# **Analysis: Catch Them All**

## **Catch Them All: Analysis**

#### **Small dataset**

We can start by computing the shortest distance between each pair of locations using the <u>Floyd-Warshall</u> algorithm. We will use dis[i, j] to represent the shortest distance between locations i and j.

Let dp[K, L] be the expected time needed to catch K Codejamons when starting from location L. Then we can use a dynamic programming algorithm with the following state transition equation:

```
if (K == 0):

dp[K, L] = 0;

else:

dp[K, L] = \Sigma_{i!=L}(dp[K-1, i] + dis[L, i]) / (N-1).
```

The algorithm above takes  $O(N^2P)$  time, which is fast enough to solve the Small dataset.

### Large dataset

We can find that for each dp[K, L], the answer is a linear expression of dp[K-1, i] when K = 0. So, we can rewrite the state transition equation as the product of a matrix and a column vector, as shown below.

```
Let S[i] = \sum_{i!=i} (dis[i, j]).
```

Let  $F_K$  denote the column vector of dp[K, i], and let A denote the transition matrix. Then we have  $F_K = A * F_{K-1} = A^K * F_0$ .

With the approach above, we can use <u>exponentiation by squaring</u> to accelerate the computation of  $A^K$ . This gives us an  $O(N^3 \log P)$  algorithm which can solve the Large dataset.

#### Other solutions

Let  $Pr_t$  be the probability of being at location 1 after catching t Codejamons. Initially,  $Pr_0$  = 1. Since at any time, the probabilities of being at locations 2, 3, ..., **N** (let's call them the "other locations") are the same, we can calculate the probability of the next Codejamon appearing at location 1 by multiplying the probability of being at any of the other locations by the probability of location 1 being chosen. Therefore, we have  $Pr_t = (1 - Pr_{t-1}) / (N-1)$ .

After computing the values of  $Pr_i$  for i = 1, 2, ..., P-1, the answer to the problem is:

 $\Sigma(Pr_i^* (expected distance from location 1) + \Sigma(1 - Pr_i) / (N-1)^* (expected distance from location j) for j = 2, 3, ..., N) for i = 0, 1, ...,$ **P**-1.

Note: the expected distance from location i equals  $\Sigma$ dis[i, j] / (N-1) for j = 1, 2, ..., N.

We might not have enough time to compute all  $\mathbf{P}$  values of the sequence  $\Pr_i$ , but one may notice that this sequence converges quickly (except for  $\mathbf{N}=2$ , which we can handle separately). Intuitively, as the game

progresses, the probabilities of you being at each of the **N** locations become more equal. For example, if **N**=4, the first few values for  $Pr_t$  are 1, 0, 0.333, 0.222, 0.259, 0.247, 0.251, 0.250, ... Once this value gets very close to 1 / **N**, after some threshold like i = 100 (depending on the numerical error allowed), we can simply approximate  $Pr_i = 1/N$  for all i larger than the threshold. Then i = 100, 101, ..., **P**-1 in the previous summation can be replaced by a multiplication: (**P**-100) \*  $\Sigma$ (expected distance from location j) / **N** for j = 1, 2, ..., **N**.

The time complexity for the above algorithm is  $O(N^3 + C)$ , where C depends on the numerical error allowed.

Using the sequence of  $Pr_t$  values, it is also possible to calculate the exact answer. Let's make another sequence  $A_i = Pr_i - 1/N$ . This sequence converges to 0 and is a geometric progression because:

- $Pr_i = (1 Pr_{i-1}) / (N-1)$
- $A_i + 1/N = (1 A_{i-1} 1/N) / (N-1)$
- $A_i + 1/N = -A_{i-1} / (N-1) + 1/N$
- $A_i = -A_{i-1} / (N-1)$ .

 $\Sigma A_i$  for i = 0, 1, ..., P-1 can be calculated using the formula for the sum of a <u>geometric series</u>. With  $\Sigma A_i$ , we can obtain  $\Sigma Pr_i = 1/N * P + \Sigma A_i$ , and ultimately the answer. The time complexity for the above algorithm is  $O(N^3)$ .