APMA 1650 Final Exam Checklist

The exam will be broken down into 50% on new material (limit theorems, estimators and confidence intervals) 25% on material from midterm 1 and 25% material from midterm 2. Some fundamental concepts like random variables, expectation, conditional probability, etc are cumulative and permeate everything we have done in the class.

Stuff from Midterm 1 (25% of the exam)

- 1. Basics of set theory: Make sure you understand how to manipulate sets and the various operations that can be performed.
 - (a) Union, intersection, complement, mutually exclusive, Venn diagrams
 - (b) DeMorgans Law and algebraic properties of sets
- 2. Fundamentals of probability
 - (a) laws of addition, complement, difference
 - (b) how to calculate probabilities by splitting up events and adding and subtracting their probabilities
 - (c) Venn diagrams
- 3. Discrete probability and combinatorics
 - (a) Equally likely events, counting, rule of products, permutations, combinations, partitions
 - (b) Sampling with/without replacement un/ordered
 - (c) Make sure you understand the formulas for permutations, combinations and partitions
- 4. Conditional probability
 - (a) Know the definition and what it means intuitively
 - (b) Law of multiplication, law of total probability, probability trees
 - (c) Independence, what does it mean, how to check it
 - (d) Bayes rule, and inverting conditional probabilities
- 5. Discrete random variables and distributions
 - (a) PMF and CDF
 - (b) Expected value and variance. Know their properties and how use use them (eg: linearity and how variance rescales)
 - (c) Special distributions: Bernoulli, Binomial, Geometric, Poisson, Pascal. Know their properties, what they describe and how they relate.
 - (d) What happens to a PMF of an RV when you apply a function to the RV?

Stuff from Midterm 2 (25% of the exam)

- 1. Continuous random variables and distributions
 - (a) Probability density function, cumulative distribution functions
 - (b) Expected value and variance again
 - (c) Special Distributions: Uniform distribution, exponential distribution (how it relates to geometric and Poisson), normal distribution, Gamma distribution (how it relates to exponential and Pascal).
 - (d) CDF method. What is the distribution of h(X)?
- 2. Multivariate distributions
 - (a) Joint distributions and CDFs, iterated integrals and integrating over regions
 - (b) Marginals, conditional distributions, independent random variables
 - (c) Covariance and correlation. Relation to independence. Can two random variables have no correlation, but be dependent? Do you have an example?
 - (d) Conditional expectation and conditional variance. How to use them to compute unconditional variance and expectation (law of total expectation law of total variance)

New material (50% of the exam)

- 1. WLLN, CLT and convergence of random variables
 - (a) What is the weak law of large numbers and what does it say?
 - (b) Chebyshev's inequality, what does it mean and how to apply it to bound probabilities.
 - (c) How to prove the weak law of large numbers from Chebyshev's inequality.
 - (d) What is the central limit theorem? Why is it so important? How do you apply it?
 - (e) Compare CLT to WLLN.
 - (f) What is convergence in probability and distribution? Which one is stronger? How do they relate to WLLN and CLT? Does convergence in probability and distribution still hold if you apply a continuous function? What about if you multiply two random variables converging in distribution?
- 2. Sampling and estimation
 - (a) iid random samples what are they?
 - (b) Point estimators, what are they? What are some examples?
 - (c) Bias of an estimator. How to make a biased estimator unbiased.
 - (d) Mean square error (MSE), it's relation to variance and bias of an estimator, know how to compute.
 - (e) How to show that one estimator is better than another using MSE.
- 3. Consistency

- (a) What does it mean to show an estimator is consistent. How to use the MSE to show consistency.
- (b) How to use the weak law of large numbers to show consistency. How does this relate to MSE?
- (c) Properties of convergence in probability and how to use it to show consistency.
- 4. Methods of estimation
 - (a) Maximum likelihood estimators (MLE). How to compute them. How are they related to sufficient statistics?
 - (b) Why taking the logarithm makes things easier (also very useful in numerics and semidefinite programing for the computer science folks)
 - (c) Properties of MLEs? Consistency, asymptoticly unbiased, CLT.
- 5. Interval estimators
 - (a) What is a (1α) confidence interval? Why is it useful?
 - (b) Pivotal quantities. What are they? How to use them to calculate confidence intervals.
 - (c) How to use the CLT to get confidence intervals of means if the sample size is large.
 - (d) $\chi^2(n)$ distribution. What does it describe? What are it's properties? What is $\chi^2_{p,n}$? How do you find it's value?
 - (e) How to find (1α) confidence intervals for the variance in normal samples using $\chi^2(n)$.
 - (f) t-distribution. Why is it useful? What are it's properties? What is $t_{p,n}$? How do you find it's value?
 - (g) How to use the t-distribution to calculate 1α confidence intervals if the sample population is normal (and the sample size is not large)