## **General Exam Instructions**

If you are planning to take the Probability part of your general exam with me, you need to go over my notes that cover both Probability I and II. The latest version is located here:

## http://merlot.stat.uconn.edu/~boba/stat6894/probabilityl.pdf

I expect you to memorize all the major definitions and major theorem statements. I will not ask you to reproduce relatively complicated proofs. However, you have to know how to do simple standard tricks. Here are some examples: proving Chebyshev's Inequality, deriving formulas for characteristic function of standard distributions, checking that a sequence of random variables forms a martingale, giving an example of a sequence of random variables that converges in probability but not with probability one.

Here is the list of definitions, concepts, and theorem statements that you absolutely must know. A failure to answer any of these questions automatically means that you failed the exam.

- 1. Definitions of  $\sigma$ -field.
- 2. Probability space.
- 3. Continuity and  $\sigma$ -additiveness of additive measure.
- 4. Random variable.
- 5. Definition of independence.
- 6. Borel-Cantelli lemmas.
- 7. Expectation of simple random variables.
- 8. Expectation of random variables.
- 9. Monotone convergence theorem.
- 10. Fatou's lemma. Dominated convergence theorem.
- 11. Chebyshev's inequality.
- 12. Cauchy-Schwarz's inequality.
- 13. Jensen's inequality.
- 14. Radon-Nikodym theorem.
- 15. Product spaces and Fubini's theorem.
- 16. Characteristic Functions.
- 17. Different Types of Convergence: with probability 1, in  $L_p$ , in probability, in distribution.
- 18. Continuity Theorem
- 19. Week Laws of Large Numbers for iid random variables.
- 20. Central Limit Theorem for iid random variable.
- 21. Conditional Expectation.
- 22. Martingale: Definition.
- 23. Doob's Optional Stopping Theorem.
- 24. Doob's Submartingale Inequality.

For example, I can ask you to explain how we define the expectation of non-negative random variable X. Here is the answer.

Since any random variable is a measurable mapping of  $(\Omega, \mathcal{F})$  to  $(\mathbb{R}, \mathcal{B})$ , we can construct a sequence of non-negative simple random variables  $\{X_n\}$  such that  $X_n(\omega) \uparrow X(\omega)$  for any  $\omega \in \Omega$ . The expectation of a simple random variable is given by an explicit formula. Then  $E(X) = \lim_{n \to \infty} E(X_n)$ . This limit exists and does not depend on the approximation of X by  $X_n$ .

Additionally, I can ask you to define measurability, to prove the existence of the limit, to define simple random variable etc. I will not ask you to show that the limit is the same for any two approximations.