CONTACT INFORMATION	Department of C Princeton Univer Princeton, NJ US	sity	website: willtownes.github.io email: ftownes@princeton.edu
Summary	I am a postdoctoral researcher focused on biomedical data science. I work with Barbara Engelhardt in the Princeton Computer Science Department. My interests include unsupervised machine learning, genomics, and neuroscience. My goal is to produce statistical methods and open source software implementations to help scientists achieve their research goals more effectively.		
Education	<ul> <li>Ph.D. Biostatistics, Harvard University, Cambridge, MA USA         Doctoral Committee: Rafael Irizarry (Dana-Farber Cancer Institute, Data Science)</li></ul>		
Professional			
Experience	2019 - present	Postdoctoral Researcher, Princeton Computer	er Science Department
	2011 - 2014	Software Test Engineer, SRA International, And Developed automated testing of web services and Energy Star Qualified Product Exchange. Trained APIs and version control.	d ETL processes for the
	2008 - 2010	Software Analyst, Perrin Quarles Associates, Supported development of the Environmental Provided Emissions Inventory System.	•
	2007 - 2008	Fulbright Scholar, University of the Philippine Conducted independent forest ecology research vindigenous Aeta community in Bataan Province.	while living with an

### Research In Press

[1] Hecker J, **Townes FW**, Kachroo P, Lasky-Su J, Ziniti J, Cho MH, Weiss ST, Laird NM, Lange C. A unifying framework for rare variant association testing in family-based designs, including higher criticism approaches, SKATs, and burden tests. *Bioinformatics* (2020). Preprint doi:10.1101/815290.

## Peer-reviewed Journal Articles

- [1] **Townes FW**, Carr K, Miller JW. Identifying longevity associated genes by integrating gene expression and curated annotations. *PLOS Computational Biology* (2020). doi:10.1371/journal.pcbi.1008429.
- [2] **Townes FW**, Irizarry RA. Quantile normalization of single-cell RNA-seq read counts without unique molecular identifiers. *Genome Biology* (2020). doi:10.1186/s13059-020-02078-0.
- [3] **Townes FW**, Hicks SC, Aryee MJ, Irizarry RA. Feature selection and dimension reduction for single-cell RNA-Seq based on a multinomial model. *Genome Biology* (2019). doi:10.1186/s13059-019-1861-6.
- [4] Marsh DM, **Townes FW**, Cotter K, Farroni K, McCreary K, Petry R, Tilghman J. Thermal Preference and Species Range in Mountaintop Salamanders and Their Widespread Competitors. *Journal of Herpetology* (2019). doi:10.1670/18-110.

- [5] Hicks SC, **Townes FW**, Teng M, Irizarry RA. Missing Data and Technical Variability in Single-Cell RNA-Sequencing Experiments. *Biostatistics* (2018). doi:10.1101/025528.
- [6] Hecker, J, Xu X, Townes FW, Fier HL, Corcoran C, Laird N, Lange C. Family-Based Tests for Associating Haplotypes With General Phenotype Data: Application to Asthma Genetics. Genetic Epidemiology (2017). doi:10.1002/gepi.22094.
- [7] Valeri L, Patterson-Lomba O, Gurmu Y, Ablorh A, Bobb J, Townes FW, Harling G. Predicting Subnational Ebola Virus Disease Epidemic Dynamics from Sociodemographic Indicators. *PLOS One* (2016). doi:10.1371/journal.pone.0163544.
- [8] **Townes W**. Seed dispersal of the genus Leea in forest patches of Bataan, Philippines. *Ecotropica* (2010).

## Technical Reports

- [1] **Townes FW**. Review of Probability Distributions for Modeling Count Data. arXiv (2020). https://arxiv.org/abs/2001.04343.
- [2] **Townes FW**. Generalized Principal Component Analysis. arXiv (2019). https://arxiv.org/abs/1907.02647.

#### Talks Invited Talks

- [1] Don't Normalize: the GLM-PCA approach to normalization. *Normjam, New York Genome Center.* November 2019.
- [2] Biostatistics and Data Science as a Career Path. Biology Department, Washington & Lee University. October 2019.

#### Contributed Talks

- [1] Quantile normalization of single-cell RNA-seq read counts without unique molecular identifiers. Joint Statistical Meetings (virtual conference). August 2020.
- [2] Dimension reduction for massive single-cell datasets. *Bioconductor Annual Meeting* (virtual conference). July 2020. slides doi:10.7490/f1000research.1118084.1
- [3] Dimension Reduction for Single Cell RNA-Seq based on a Multinomial Model. *Joint Statistical Meetings* (Denver, CO). August 2019.
- [4] Varying-Censoring Aware Matrix Factorization for Single Cell RNA-Sequencing. *Bioconductor Annual Meeting* (Boston, MA). July 2017.

