

LECTURE 21: The Bernoulli process

- Definition of Bernoulli process
- Stochastic processes
- Basic properties (memorylessness)
- The time of the k th success/arrival
- Distribution of interarrival times
- Merging and splitting
- Poisson approximation

The Bernoulli process

- A sequence of independent Bernoulli trials, X_i

- At each trial, i :

$$P(X_i = 1) = P(\text{success at the } i\text{th trial}) = p$$

$$P(X_i = 0) = P(\text{failure at the } i\text{th trial}) = 1 - p$$

$$0 < p < 1$$

- Key assumptions:

- Independence
- Time-homogeneity

- Model of:

- Sequence of lottery wins/losses
- Arrivals (each second) to a bank
- Arrivals (at each time slot) to server
- ...



- Jacob Bernoulli
(1655–1705)

(Image is in the public domain.
Source: [Wikipedia](#))

Stochastic processes

infinite

- First view: sequence of random variables X_1, X_2, \dots

Interested in: $\mathbf{E}[X_i] = p$ $\text{var}(X_i) = p(1-p)$ $p_{X_i}(x) = \begin{matrix} p & x=1 \\ 1-p & x=0 \end{matrix}$

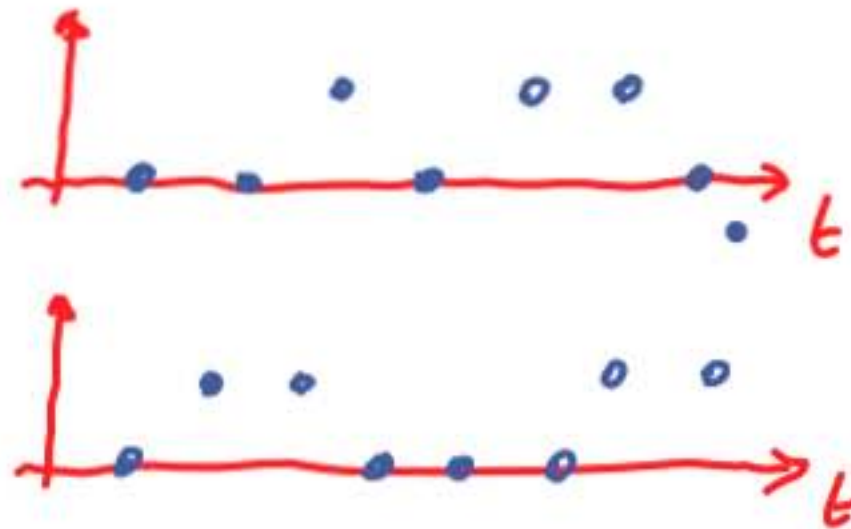
$p_{X_1, \dots, X_n}(x_1, \dots, x_n) = p_{X_1}(x_1) \dots p_{X_n}(x_n)$

for all n

- Second view – sample space:

$\Omega =$ *set of infinite sequences of 0's and 1's*

- Example (for Bernoulli process):



$$\mathbf{P}(X_i = 1 \text{ for all } i) = 0 \quad (p < 1)$$

$$\leq \mathbf{P}(X_1 = 1, \dots, X_n = 1) = p^n, \text{ for all } n$$

Number of successes/arrivals S in n time slots

- $S = X_1 + \dots + X_n$

- $P(S = k) = \binom{n}{k} p^k (1-p)^{n-k} \quad k = 0, \dots, n$

- $E[S] = np$

- $\text{var}(S) = np(1-p)$

Time until the first success/arrival

- $T_1 = \min \{ i : X_i = 1 \}$

- $P(T_1 = k) = P(\underbrace{0 \dots 0}_{k-1} 1) = (1-p)^{k-1} p$
 $k = 1, 2, \dots$

