## Learning Goals

- Basic definitions of finite probabilities: sample space, probability, events
- State and apply union bound.
- Define independence, and apply its properties in probability calculations
- Contention resolution with random access, and analysis of its efficiency


## Borges's Garden of Forking Paths



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- A probability space is defined by weights on those realizations.


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- If everything is fair, then each outcome has probability mass $1 / 36$.
- Let $\mathcal{E}$ be the event that the sum of the two numbers is 11 , then $\mathcal{E}=\{(6,5),(5,6)\}$, so $\operatorname{Pr}[\mathcal{E}]=1 / 18$.


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Exercise: If $A$ and $B$ are independent, then so are $\bar{A}$ and $B$, and so are $\bar{A}$ and $\bar{B}$.

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- Trivial if the tasks can agree on some ordering and requests the service one by one.
- Problem: The tasks cannot talk with each other and there is no central authority.
- Randomized strategy: In each time step, each task requests with some small probability $p$, independently.


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- Let $S[i, t]$ denote the event that task $i$ sends a request at time $t$ and gets served, then

$$
\operatorname{Pr}[S[i, t]]=\operatorname{Pr}\left[A[i, t] \cap \bigcap_{j \neq i} \overline{A[j, t]}\right]=p(1-p)^{n-1}
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- To maximize $\operatorname{Pr}[S[i, t]$, set $p=1 / n$.


## Rate of success at each time step

We set $p$ to maximize $\operatorname{Pr}[S[i, t]]$ to $\frac{1}{n}\left(1-\frac{1}{n}\right)^{n-1}$. How good is this?

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## Proposition

(1) The function $\left(1-\frac{1}{n}\right)^{n}$ converges monotonically from $\frac{1}{4}$ up to $\frac{1}{e}$ as $n$ increases from 2.
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So $1 /(e n) \leq \operatorname{Pr}[S[i, t]] \leq 1 /(2 n)$. Therefore $\operatorname{Pr}[S[i, t]]$ is asymtotically $\Theta(1 / n)$.

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- Remark: in many situations, the two give answers that are close: sometimes one may show that the random quantity concentrates around its expectation. Tail bounds are used to prove this.
- Probability with which task $i$ does not succeed in the first $t$ steps:

$$
\operatorname{Pr}\left[\cap_{r=1}^{t} \overline{S[i, r]}\right]=\prod_{r=1}^{t}[1-\operatorname{Pr}[S[i, r]]]=\left[1-\frac{1}{n}\left(1-\frac{1}{n}\right)^{n-1}\right]^{t}
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\operatorname{Pr}\left[\cap_{r=1}^{t} \overline{S[i, r]}\right] \leq\left[1-\frac{1}{e n}\right]^{t}=\left[1-\frac{1}{e n}\right]^{e n \cdot \frac{t}{e n}} \leq e^{-t / e n}
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- Big picture (useful rough estimations): if we have a biased coin that gives Heads with probability $1 / k$ :
- In about $k$ independent tosses, one "expects" to see a Heads;
- However, with constant probability, a Heads doesn't show in $k$ tosses;
- But if one tosses the coin $\Theta(k \log k)$ times, the probability that no Heads shows up quickly tends to 0 .


## Waiting time for all tasks to succeed

- Let $F[i, t]$ denote the event that task $i$ fails in the first $t$ steps, we have shown $\operatorname{Pr}[F[i, t]] \leq e^{-t / e n} \leq n^{-c}$ for $t=\lceil e n \cdot c \ln n\rceil$.


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- The event that some task keeps failing in the first $t$ steps is then $\cup_{i=1}^{n} F[i, t]$.
By the union bound, we have

$$
\operatorname{Pr}\left[\cup_{i=1}^{n} F[i, t]\right] \leq \sum_{i=1}^{n} e^{-t / e n}=n e^{-\frac{t}{e n}}
$$

So for $t=\lceil 2 e n \ln n\rceil$, this is at most $\frac{1}{n}$.

