Stochastic Processes: Syllabus and Exercise

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§ This course is suitable for those who have taken Basic Probability; some knowledge of Measure Theory is better (for Chapters 4,5).

§ The content and exercise are adapted from the followings:


J Lamperti: Probability (1966); Stochastic Processes (1977)

R Durrett: Essentials of Stochastic Processes (1998) [This one is recommended for your own reading]

§ Grading:

HmWk and OnBb: 40 percent; Exam(s): 60 percent. Subject to adjust.
Chapter 1: Basic Theory of Stochastic Processes

§ math frames, finite-dim’l distributions

§ Kolmogorov Existence Theorem

§ discrete time, continuous time, multiparameter time(random field)

§ state space

§ sample path continuity(Kolmogorov Continuity Theorem)

§ sample path jump property

§ equivalence of processes

§ convergence of processes

§ exercise of Chapter 1

1. BPR(Basic Probability Review): Prove the following monotone continuity of “probability measure”: If \( A_{n+1} \subset A_n \) and \( \cap A_n = \emptyset \) then \( \lim_{n \to \infty} P(A_n) = 0 \).

2. BPR: Prove the right continuity and the left limit existence of a distribution function \( F(x) \) on \( x \). How about the multivariate case \( x \in \mathbb{R}^n \).

3. PRR: Prove that, (i) if \( \{X_n\} \) is an uncorrelated random sequence, all with mean 0 and variance 1, then WLLN holds; (ii) if \( \{X_n\} \) is iid with fourth moments then SLLN holds.

4. BPR: Let \( \{X_n\} \) be iid \( B(p) \) and \( \bar{X}_n \) be sample mean, and \( f \) be a continuous function on [0,1]. Prove that \( B_n(p) := E^p[f(X_n)] \) is an \( n-th \) order polynomial and that \( B_n(x) \) converges to \( f(x) \), uniformly for \( x \in [0,1] \).
5. Refer to K Existence Theorem, prove that the class $F$ of all sets of the form 
\[ \{ \omega; a_i < \omega(t_i) \leq b_i, i = 1, \cdots, k, t_i, a_i, b_i, k \text{ vary freely}, \} \]
form a field, i.e. $\Omega \in F$, $C_1 \cap C_2 \in F$ and $\bigcup_{i=1}^{m} C_i \in F$ whenever the assumed members are in $F$. Prove that $P$ defined in K theorem is an “additive set function” on $F$.

6. Let $F_n$ be a sequence (finite or infinite) of distribution functions on $R$. Refer to K Theorem, prove that there is a probability space $(\Omega, \mathcal{F}, P)$ and a sequence of rv’s $X_n$ defined on $\Omega$ such that $X_n$ are independent and each $X_n$ is of distribution $F_n$.

7. Let $F_n$ in Ex 6 be of the form $F_n(x) = 1, x < 0, = 1/2, 0 \leq x < 1, = 1, x \geq 1$. Show that how the probability space in Ex 6 corresponds to the unit interval $[0,1]$ with $P$ corresponding to Lebesgue-Borel measure.

8. Let $X_t$ be a continuous time process such that the increments $X_t-X_s$ are Gaussian with mean 0 and var $t-s$. Prove that $X$ has continuous sample paths, by applying K Continuity Theorem and the moments of mean 0 Gaussian rv which is given in BP.

9. How do you figure K Existence Theorem and K Continuity Theorem for multi-parameter time (random field) case?

10. BPR: A sequence of distributions $F_n$ on $R$ is said to converge to a distribution $F$ if $F_n(x) \to F(x)$ at each continuity point $x$ of $F$. It is then also said the corresponding $X_n$ converges to $X$ in distribution. Prove that, if $X_n$ converges to $X$ in probability then it converges in distribution.
Chapter 2: Discrete Time Markov Chains

§ definition, transition probabilities

§ finite–dim’l distributions

§ random walks

§ examples of finite MC: on–off, weather, gambling, Ehrenfest, Wright–Fisher

§ examples of denumerable MC: queue, branching, birth and death

§ n–step transition probabilities, Chapman–Kolmogorov equation

§ $i \rightarrow j$, $i \leftrightarrow j$, irreducibility

§ periodicity

§ recurrent states and transient states

§ usage of probability generating function(pgf)

§ recurrence theorem

§ recurrence and transience of SRW

§ discrete renewal equation

§ basic limit theorem

§ ergodicity

§ stationary equation and stationary distribution

§ examples of stationary MC

§ Galton–Watson branching process(GWBP), usage of pgf

§ GWBP, extinction probability
§ examples of GWBP

§ exercise of Chapter 2

1. BPR: conditional probability, multiplicative formula and formula of total probabilities. \( P(\cdot | E) \) is also a probability measure.

