BSTA 787: Methods for Statistical Genetics and Genomics in Complex Human Diseases Spring 2023

- **Description:** This is an introductory course for graduate students in Biostatistics, Epidemiology, Computational Biology and other BGS disciplines, which will cover statistical methods for the analysis of genetic and genomic data. Topics covered include linkage analysis, association analysis, haplotype analysis, genome-wide association studies (GWAS), expression quantitative trait locus analysis (eQTL) and allele-specific expression (ASE) analyses, analysis of copy number variations, next-generation sequencing data (DNA), analysis of bulk RNA-seq data, singlecell RNA-seq data, single-cell multi-omics data integration, and spatial transcriptomics data. Students will be exposed to the latest methodology and computer tools in statistical genetics and genomics.
- **Prerequisites:** Introductory graduate-level courses in statistics or biostatistics (e.g., BSTA 630-631, EPID 520-521), or permission of the instructor.

Instructors: Rui Xiao, PhD, <u>rxiao@pennmedicine.upenn.edu</u>, <u>xiaor@chop.edu</u>, 206 Blockley Hall Mingyao Li, PhD, <u>mingyao@pennmedicine.upenn.edu</u>, 213 Blockley Hall

Lecture Time: 12:00pm-1:30pm Tuesday and Thursday

Location: 701 Blockley Hall

Office Hour: by appointment

Evaluation:

- 1. <u>Two homework assignments (20% of the final grade)</u>. Submit your homework on Canvas in word or pdf. Late homework is NOT accepted.
- 2. One paper presentation (25% of the final grade). Students will form 2-person teams and each team will present one paper from the list provided by the instructors. The presenters are expected to clearly explain the content of the assigned paper and lead discussions (about 75 minutes) during class. All non-presenters are required to write down their questions and submit the questions to the instructor before the presentation starts. Non-presenters are also required to participate in the discussion when their questions are discussed.
- <u>Final project (30% of the final grade).</u> Students will submit a written report (at least 2,000 words in the format of a scientific article with Abstract, Introduction, Methods, Results, and Discussion) at the end of the semester. Students can 1) analyze real data using existing methods or new methods proposed by yourself, or 2) evaluate existing methods by simulations and applications to real data, or 3) do a comprehensive literature review on a specific topic approved by the instructors. You are welcome to discuss your project with the instructors. The abstract of the project is due on April 7, 2023. The final project is due on May 5, 2023. Each student will do a 10-minute presentation about their final project at the end of the semester.
- 4. Class participation and paper presentation discussion (15% of the final grade).
- 5. Lecture attendance (10% of the final grade).

Recommended (not Required) Textbook:

- 1. A Statistical Approach to Genetic Epidemiology: Concepts and Applications. Andreas Ziegler, Inke R. Koenig (2006).
- 2. Handbook on Analyzing Human Genetic Data: Computational Approaches and Software. Shili Lin, Hongyu Zhao (2010). <u>http://www.springerlink.com/content/978-3-540-69263-8</u>
- 3. Statistical Analysis of Next Generation Sequencing Data. Somnath Datta, Daniel Nettleton (2014). http://www.springer.com/us/book/9783319072111.
- 4. Computational Methods for Single-Cell Data Analysis. Guo-Cheng Yuan (2019). https://www.springer.com/gp/book/9781493990566.
- 5. Museum of Spatial Transcriptomics. Lambda Moses, Lior Pachter (2021). https://www.biorxiv.org/content/10.1101/2021.05.11.443152v2

BSTA 787 2023 Spring Course Schedule

#	Date	Content	Instructor
		1 st Half focusing on statistical genetics	
1	1/12	Introduction	Xiao
2	1/17	Linkage analysis	Xiao
3	1/19	Basics of genetic association studies + population stratification	Xiao
4	1/24	Haplotype analysis	Xiao
5	1/26	GWAS	Xiao
6	1/31	eQTL + allele-specific expression (ASE)	Xiao
7	2/2	Paper presentation:	Students
		Genomewide Association Study of Severe Covid-19 with Respiratory	
		Failure. New England Journal of Medicine 383(16):1522-1534 (2020)	
8	2/7	Imputation	Xiao
9	2/9	Paper presentation:	Students
		Fast and efficient QTL mapper for thousands of molecular phenotypes. Bioinformatics. 32(10): 1479-1485 (2016)	
10	2/14	Paper presentation:	Students
		The personal and clinical utility of polygenic risk scores. <i>Nat Rev Genet</i> 19:	
		581–590 (2018)	
11	2/16	Copy number variations (CNVs)	Guest
			Lecture
12	2/21	Paper presentation:	Students
		STAARpipeline: an all-in-one rare-variant tool for biobank-scale whole-	
		genome sequencing data. Nat Met 19: 1532–1533 (2022)	
13	2/23	Paper presentation:	Students
		Integrative approaches for large-scale transcriptome-wide association	
		studies. Nat Genet 48(3):245-52 (2016)	
		2 nd Half focusing on statistical genomics	
14	2/28	Intro of RNA-seq and single-cell RNA-seq	Li
15	3/2	Single-cell RNA-seq clustering and cell type annotation	Li
16	3/14	Paper presentation:	Students
		Moderated estimation of fold change and dispersion for RNA-seq data with	
		DESeq2. Genome Biology 15:550 (2014)	
17	3/16	Denoising in single-cell RNA-seq	Li
	3/21	No class	
18	3/23	Paper presentation: Joint probabilistic modeling of single-cell multi-omic	Students
		data with totalVI. Nature Methods 18:272-282 (2021)	
19	3/28	Cell-type deconvolution in RNA-seg	Li
20	3/30	Intro of spatial transcriptomics clustering and deconvolution	Li
21	4/4	Paper presentation:	Students
		Cell type-specific inference of differential expression in spatial	
		transcriptomics. Nature Methods 19:1076-1087 (2022)	
22	4/6	How to model histology information in spatial transcriptomics	Li
23	4/11	Super-resolution gene expression in spatial transcriptomics	Li
24	4/13	Paper presentation:	Students
		Robust alignment of single-cell and spatial transcriptomes with	
		CytoSPACE. BioRxiv doi: https://doi.org/10.1101/2022.05.20.488356 (2022)	
		(accepted by Nature Biotechnology)	
25	4/18	Integration of single-cell RNA-seq and spatial transcriptomics	Li
26	4/20	Paper presentation:	Students
	-	Graph deep learning for the characterization of tumor microenvironments	
		from spatial protein profiles in tissue specimens. Nature Biomedical	
		Engineering, in press (2022)	
27	4/25	Final Project Presentation	Students
28	4/27	Final Project Presentation	Students