- $E[T_1] = \frac{1}{p}$

- $\text{var}(T_1) = \frac{1-p}{p^2}$

Independence, memorylessness, and fresh-start properties

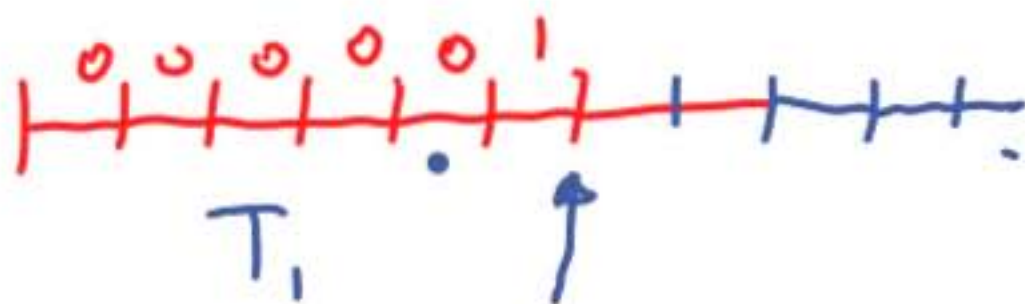
$$\{X_i\} \sim \text{Ber}(p)$$



$$\begin{aligned} Y_1 &= X_6^{X_{n+1}} \\ Y_2 &= X_7^{X_{n+2}} \\ &\vdots \end{aligned} \quad \{Y_i\} \quad i=1,2,\dots$$

- ① $\{Y_i\}$ independent of X_1, \dots, X_n
- ② $\text{Ber}(p)$

- Fresh-start after time n



$$\begin{aligned} Y_1 &= X_{T_1+1} \\ Y_2 &= X_{T_2+2} \\ &\vdots \end{aligned}$$

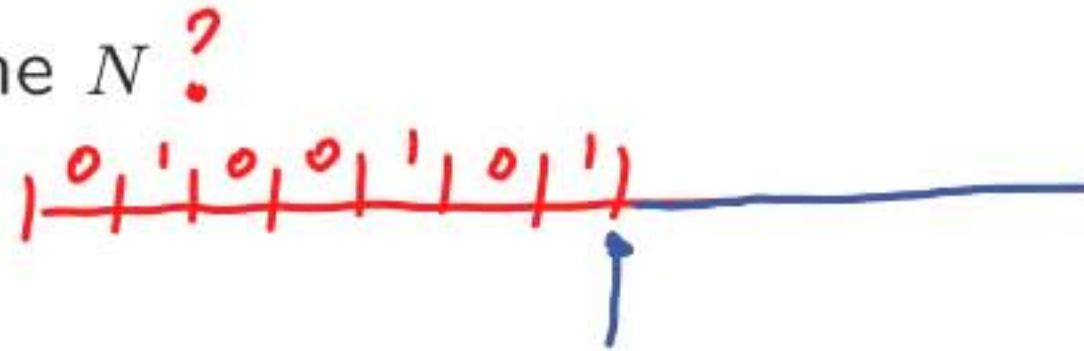
- ① $\{Y_i\}$ independent of X_1, \dots, X_{T_1}
- ② $\text{Ber}(p)$

- Fresh-start after time T_1

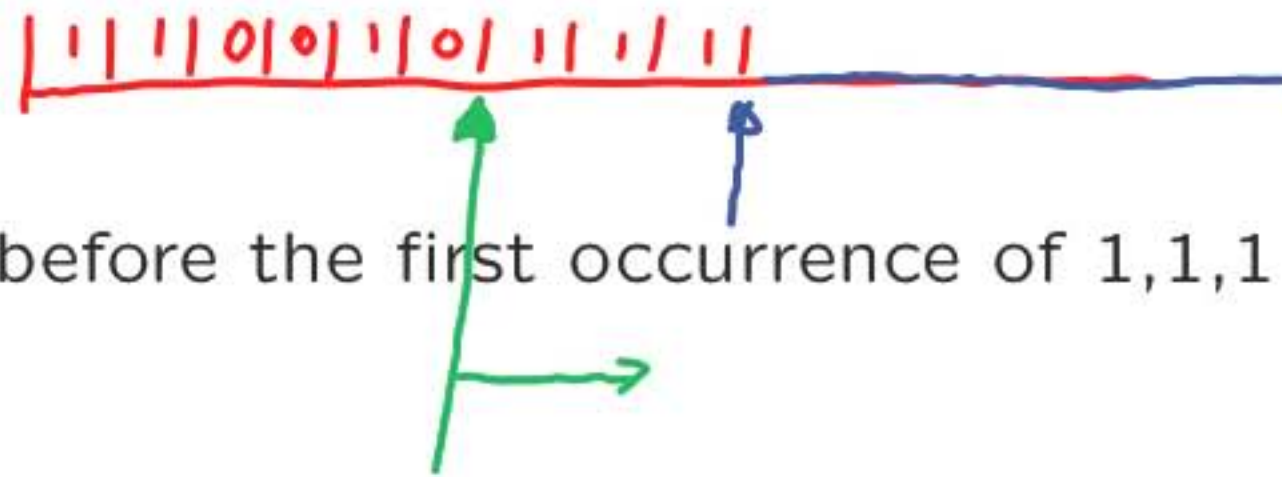
Independence, memorylessness, and fresh-start properties

