

LECTURE 23: RANDOM PROCESSES

• TOPICS TO COVER (BASED ON CH 3.5)

→ INTRODUCTION

→ BERNOULLI PROCESSES

→ POISSON PROCESSES

→ INTRODUCTION

A RANDOM PROCESS IS AN INFINITE SEQUENCE OF RANDOM VARIABLES USUALLY OCCURRING OVER TIME.

→ BERNOULLI PROCESSES

 $x_1, x_2, x_3, \dots$ $x_i \sim \text{BERNOULLI}(p)$ & i INDEPENDENTLY

RANDOM OUTCOME OF THE i th BERNOULLI TRIAL
IN AN INFINITE SEQ. OF TRIALS

E.G. SYSTEM FAILURE EVENTS OVER AN INFINITE SEQ. OF DISCRETE TIME POINTS

RECALL $(\Omega, \mathcal{F}, P)$: PROBABILITY SPACE

FOR ANY $\omega \in \Omega$: $x_k(\omega)$: REALIZED VALUE OF x_k
TIME

$x_1(\omega), x_2(\omega), x_3(\omega), \dots, x_k(\omega), \dots$: SAMPLE PATH OF A BERNOULLI

PROCESS FOR OUTCOME $\omega \in \Omega$

REFER TO THE FOLLOWING FIGURE :

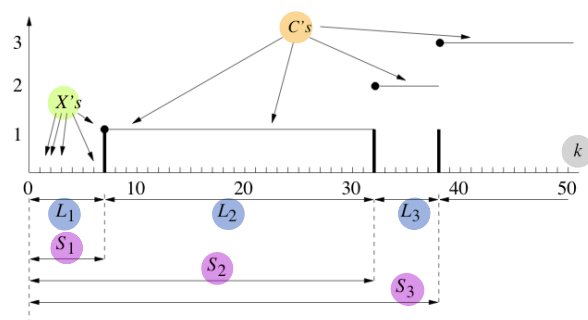


Figure 2.6: Depiction of a sample path of a Bernoulli process.

L_1 : # OF TRIALS NEEDED UNTIL THE OUTCOME OF A TRIAL IS ONE

$L_1 \sim \text{GEOMETRIC}(p)$

L_2 : # OF TRIALS, AFTER THE FIRST L_1 TRIALS, UNTIL AGAIN THE OUTCOME OF A TRIAL IS ONE.

$L_2 \sim \text{GEOMETRIC}(p)$

L_j : # OF TRIALS NEEDED AFTER THE FIRST $L_1 + L_2 + \dots + L_{j-1}$ TRIALS, UNTIL THE OUTCOME OF A TRIAL IS ONE.

$L_1, L_2, \dots, L_j, \dots$ INDEPENDENT RVS EACH FOLLOWING $\text{GEOMETRIC}(p)$

NOTE THAT THE VARIABLES $L_1, L_2, \dots, L_j, \dots$ ARE DETERMINED BY $X_1, X_2, \dots, X_j, \dots$

AND VICE-VERSA!

WE HAVE TWO MORE WAYS TO DESCRIBE THE BERNOULLI RANDOM PROCESS.

→ S_j : # OF TRIALS, COUNTING FROM THE VERY FIRST ONE, UNTIL A TOTAL OF j TRIALS HAVE OUTCOME ONE

$$S_j = L_1 + L_2 + \dots + L_j \quad j \geq 1$$

ALSO, $L_j = S_j - S_{j-1}$ WITH $S_0 = 0$

$$S_j \sim \text{NEGATIVE BINOMIAL}(j, p)$$

→ C_k : CUMULATIVE # OF ONES IN THE FIRST k TRIALS

$$C_k = X_1 + X_2 + \dots + X_k \quad k \geq 1$$

ALSO, $X_k = C_k - C_{k-1}$ WITH $C_0 = 0$

$$C_k \sim \text{BINOMIAL}(k, p)$$



POISSON PROCESSES

RECALL THE BERNOULLI PROCESS DISCUSSED ABOVE. LET $h > 0$, WITH h REPRESENTING THE AMOUNT OF TIME. SUPPOSE EACH TRIAL IN THE BERNOULLI PROCESS TAKES h TIME UNITS TO PERFORM. A TIME-SCALED BERNOULLI RANDOM PROCESS TRACKS THE # OF COUNTS VERSUS TIME. REFER TO THE FOLLOWING FIGURE:

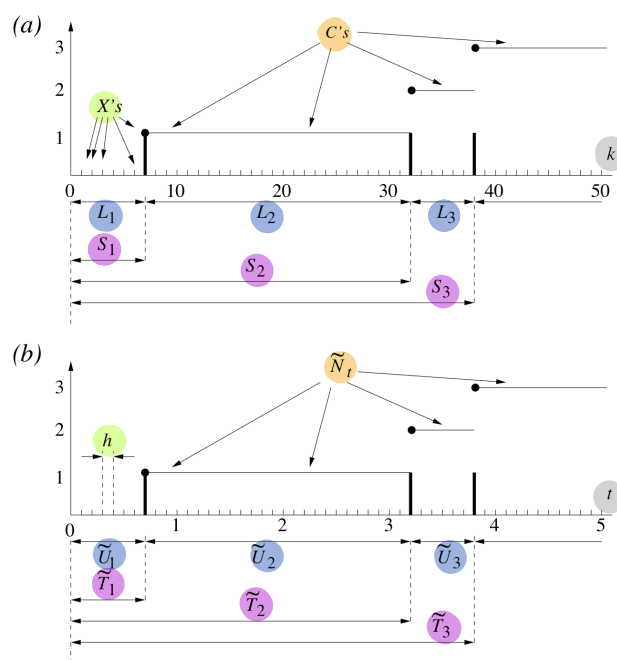


Figure 3.8: (a) A sample path of a Bernoulli process and (b) the associated time-scaled sample path of the time-scaled Bernoulli process, for $h = 0.1$.

