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Today: More on 2-D and 3-D Continuous Data

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Today: Defining Contour Plots and Heat Maps  
Visualizing High-D Structure

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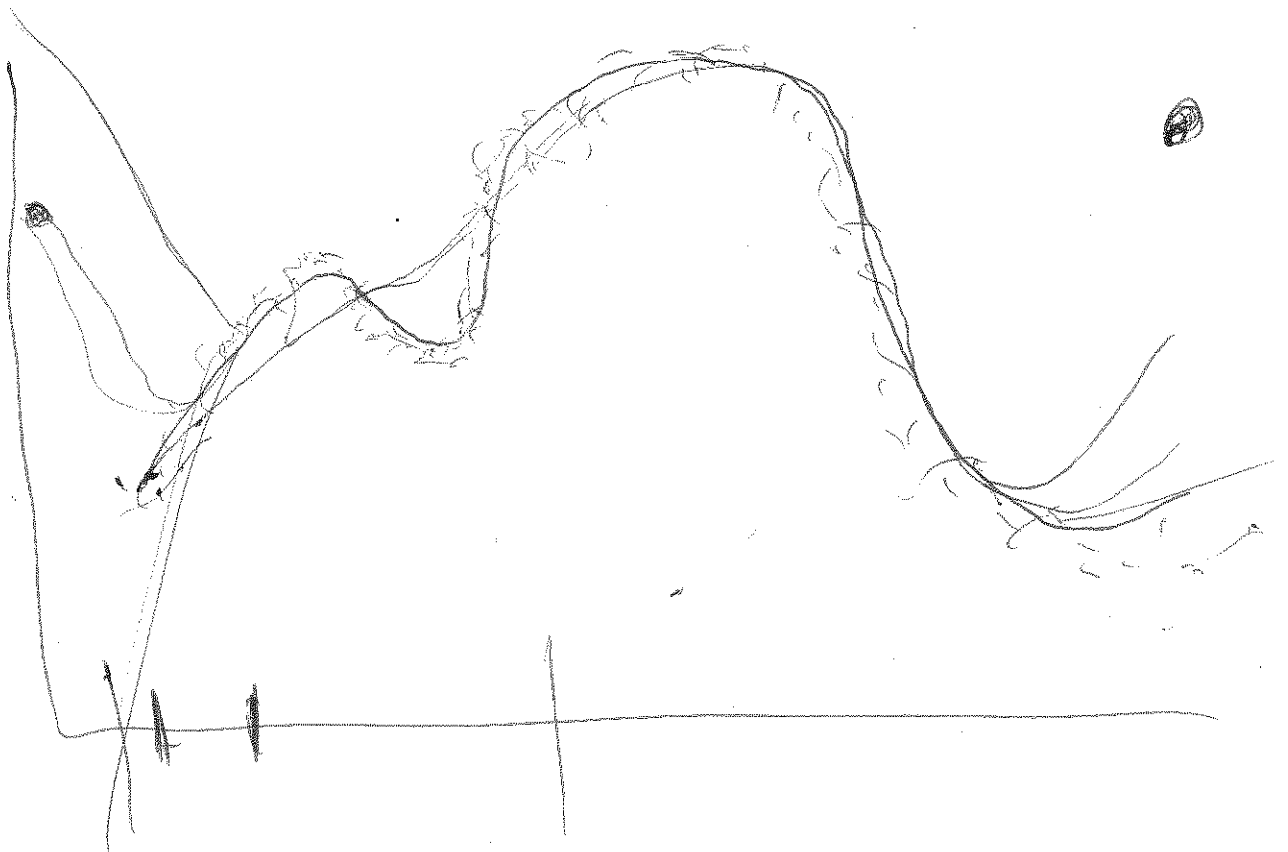
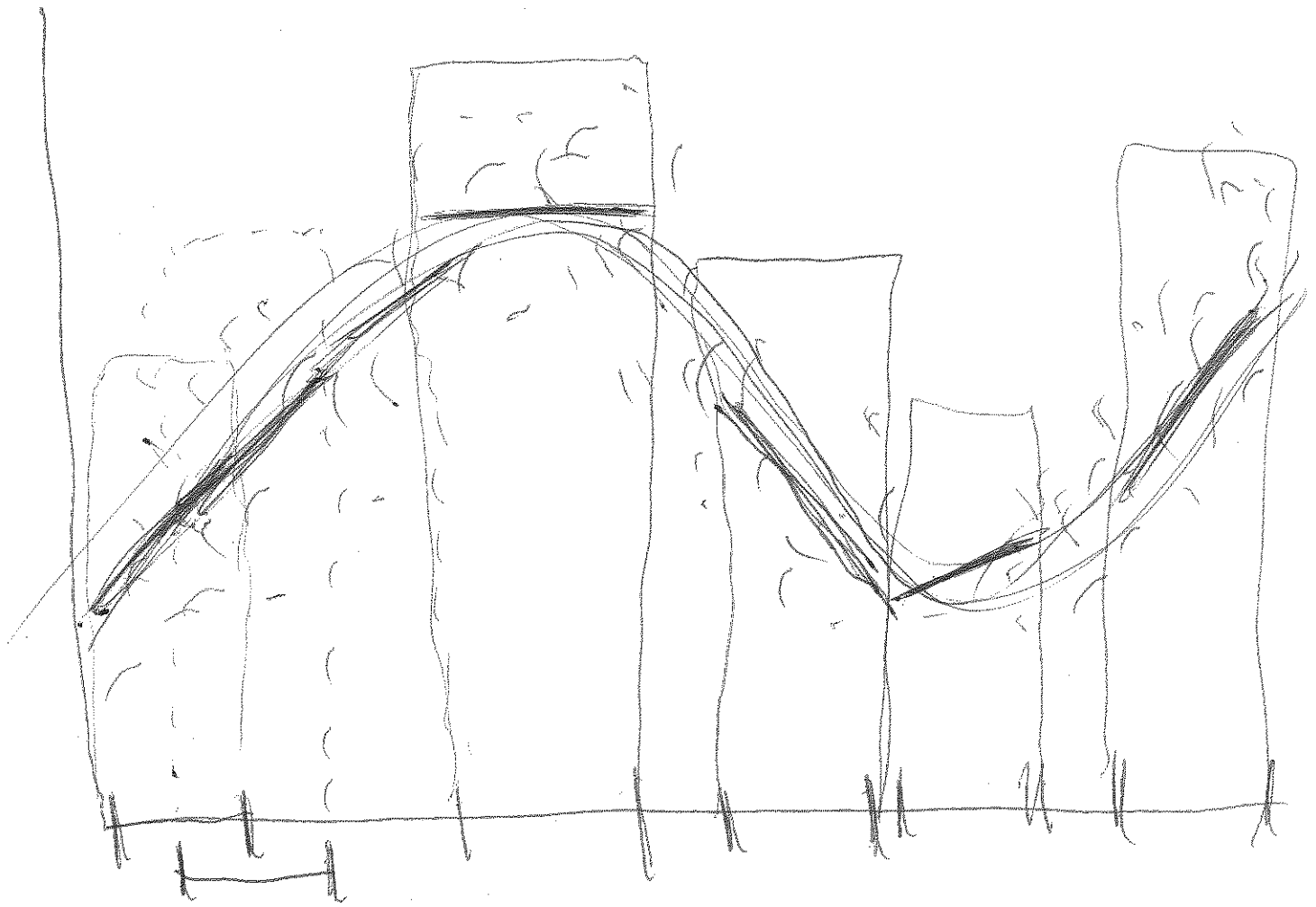
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Lab Exam, 1-D KDE, Writing about 2-D Continuous Data