- Fresh-start after a random time N ?

$N =$ time of 3rd success



$N =$ first time that 3 successes in a row have been observed



N = the time just before the first occurrence of 1,1,1

N is causally determined

N not causally determined

The process X_{N+1}, X_{N+2}, \dots is:

– a Bernoulli process

– independent of N, X_1, \dots, X_N

(as long as N is determined "causally")

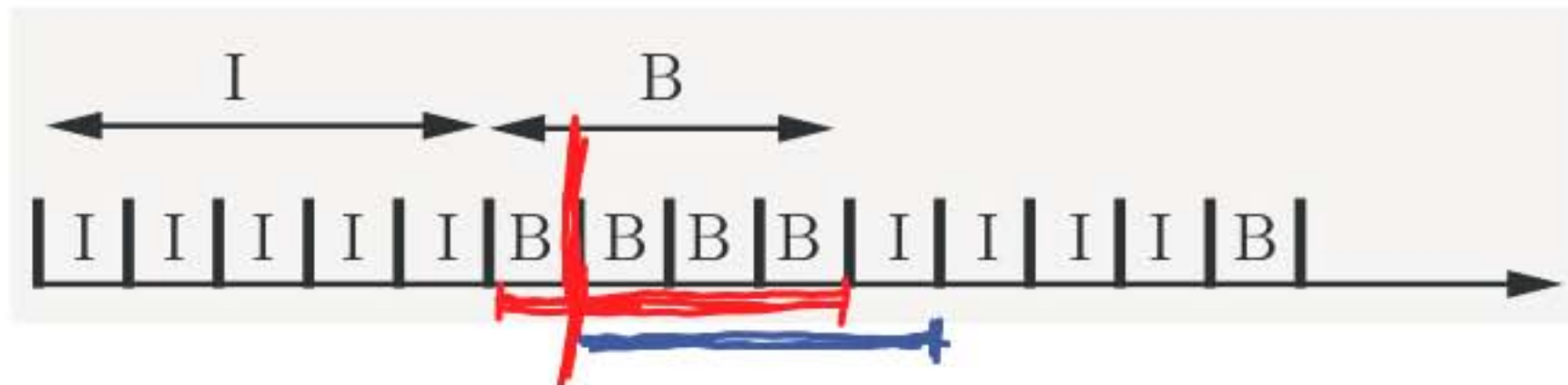
The distribution of busy periods

- At each slot, a server is busy or idle (Bernoulli process)

