

# DATA 140



Fall 2024

## WEEK 14 STUDY GUIDE

### The Big Picture

We study the most important joint distribution in data science. We then see how this is connected with simple regression.

- A random vector with a multivariate normal joint density has a few equivalent definitions, chief among which is that the multivariate normal vector can be represented as an invertible linear transformation of i.i.d. standard normals. Linear combinations of such a random vector are normal; multiple linear combinations are multivariate normal; pairwise uncorrelated multivariate normal variables are independent.
- Simple linear regression predicts  $Y$  as a linear function of a single  $X$ . No matter what the joint distribution of  $X$  and  $Y$ , there is always a least squares line. If  $X$  and  $Y$  are bivariate normal, this line turns out to be the best among all predictors.

### Week At a Glance

Mon 11/25	Tue 11/26	Wed 11/27	Thu 11/28	Fri 11/29
	Lecture			
Lab 8 Due				
HW 13 Due HW 14 (Due 5PM Mon 12/2)				
Take it easy	Happy Thanksgiving!			

## Reading, Practice, and Class Meetings

Book	Topic	Lectures: Prof. A.	Sections: TAs	Optional Additional Practice
Ch 23	<p><b>Multivariate Normal Vectors, contd.</b></p> <ul style="list-style-type: none"> <li>- 23.3 examines the multivariate normal joint density</li> <li>- 23.4 shows that for multivariate normal variables, being pairwise uncorrelated is equivalent to independence</li> </ul>	<p>Tuesday 11/26</p> <ul style="list-style-type: none"> <li>- Multivariate normal joint density</li> <li>- Independence</li> </ul>	None	None; focus on the homework.
Ch 24	<p><b>Simple Regression</b></p> <ul style="list-style-type: none"> <li>- 24.1 derives the equation of the regression line</li> <li>- 24.2 constructs bivariate normal random variables so that the relation between can be expressed in terms of “linear signal plus noise”</li> <li>- 24.3 looks at least-squares prediction in the context of the bivariate normal, and the connection with linear regression</li> <li>- 24.4 writes the regression equation in multiple different ways, each one illuminating a different property</li> </ul>	<ul style="list-style-type: none"> <li>- Simple regression: general case</li> <li>- Bivariate normal</li> <li>- Regression and the bivariate normal</li> </ul>		