DEFINE THE FOLLOWING RVS :

• $\tilde{U}_j = h L_j$: AMOUNT OF TIME BETWEEN THE $(j-1)$ TH COUNT AND THE j TH COUNT

• $\tilde{T}_j = h S_j$: THE TIME THE j TH COUNT OCCURS

• $\tilde{N}_t = C_{\lfloor t/h \rfloor}$: # OF COUNTS UPTO TIME t

TIME-SCALED BERNOULLI PROCESS

WE NEXT INTRODUCE THE POISSON PROCESS.

LET λ IS FIXED AND h IS SO SMALL THAT $p = \lambda h$ IS MUCH SMALLER THAN 1

WE CAN APPROXIMATE THE SCALED-BERNOULLI PROCESS AS FOLLOWS:

• $L_j \sim \text{GEOMETRIC}(p)$

AND SCALED VERSION OF L_j , NAMELY $\tilde{U}_j = h L_j \sim \text{EXPONENTIAL}(\lambda = p/h)$

WHY? $L_j \sim \text{Geometric}(p)$
 $h L_j \sim \text{CONT. ANALOG OF GEOMETRIC}$

• t FIXED, $\tilde{N}_t$: SUM OF $\lfloor t/h \rfloor$ $\text{BERNOULLI}(p)$ RVS

$$\Rightarrow \tilde{N}_t \sim \text{BINOMIAL}(\lfloor t/h \rfloor, p = \lambda h)$$

$$\therefore E \tilde{N}_t = \lfloor t/h \rfloor \cdot \lambda h \approx \lambda t$$

RECALL THAT AS $n \rightarrow \infty$ AND $p \rightarrow 0$ WITH $np \rightarrow \lambda$

$$\text{BINOMIAL}(n, p) \rightarrow \text{POISSON}(\lambda)$$

WE HAVE $h \rightarrow 0 \Rightarrow p = \lambda h \rightarrow 0$
 $\lfloor t/h \rfloor \rightarrow \infty$ $\lfloor t/h \rfloor \cdot \lambda h \rightarrow \lambda t$

THINK ABOUT IT!

$\therefore$ AS $h \rightarrow 0$, $\tilde{N}_t \rightarrow \text{POISSON}(\lambda t)$

MORE GENERALLY, IF $0 \leq s < t$, THEN

$$\tilde{N}_t - \tilde{N}_s \rightarrow \text{POISSON}(\lambda(t-s)) \quad \text{INDEPENDENTLY}$$

FORMAL DEFINITION OF A POISSON PROCESS

GIVEN $(\Omega, \mathcal{F}, P)$ A SAMPLE PATH OF A POISSON PROCESS (i.e., THE FUNCTION OF TIME THE PROCESS YIELDS FOR SOME PARTICULAR $\omega \in \Omega$) IS SHOWN IN THE FOLLOWING FIGURE:

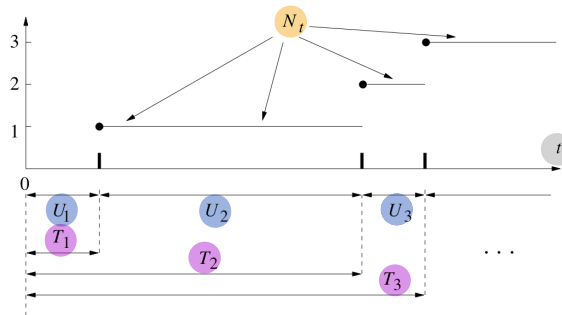


Figure 3.9: A sample path of a Poisson process.

- $t \geq 0$, N_t : CUMULATIVE # OF COUNTS UP TO TIME t

- $T_1, T_2, \dots$: COUNT TIMES

- $U_1, U_2, \dots$: INTER COUNT TIMES

$$N_t = \sum_{n=1}^{\infty} I_{\{t \geq T_n\}} \quad \text{WHERE } I_A = \begin{cases} 1, & A \text{ TRUE,} \\ 0, & \text{OTHERWISE.} \end{cases}$$

$$T_n = \min \{t : N_t \geq n\}$$

$$T_n = U_1 + U_2 + \dots + U_n$$

LET $\lambda \geq 0$. A POISSON PROCESS WITH RATE λ IS A RANDOM COUNTING PROCESS

$N = (N_t : t \geq 0)$ SUCH THAT

N.1 N HAS INDEP. INCREMENTS: IF $0 \leq t_0 \leq t_1 \leq \dots \leq t_n$, THE INCREMENTS

$N_1 - N_{t_0}$, $N_{t_2} - N_{t_1}$, $\dots$, $N_{t_n} - N_{t_{n-1}}$ ARE INDEPENDENT.

N.2 THE INCREMENT $N_t - N_s$ HAS THE POISSON $(\lambda(t-s))$ DISTRIBUTION FOR $t \geq s$.

PROPOSITION: LET N BE A RANDOM COUNTING PROCESS AND LET $\lambda \geq 0$. THE FOLLOWING

ARE EQUIVALENT:

(a) N IS A POISSON PROCESS WITH RATE λ .

(b) THE INTERCOUNT TIMES $U_1, U_2, \dots$, ARE MUTUALLY INDEP., EXPONENTIALLY DISTRIBUTED

RVS WITH PARAMETER λ .