p

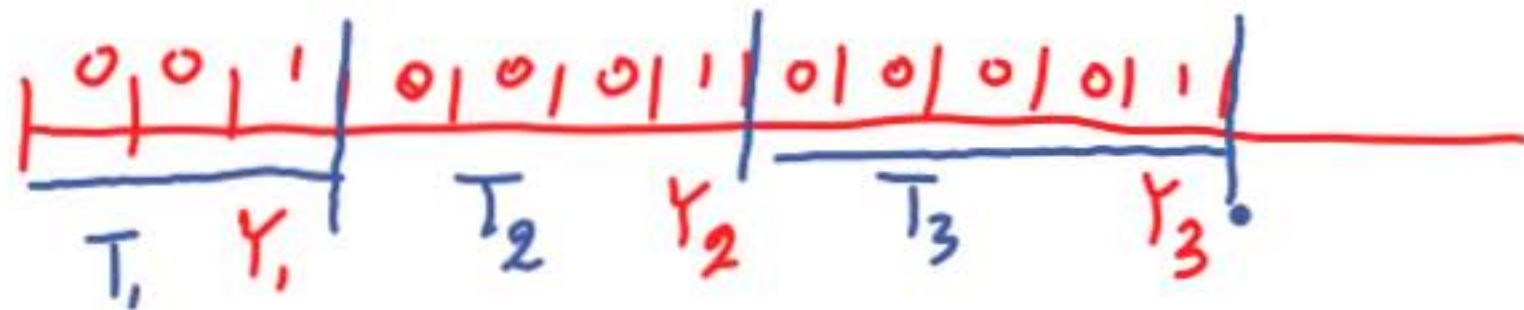
- First busy period:
 - starts with first busy slot
 - ends just before the first subsequent idle slot

$Geo(1-p)$



$Geo(1-p)$

Time of the k th success/arrival



- Y_k = time of k th arrival
- T_k = k th inter-arrival time = $Y_k - Y_{k-1}$ ($k \geq 2$)
- The process starts fresh after time T_1
- T_2 is independent of T_1 ; Geometric(p); etc.