2. Assume the weather of any day depends the weather condition of two previous days. To set a weather chain under this assumption by introducing the product space of the original states \( S, C, R \).

3. An inventory chain is a model for continuing demand of some goods, say video game, in one, say, PC, store. If at the end of one business day, the number of units they have at hand is 1 or 0, then they order some new units so that their total on hand is, say, \( S=5 \). Assume the new units can arrive before the new business day. Assume the number of customers to buy one unit at one day is 0, 1, 2, or 3 with probabilities 0.3, 0.4, 0.2 and 0.1. Describe the transition matrix with state space \( 0, \cdots, 5 \). In general, when the stock at hand is \( \leq s \) then we order new units to bring the stock back to \( S \). Let \( d_{n+1} \) be the demand at the day \( n+1 \), use the positive part function \( x^+ = \max\{x, 0\} \) to write the \( X_{n+1} \) in terms of \( X_n, S, d_{n+1} \).

4. In the above video game store, assume that it makes USD 12 profit on each unit sold but it takes USD 2 per day to store one unit. What is the long–run profit per day of \( S = 5, s = 1 \) inventory policy? How to choose \( S, s \) to make the maximum profit?

5. In the gambling chain, let \( h(x) \) denote the probability of A to win all money, i.e. B is ruined, when A starts with X dollars. Calculate the \( h(x) \) by solving a certain
difference equation of \( h(x) \), with boundary conditions \( h(0) = 0, h(N) = 1 \). Note that the answer is of different type for the case \( p = q = 1/2 \) and the case \( p \neq q \).

6. In Ex 4, calculate the expected duration time of the game, in the fair case \( p = q = 1/2 \), \( H(x) = E^{x}[T], T := \min\{n : X(n) = 0, N\} \). Again, to solve a certain difference equation of \( H(x) \), with boundary condition \( H(0) = H(N) = 0 \).

7. Discuss the barrier \( 0, N \) for WF model with mutation.

8. Write down the trnasition matrix for birth and death chain.

9. Use CK Equation to prove the relation \( i \rightarrow j \) is transitive.

10. Determine the classes and the periodicity of all states in the following Markov matrix

\[
\begin{bmatrix}
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 \\
0 & 1 & 0 & 0 \\
1/3 & 0 & 2/3 & 0
\end{bmatrix}
\]

11. Use CK Eq to prove that \( i \leftrightarrow j \Rightarrow d(i) = d(j) \).

12. Use \( \sigma \)-additivity of probability to prove that \( i \) is rec iff \( \sum_{n=0}^{\infty} f_{ii}^{(n)} = 1 \).

13. Use power series and Cauchy product to prove that \( P_{ii}(s) - 1 = F_{ii}(s)P_{ii}(s) \), and that \( P_{ij}(s) = F_{ij}(s)P_{ij}(s), i \neq j \).

14. Use CK to prove that \( i \leftrightarrow j \iff i, j \) either both rec or both tran.

15. Try to use three states MC to show that the expected number for tossing a fair coin to get two consecutive Tails is 6. Durrett p70.

16. Let \( x = \{x_i\} \) satisfy the stationary eq \( x = xP \). Prove that, for each \( n, x_j = \sum_i x_i p_{ij}^{(n)} \).
17. A Markov matrix is called doubly stochastic if \( \sum_i p_{ij} = 1 \) for each \( j \). Prove that the corresponding MC then has at most one stationary prob distribution.

18. Prove that for the merry–go–round model, the stationary distribution is \( \pi_i = 1/l, \forall i \).

19. Find the stationary distribution for Ehrenfest chain.

20. Write down and try to solve, by exact or by numerical, the extinction probability eq’s for the binary and the binomial GWBP’s.

21. Calculate \( f_n(s) = f(f_{n-1}(s)) \) explicitly for \( f(s) := \frac{\alpha + \beta s}{\gamma + \delta s}, \quad \alpha \delta - \beta \gamma \neq 0 \).

22. Apply Ex 21 to get \( P\{X_n = 0\} \), and then verify the extinction formula in this case.
§ continuous times Markov transitions probabilities and CK Equation

§ contraction from discrete time to continuous time MC

§ Poisson process

§ Yule process

§ birth and death process

§ infinitesimal generating matrix

§ renewal process and its SLL, CLT

§ renewal equation and renewal theorem

§ continuous time Markov branching process, usage of pgf

§ continuous time Markov branching process, extinction probability

§ exercise of Chapter 3

1. BPR: show that an exponential distribution $T$ with parameter $\lambda$ has the following “lack of memory” property: $P(T > t + s|T > t) = P(T > s)$.

2. Try the following Matlab graphic: Let $U_i$ be $U(0,1)$ distributed and let $\tau_i = \ln U_i$. Define, as your will, the transition of $Y_n$ by some random numbers. Graph $X_t$ based on $Y_n$ and $\tau_i$.

