$  is the number of leaves.

What we have here is very much like a branching process. Every node has either two or three children (until level  $n$  at which the branching process is cut off), and the savings in the randomized algorithm comes from showing that, in the notation of the next section,  $\bar{\mu} \leq 8/3$ .

**Digression: branching processes**

Let  $\mu$  be a probability distribution on the nonnegative integers. The *branching process* or *Galton-Watson process* with distribution  $\mu$  is the following tree-valued random variable  $T$ :

$T$  has a root. Each vertex  $v$  of  $T$  gets some  $N_v$  children, for  $N_v$  independently distributed according to  $\mu$ . Let  $\bar{\mu} = E(N_v)$  (possibly infinite).

**Theorem 6.** *TFAE:*

1.  $T$  is a.s. finite. (a.s. = Almost Surely = With Probability 1)
2.  $\bar{\mu} \leq 1$  and  $\mu_1 < 1$ .

The intuition is this. In the subcritical regime, i.e.,  $\bar{\mu} < 1$ , each parent has less than 1 child on average—so no wonder the generations die out with probability 1. In the supercritical regime,  $\bar{\mu} > 1$ , things are not so definite—it could happen for example that the root has no children at all—but, the number of vertices at a level is generally drifting upwards, and as it grows, the likelihood of population collapse decreases drastically, so overall, the probability of the tree being infinite is positive. The critical case  $\bar{\mu} = 1$  is (as always in these kinds of problems) hardest to determine and here we have two cases. One is that  $\mu_1 = 1$  in which case the process is deterministic,  $T$  is infinite and there is nothing more to say. The other is that  $\mu_1 < 1$  which implies that  $\mu_0 > 0$ . Now the number of children at each level of the tree is just drifting without bias up or down. However, there is an absorbing boundary at 0: extinction is forever. This process is not a random walk with bounded step size, such as we have studied in “gambler’s ruin”, but intuitively it behaves similarly, and it goes extinct with probability 1 for basically the same reason.

Here we prove formally only a weaker version of (2)  $\Rightarrow$  (1); after that we’ll give some idea of the other direction but point to Grimmett and Stirzaker §5.4 Thm (5) for a full proof [18].

If  $\bar{\mu} < 1$  then  $T$  is a.s. finite.

*Proof.* We will use the shorthand  $\mu_{\geq i} = \sum_{j \geq i} \mu_j$ .

When a vertex has  $N$  children, we list them in an arbitrary “birth order” as children  $1, \dots, N$ . The “address” of a vertex of the tree is a finite string of positive integers  $(X_1 \dots X_\ell)$ : the root is represented by the empty string and the address of a vertex is its parent’s address followed by its place in the birth order.

For a string  $(X_1 \dots X_\ell)$ , let  $\llbracket X_1 \dots X_\ell \rrbracket$  be the (indicator rv of) the event that this address exists in the tree. An equivalent characterization of this event is that

1. The root has at least  $X_1$  children, and
2. The vertex  $(X_1)$  has at least  $X_2$  children, and ...
3. The vertex  $(X_1 \dots X_{\ell-1})$  has at least  $X_\ell$  children.

Note that  $\bar{\mu} = \sum_{i \geq 0} i\mu_i = \sum_{i \geq 1} \mu_{\geq i}$ .

We have the following:

$$\Pr(\llbracket X_1 \dots X_\ell \rrbracket) = \prod_{j=1}^{\ell} \mu_{\geq X_j}$$

The event that  $T$  is infinite is equivalent to the event that  $\sum_{\vec{X}} \llbracket \vec{X} \rrbracket = \infty$ . Let’s calculate the expectation of the LHS. Note that in the following calculation all products are of finitely many terms.

$$\begin{aligned}
 E(\sum_{\vec{X}} \mathbb{I}[\vec{X}]) &= \sum_{\vec{X}} \Pr(\mathbb{I}[\vec{X}]) = \sum_{\ell \geq 0} \sum_{\vec{X}: |\vec{X}| = \ell} \prod_{j=1}^{\ell} \mu_{\geq X_j} \\
 &= \sum_{\ell \geq 0} \prod_{j=1}^{\ell} \sum_{X \geq 1} \mu_{\geq X} \\
 &= \sum_{\ell \geq 0} \prod_{j=1}^{\ell} \bar{\mu} \\
 &= \sum_{\ell \geq 0} \bar{\mu}^{\ell} \\
 &= \frac{1}{1 - \bar{\mu}} < \infty \quad \text{here we finally use } \bar{\mu} < 1
 \end{aligned}$$

Now let's recall the first Borel-Cantelli lemma (??) (*Reminder: Let  $B_i$  be countably many events s.t.  $\sum_{i \geq 1} \Pr(B_i) < \infty$ . Then  $\Pr(\limsup B) = 0$ .)*) It follows that  $T$  is almost surely finite.  $\square$