#### Seminars

- [1] Primer: Generalized linear models and latent factor models. *Broad Institute Models, Inference, and Algorithms.* October 2020.
- [2] Feature Selection and Dimension Reduction for Single Cell RNA-Seq. Greene Lab, Department of Systems Pharmacology and Translational Therapeutics, Perelman School of Medicine, University of Pennsylvania. February 2019.
- [3] Unsupervised Learning for Single Cell Gene Expression. *Harvard Biostatistics Cancer Working Group*. September 2018.
- [4] Single Cell Housekeeping Genes and Normalization. Dana-Farber Cancer Institute cBio Seminar. April 2018.
- [5] Informative Missing Data in Single Cell RNA-Seq. Dana-Farber Cancer Institute Genomics Seminar. October 2016.
- [6] Family Based Association Tests for Rare Variants. Brigham & Women's Hospital Channing Network Medicine Seminar. September 2015.

#### Posters

- [1] Townes FW, Shukla C. Gene Expression Autoencoders. Harvard Biomedical Informatics 707 (Deep Learning in Healthcare). April 2018.
- [2] Townes FW, Marquez-Luna C. Mixture of Experts Analysis of Infectious Disease Outbreak Characteristics. Harvard Computer Science 281 (Advanced Machine Learning). December 2015.
- [3] Townes FW, Karaayvaz M, Gillespie S, Bernstein B, Ellisen L, Aryee M. Single Cell RNA-Seq Technical and Biological Confounders. Program in Quantitative Genomics Conference, Harvard Medical School. November 2015.

PEER REVIEW	Journals: Bioinformatics, Biostatistics, Genome Biology, NAR Genomics and Bioinformatics, Nature Communications		
Funding		NIH T32 Training Grant: Cancer Chan-Zuckerberg Foundation Travel Grant	
		Human Cell Atlas Jamboree. European Bioinformatics Institute, Hinxton, UK	
		NIH T32 Training Grant: Big Data to Knowledge	
		NSF-CBMS Travel Award: Topology, Geometry, and Statistics	
		NIH T32 Training Grant: Clinical Epidemiology of Lung Diseases	
Teaching	2018 Fall	Teaching Assistant, Applied Regression (Harvard - BST 210)	
Experience	2010 Fall	Grade homeworks, hold office hours, and teach a weekly lab section.	
	2018 Summer	Co-Instructor, StatStart (Harvard)	
	Statistics program for under-represented high school students.		
		Gave two interactive lectures on graphing data and regression.	
	2017 Fall	Teaching Assistant, Applied Bayesian Analysis (Harvard - BST 228)	
		Grade homeworks, provide solutions, and hold office hours.	
	2017 Summer	Lead Instructor, Introduction to Data Science (PARSE Ltd.)	
		Nonprofit, week-long statistics program for high school students.	
		Developed course material and gave interactive lectures using R.	
	2017 Spring	Teaching Assistant, Applied Longitudinal Analysis (Harvard - BST 226)	
		Grade homeworks and hold office hours.	
	2016 Fall	Teaching Assistant, Intro to Statistical Methods (Harvard - BST 201)	
	Grade homeworks, provide solutions, hold office hours, and teach a weekly lab section		
	2016 Summer	Project Mentor, Pipelines into Biostatistics (Harvard)	
		Program for under-represented undergraduates.	
		Guided three students on a pharmacogenomics data analysis.	
	2016 Spring	<b>Teaching Assistant</b> , Rates and Proportions (Harvard - BST 210)	
		Grade homeworks, provide solutions, hold office hours, and teach a weekly lab section.	

# SOFTWARE PACKAGES

- [1] glmpca: dimension reduction for non-normally distributed data. R package: https://cran.r-project.org/package=glmpca, python package: https://pypi.org/project/glmpca/
- [2] scry: Small-Count Analysis Methods for High-Dimensional Data. Bioconductor R package: https://bioconductor.org/packages/release/bioc/html/scry.html
- [3] quminorm: Quantile normalization of non-UMI single cell read counts. R package: https://github.com/willtownes/quminorm

## TECHNICAL SKILLS

Operating Systems: Mac OS X, Linux, Windows

Machine Learning Frameworks: Caret, Sklearn, Keras, Tensorflow, Stan, JAGS