3. BPR: in the waiting times or inter-arrival times $T_i, i = 1, 2, \cdots$, of a Poisson process, show that the $S_n = T_1 + \cdots T_n$ is a $\Gamma$ distribution by showing the convolutions.

4. Compute the covariance of $X(t)$ and $X(t + s)$ of a Poisson process.
5. Let \( Y_1, Y_2, \ldots \) be iid random sequence, and let \( N(t) \) be a Poisson process, independent on \( Y_1, Y_2, \ldots \). Define \( X(t, \omega) = Y_1 + \cdots + Y_{N(t, \omega)} \), with \( X(t, \omega) = 0 \) when \( N(t, \omega) = 0 \). It is called \( Y(t) \) a compound Poisson process. Calculate \( E(X(t)), \text{var}(X(t)) \) by following Wald identities. Write \( X, N \) for \( X(1), N(1) \), show that \( E(X) = EN \cdot EY_1, \text{Var}(X) = EN \cdot \text{var}(Y_1) + \text{Var}(N)(EY_1)^2 \). Key: look at the formulae when \( N \) is nonrandom, then calculate \( E(X) \) by using total probability formula of conditional probability based on \((N = n)\), how the independence assumption will be used? Durrett p138

6. Let \( N_1(t), N_2(t) \) be two independent Poisson processes, with parameter \( \lambda_1, \lambda_2 \), prove that the sum process \( X(t) = N_1(t) + N_2(t) \) is also a Poisson process, with parameter \( \lambda_1 + \lambda_2 \).

7. Verify Yule process, with parameter \( \beta \), has \( p_k(t) = e^{-\beta t} (1 - e^{-\beta t})^{k-1}, k = 1, 2, \ldots \). Then verify the pgf of \( X(t) \) is

\[
 f_t(s) = \frac{se^{-\beta t}}{1 - (1 - e^{-\beta t})s}, -1 \leq s \leq 1.
\]

8. Consider a Yule process with \( X(0) = N, N > 1 \). Regard the process as the sum of \( N \) independent Yule processes, each starts with 1. Then obtain the pgf of \( X(t) \).

9. In Ex 7, obtain \( p_k(t) \) by using the Taylor expansion of \((1/(1 - x))^N\).

10. In Yule process, with parameter \( \beta \), let \( T_n \) be the time, start with 1 particle, to get \( n + 1 \) particles. Prove that \( ET_n \approx (\ln n)/\beta, \) as \( n \) large.

11. In birth and death process, assume that \( \sum_k \pi_k < \infty \) and that \( \sum_k = 1 \), prove that \( p_j = \frac{\pi_j}{\sum_k \pi_k}, j = 1, 2, \ldots \).
12. In birth and death process, $\lambda_k = \lambda k + a, \mu_k = \mu k$, let $M(t) = EX(t)$, use 2nd Kolomogrov equation to derive that $M'(t) = a + (\lambda - \mu)M(t)$. Try to solve this d.e. ($\lambda =, >, < \mu$ separately)


14. Carry out the proof of CLT for renewal process

15. In renewal process, assume at $i$th renewal you can earn a reward(or say you need to pay the cost) $r_i$. Write $R(t) = \sum_{i=1}^{N(t)} r_i$. Try to explain the limit $\frac{R(t)}{t} \to \frac{E r_i}{E X_i}$.

17. In continuous time Markov branching process, calculate the $Var(X(t))$ in terms of $u'(1), u''(1)$ by the similar way in GWBP in Chapter 2. There are two cases $u'(1) \neq 0$ and $u'(1) = 0$. KT p438

18. In continuous time Markov branching process, let $u(s) = s^k - s$, say $k = 2$, solve $\frac{\partial \phi(t,s)}{\partial t} = u(\phi(t,s), \phi(0,s) = s$ to get $\phi(t,s)$. KT p 439

19. Repeat Ex 18 with $u(s) = 1 - s - \sqrt{1 - s}$. KT p439

20. If the life time of the particle is a distribution with density $(1/2)te^{-t}$, then the corresponding continuous time branching process is not Markovian. However the distribution is the sum of two independent exponent distributions, with parameter 1. So we can think the branching process as $X(t) = (X_1(t), X_2(t))$, each $X_i$ is Markov branching, try derive the $u$ function of this two–type branching process, here $u$ should be a vector $u = (u_1, u_2)$ and $u_i$ have two variables $s_1, s_2$. KT p431
Chapter 4: Second Order Processes

§ review of Hilbert space $L^2(\Omega, \mathcal{F}, P)$: CS inequality, inner product, completeness, GS orthonormalization, decomposition theorem

§ 2nd order process and its covariance function

§ continuity and differentiation of continuous time 2nd order processes

§ integration of continuous time 2nd order processes

§ covariance–stationary 2nd order processes

§ spectral measure of covariance function

§ spectral representation of 2nd order processes

§ predictions: linear prediction and optimal prediction

§ Wold decomposition

§ Gaussian process and system

§ exercise of Chapter 4

1. Prove CS inequality and GS o.n. procedure.

2. BPR: WLLN and SLLN; let $\xi_n$ be iid mean 0 and var 1. Let $X_n = \frac{1}{n} \sum_{k=1}^{n} \xi_k$.

Prove that $X_n \to 0$ both in probability and in the mean square sense. Give an example show that in general the convergence cannot be w.p. 1. Prove that it can be w.p.1 if $\xi_n$ is assumed to have 4th moments.