$$Y_k = T_1 + \dots + T_k$$

Time of the k th success/arrival

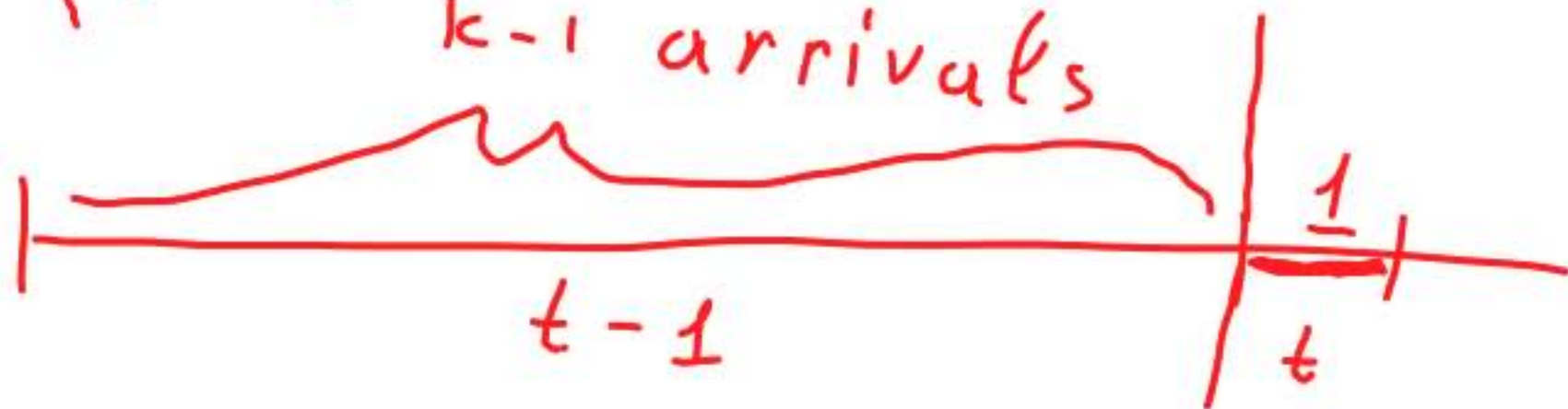
$$P(Y_k = t)$$

$$= P(k-1 \text{ arrivals in time } t-1)$$

$$\bullet P(\text{arrival at time } t)$$

$$= \binom{t-1}{k-1} p^{k-1} (1-p)^{t-k} \cdot p$$

$k-1$ arrivals



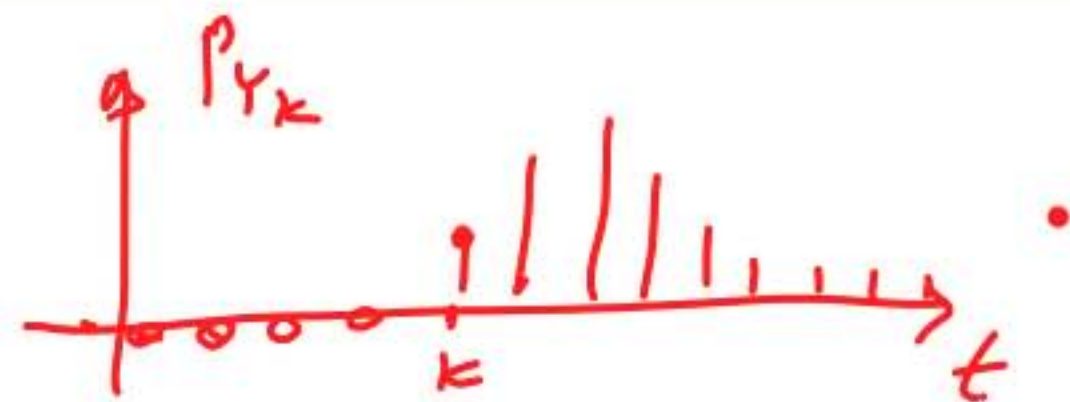
$$Y_k = T_1 + \dots + T_k$$

the T_i are i.i.d., Geometric(p)

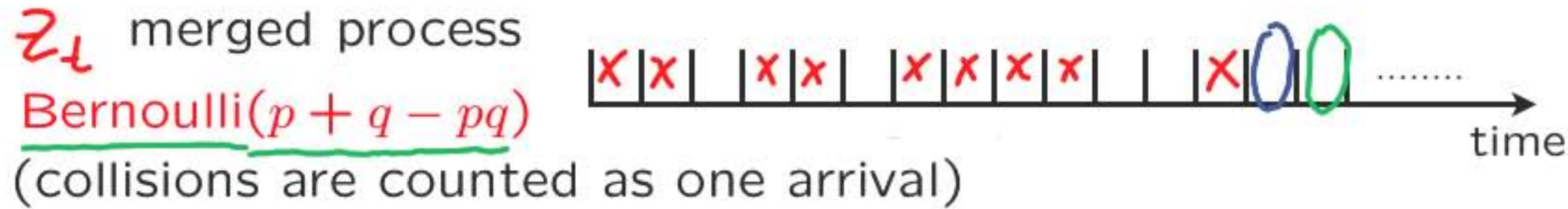
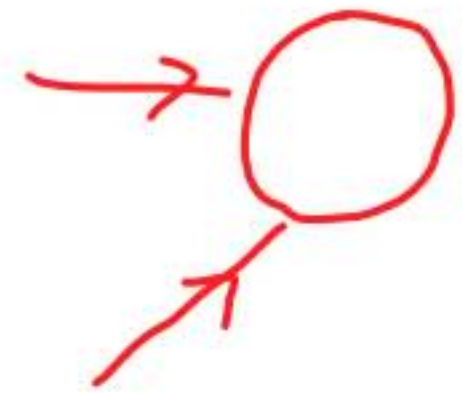
$$E[Y_k] = \frac{k}{p} \quad \text{var}(Y_k) = \frac{k(1-p)}{p^2}$$

