### **Course goals:**

Successful students:

- 1. will be familiar with modern ideas from high-dimensional probability.
- 2. will be able to apply these ideas to new problems.
- 3. will master rigorous proofs underlying these results, and will be able to adapt them to new settings.

#### **Course format:**

The course will have regular lectures and optional sections. Regular attendance to the lectures is strongly encouraged.

### **Typical enrollees:**

Probability at the level of STAT 210, STAT 212 and real analysis at the level of MATH 112 are required. This is an advanced graduate course, so we will assume a certain degree of mathematical maturity.

### When is course typically offered?

This course will be offered in the Fall semester.

### What can students expect from you as an instructor?

I like to create an engaging classroom environment with open discussions. I welcome interruptions during lecture to clarify any doubts that might arise.

#### Assignments and grading:

The course grade will be based on Homeworks (approximately bi-weekly) and a final project. We will not have a final exam. Approximate grade contributions: Homeworks (60%)+ Project (40%).

### Sample reading list:

- 1. Roman Vershynin, High-dimensional Probability, Cambridge University Press, 2019.
- 2. Ramon Van Handel, Probability in High-dimension, 2016.
- 3. Romain Couillet, Zhenyu Liao, Random Matrix Methods for Machine Learning, Cambridge University Press, 2022.

# Enrollment cap, selection process, notification:

No enrollment cap for this class.

# **Tentative syllabus:**

Here is a tentative syllabus for the class.

# Absence and late work policies:

We strongly encourage regular attendance for the lectures. Students will have 2 late days for the homework submissions that they can use through the semester (without any prior approval).