3. Express the mean square $||X_t - X_s||^2$ by the covariance function $K$. Then use the completeness to prove that if $K(s, t)$ continuous at $(t_0, t_0)$ then the process is continuous
4. A Wiener process is a continuous time 2nd order process such that the process has independent and stationary increment (recall the Poisson process) and that each $X_t$ is Gaussian with mean 0 var $\sigma t$, for some constant $\sigma > 0$. Compute its cov function, and see that the process is continuous yet not differentiable.

5. Do Ex 4 for centered Poisson process $N_t - \lambda t$, where $N_t$ is a Poisson process with parameter $\lambda$.

6. Try to formulate “fundamental theorem of Riemann calculus” for the 2nd order continuous time process.

7. Let $X(t), Y(t)$ be two independent 2nd order processes, with cov function $K_1, K_2$. Prove that $K_1 \cdot K_2$ is the cov function of the product process $X(t)Y(t)$.

8. Define random cosine process by $X_t = A\cos(\eta t + \phi)$, where $A, \eta$ are real–valued rv’s. and $\phi$ is $Unif(0, 2\pi)$ and is independent of $A, \eta$. Compute the cov function of the process and see if it is cov–stationary.

9. Compute the cov function of AR process $X_n = \alpha X_{n-1} + \beta \xi_n, n \in \mathbb{Z}$.

10. Let $A_1, \cdots, A_n$ be orthogonal rv’s and $\lambda_1, \cdots, \lambda_n$ be reals. Define the process $X_t = \sum j = 1^n A_j e^{i\lambda_j t}$. Compute its cov function and spectral measure $dF(\lambda)$.

11. Let $X_t$ be a cov–stationary process; write down, formally, the spectral representations of the differentiation process $X'_t$ and the integration process $\int_0^t X(s)ds$ from that of $X_t$. What should be sufficient conditions to ensure the formal expressions?

12. Let $X_t$ be an orthogonal increments process with associated function $F$. Fix
Define integral process $Y_t := \int_{t_0}^{t} f(u) dX(u) = \int 1[t_0, t] 9u f(u) dX(u)$. Prove that $Y_t$ is also an orthogonal increments process and that the associated function of $Y$ is given by $G(t) = \int_{t_0}^{t} |f|^2 dF$. How about the integral $\int gdY = \int g f dX$.

13. BPR: recall two rv’s $X, Y$ form a bivariate Gaussian distribution, if $X, Y$ have jpdf $f(x, y) := \cdots$; write down the conditional density of $X$ with respect to $Y$.

14. Let $\xi_n$ be a noise process, and let $X_n = \xi_n - \alpha \xi_{n-1}$. Show that the spectral measure $dF$ of the process $X_n$ has a density $f(\lambda) = \frac{1}{2\pi} |1 - \alpha e^{i\lambda}|^2$.

15. Let $X_n, n \in Z, be cov stationary, with cov function $K(m), m = 0, \pm 1, \cdots$ let $\alpha_i^*, i = 1, \cdots, p$ be such that $\hat{X}_n = \alpha_1^* X_{n-1} + \cdots + \alpha_p^* X_{n-p}$ is the best linear prediction of $X_n$ w.r.t. $X_{n-1}, \cdots, X_{n-p}$. Show that $\alpha_i^*, i = 1, \cdots, p$ solve the linear equations $K(j) = \alpha_1^* K(1) + \cdots + \alpha_p^* K(p-j), j = 1, \cdots, p$. Find the prediction error. KT p471
Chapter 5: Brownian Motions

§ math def of BM: Wiener process, sample path continuity

§ finite dim'l distributions of BM

§ Fourier–Wiener expansion of BM

§ Markov and strong Markov property

§ First passage time, reflection principle

§ nondifferentiability of Brownian paths

§ quadratic variation of Brownian paths

§ BM with drift

§ $U(t) = B^2(t) - t$ and $V(t) = \exp(\lambda B(t) - (1/2)\lambda^2 t)$

§ multi-dim'l BM

§ Brownian bridge and OU process

§ exercise of Chapter 5

D p265 Ex 6.1– 6.11