$$p_{Y_k}(t) = \binom{t-1}{k-1} p^k (1-p)^{t-k},$$

$$\underline{t = k, k+1, \dots}$$



Merging of independent Bernoulli processes



$$Z_t = g(\underline{X}_t, \underline{Y}_t) \quad (Z_1, \dots, Z_t)$$

$$Z_{t+1} = g(\underline{X}_{t+1}, \underline{Y}_{t+1}) \quad 1 - (1-p)(1-q)$$

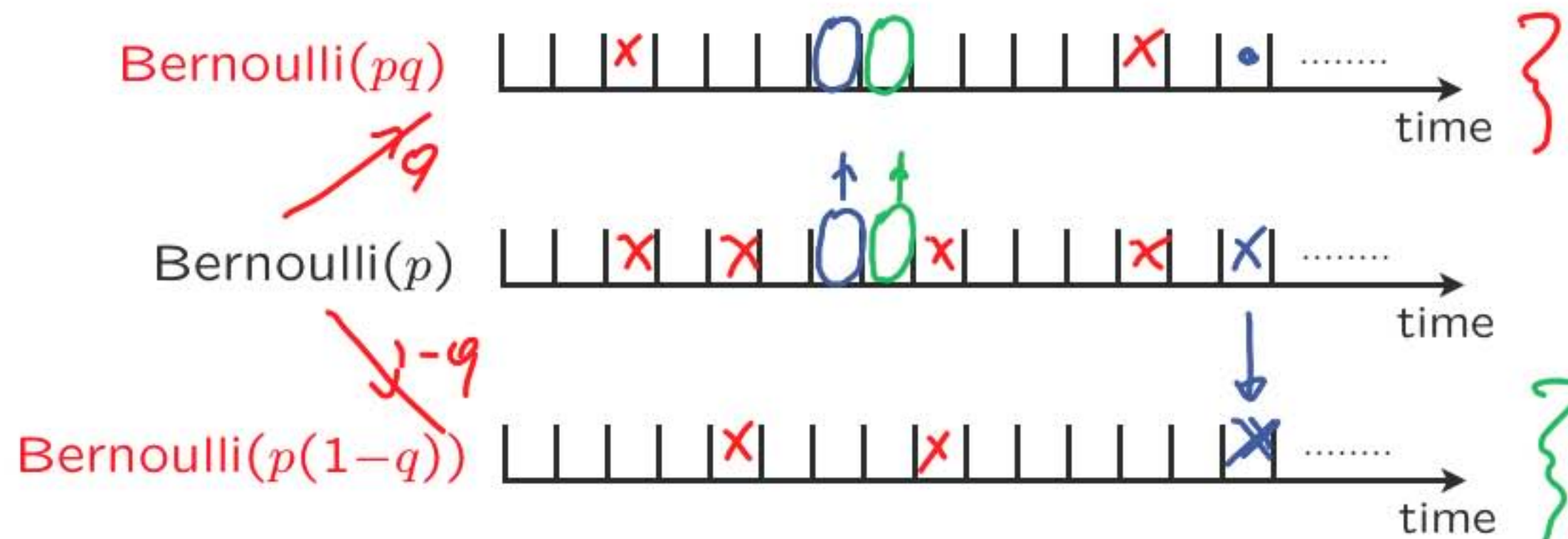
$$P(\text{arrival in first process} \mid \text{arrival}) = \frac{p}{p + q - pq}$$

	0	1
1	$(1-p)q$	pq
0	$(1-p)(1-q)$	$p(1-q)$
$\frac{Y}{X}$	0	1

Splitting of a Bernoulli process



- Split successes into two streams, using independent flips of a coin with bias q
 - assume that coin flips are independent from the original Bernoulli process



- Are the two resulting streams independent? *No*

Poisson approximation to binomial

- Interesting regime: large n , small p , moderate $\lambda = np$

$$\begin{aligned} & \bullet n \rightarrow \infty \\ & p \rightarrow 0 \quad p = \frac{\lambda}{n} \end{aligned}$$

- Number of arrivals S in n slots: $p_S(k) = \frac{n!}{(n-k)!k!} \cdot p^k (1-p)^{n-k}, \quad k = 0, \dots, n$

For fixed $k = 0, 1, \dots,$

$$p_S(k) \rightarrow \frac{\lambda^k}{k!} e^{-\lambda},$$

$$= \frac{n \cdot (n-1) \cdot \dots \cdot (n-k+1)}{k!} \cdot \frac{\lambda^k}{n^k} \left(1 - \frac{\lambda}{n}\right)^{n-k}$$

$$= \frac{n}{n} \cdot \frac{n-1}{n} \cdot \dots \cdot \frac{n-k+1}{n} \cdot \frac{\lambda^k}{k!} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-k}$$

$$\xrightarrow{n \rightarrow \infty} 1 \cdot 1 \cdot \dots \cdot 1 \cdot \frac{\lambda^k}{k!} e^{-\lambda} \cdot 1$$

- Fact: $\lim_{n \rightarrow \infty} (1 - \lambda/n)^n = e^{-\lambda}$

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<https://ocw.mit.edu>

Resource: Introduction to Probability

John Tsitsiklis and Patrick Jaillet

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