Lab Exam Timeline:

- Friday: you get the data\* and the questions\*\*





$$\hat{f}_X(x) = \frac{1}{nh} \sum_{j=1}^n K\left(\frac{x - X_j}{h}\right) \rightarrow 1D \text{ KDE}$$

## 2-D Kernel Density Estimation

$$X = (X_1, X_0)$$

Goal: Estimate the joint distribution of  $X_1, X_2$ :

Assuming  $X_1$  and  $X_2$  are independent:

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$$\hat{f}(x) = \frac{1}{n \cdot h_1 \cdot h_2} \sum_{j=1}^n K_1\left(\frac{x - X_{1j}}{h_1}\right) K_2\left(\frac{x - X_{2j}}{h_2}\right)$$

↳ b/w for  $X_1$     ↳ bandwidth  
for  $X_2$

Assuming  $X_1$  and  $X_2$  are dependent:

$$\hat{f}(x) = \frac{1}{n \cdot |H|} \sum_{j=1}^n K(H^{-1}(x - x_j))$$

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$H = \begin{bmatrix} h_{11} & h_{12} \\ h_{21} & h_{22} \end{bmatrix} \rightarrow$  bandwidth matrix  
 $\approx$  covariance matrix

## Contour Plots

Level Sets:

Contour Plots:

## Heat Maps

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## Visualizing High-D Structure / Projections

What do we do when we have **many** continuous variables?

**Projections:** Sometimes we want to project the high dimensional data into a smaller subspace without losing “important structure”.

