

# STAT 721: STOCHASTIC PROCESSES

Spring 2024

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| <b>Instructor:</b> Ray Bai        | <b>Time:</b> 1:10-2:00 pm MWF |
| <b>Email:</b> RBAI@mailbox.sc.edu | <b>Place:</b> LeConte 107     |

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**Course Page:** <https://blackboard.sc.edu/> (Check regularly for updates)

**Office Hours:** By appointment. I am also very accessible by e-mail and will typically reply to e-mails within one business day of receiving them.

**Course Description:** Stochastic processes are probabilistic (non-deterministic) systems indexed by time or space. They are very useful for modeling a variety of phenomena that vary in a random manner, such as the volatility of financial assets or the spatial distribution of crimes in a city. Stochastic processes are also very useful for many machine learning tasks such as regression, clustering, and reinforcement learning.

This course introduces methodology, applications, and theory for stochastic processes. It focuses on the following five topics: 1) point processes, 2) mathematical finance, 3) Bayesian nonparametrics, 4) Markov chain Monte Carlo (MCMC), and 5) reinforcement learning. The tentative schedule of topics is as follows:

- **Weeks 1-2:** point processes (Poisson processes, spatial point processes, nonhomogeneous Poisson processes, simulation algorithms)
- **Weeks 3-5:** mathematical finance (pricing of European options, random walks, Brownian motion, stochastic differential equations, binomial model, Black-Scholes model)
- **Weeks 6-8:** Gaussian processes (Bayesian inference, stationary and nonstationary kernels, GP regression, empirical Bayes, global approximation methods for GPs in big data)
- **Week 9-10:** Markov chain Monte Carlo (Markov chains, stationarity and convergence, Metropolis-Hastings, Gibbs sampling, Hamiltonian Monte Carlo, MCMC for Bayesian inference)
- **Week 11:** Dirichlet processes (de Finetti's Theorem, exchangeability, stick-breaking process, Chinese restaurant process, Dirichlet process mixture models)
- **Weeks 12-14:** reinforcement learning (Markov decision process, dynamic programming, exploration vs. exploitation, Q-learning, policy gradients, deep reinforcement learning)
- **Week 15:** group project presentations

## Learning Outcomes:

1. Be able to apply stochastic processes to a variety of tasks, including spatial statistics, financial modeling, Bayesian inference, regression, clustering, and training intelligent agents.
2. Develop programming and algorithm design skills by implementing the models covered in the course.
3. Understand the theory of stochastic processes for point processes, option valuation, Bayesian inference, MCMC, and reinforcement learning and appropriately apply this theory to solving problems.
4. Practice effective communication skills through writing scientific reports and making presentations.

**Prerequisites:**

- STAT 712 **or** STAT 511/512 (or equivalent)
- MATH 344 or MATH 544 (or equivalent)

Students must also be able to write functions in one programming language such as R, Python, MATLAB, or C++. These are strict requirements. Students who cannot code basic functions or who have not taken probability/statistics/linear algebra at the levels specified above will not be allowed to register for the course. Students who have taken only STAT 712 (or its equivalent) may register for the class. However, if you have *not* taken STAT 712, then you must have taken *both* STAT 511 and STAT 512 (or their equivalent).

**Main References:** We will use typed handouts prepared by the instructor.

**Computing:** This course involves programming. Please use one of the following languages: Python, R, MATLAB, C, or C++.

**Homework:** There will be five homework assignments. Students will work in small groups, and each group will submit a single, typed report for each homework assignment. The homework may consist of both conceptual/theoretical exercises and questions that involve programming.

All homework reports should be typed, including answers to math exercises and data analysis portions that may include figures, tables, etc. Your code should be submitted along with your homework report.

**Project:** Students will work in small groups to research a topic of their choosing, prepare a 15-minute presentation, and write a short report in the style of a journal article: abstract, introduction, method, data analysis, and a bibliography. The last week of the semester will be devoted to project presentations. Some potential examples of projects include:

- numerical methods for pricing American options
- Hawkes process for modeling social media contagion
- local approximation methods for Gaussian processes
- Bayesian additive regression trees
- Monte Carlo Tree Search or other reinforcement learning algorithms not covered in lecture
- Gillespie algorithm for simulating stochastic equations of molecular reactions

Projects must be approved in advance by the instructor, and no two groups may do the same topic for their project. If you have an idea of what you want to do for your project, please “claim” it early. Detailed instructions for the presentation and the report will be given at a later date.

**Grading:** Your grade will be determined by homework (70%) and the project (30%). The tentative grading scale is as follows: 90-100 for an A, 80-89 for a B+, 70-79 for a B, 60-69 for a C+, 0-59 for a C.

**Honor Code:** See the Carolinian Creed in the *Carolina Community: Student Handbook and Policy Guide*. The *minimum* punishment for violations of the USC Honor Code is a grade of zero for the work in question. In accordance with university policy, there may be other punishments, including an automatic F in the class and/or expulsion from the university.

**Accommodation:** If you need special accommodations for any aspects of the course, please contact me before or during the first week of the semester. Note that reasonable accommodations are available for students with a documented disability. If you have a disability and may need accommodations to fully participate in this class, contact the Office of Student Disability Services by phone (803-777-6142) or e-mail [sasds@mailbox.sc.edu](mailto:sasds@mailbox.sc.edu). All accommodations must be approved through the Office of Student Disability Services.