**Multi-dimensional scaling:** looks for a configuration in a  $k$ -dimensional subspace such that the distances between observations in the subspace best match the distances in the original  $p$ -dimensional space.

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Today: 2-D KDE, Contour Plots, Heat Maps,  
Distance Matrices, Dendrograms

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Wednesday: Colors (guest speaker)

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March 20, 2017

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## 2-D Kernel Density Estimation

Goal: Estimate the joint distribution of  $X_1, X_2$ :

Assuming  $X_1$  and  $X_2$  are independent:

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Contour Plots

$$X = (X_1, X_2)$$

$f_X(X) \rightarrow$  probability density function

Level Sets:

$$L(\lambda; f_X(X)) = \{X : f_X(X) > \lambda\}$$

threshold

$\rightarrow$  all areas of our feature space that have density  $> \lambda$   
 $\rightarrow$  where the observations live  
 "possible values" of our variables  $X_1, X_2$

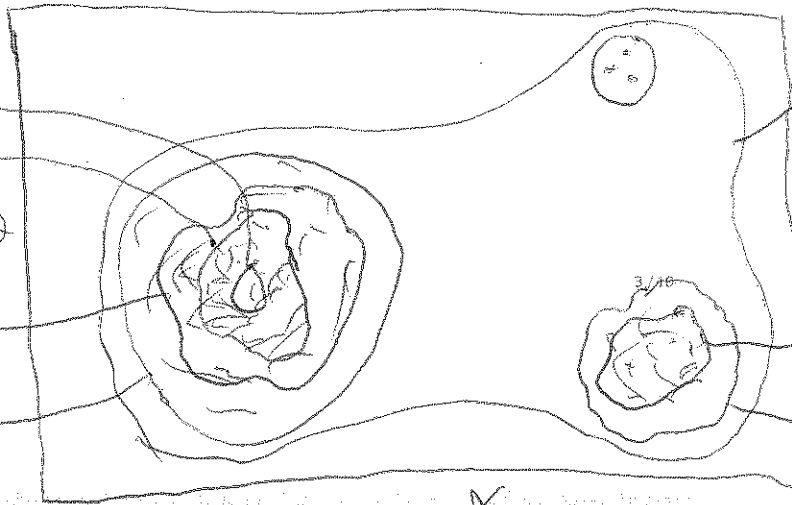
Contour Plots:

$$\lambda = \lambda_4$$

$$\lambda = \lambda_3$$

$$\lambda = \lambda_2$$

$$\lambda = \lambda_1$$



$$\lambda = \lambda_4$$

$$\lambda = \lambda_2$$

$$\lambda = \lambda_1$$

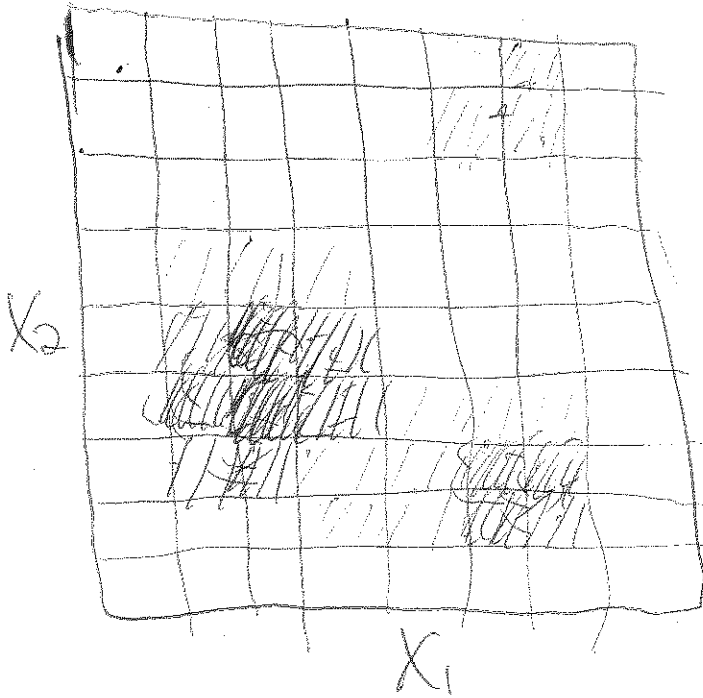
In `ggplot()`:

Heat Maps

$$\lambda_0 < \lambda_1 < \lambda_2 < \lambda_3 < \lambda_4$$

`geom_tile()`

$\rightarrow$  "birds eye view" of 2D density estimate



- Divide the feature space into a (usually 2-D) grid
- for each tile in the grid, color that tile by the average density in that space (from 2D KDE)
- Use a logical color scale/gradient to represent high/medium/low density areas