

Chapter 2: Interpretability

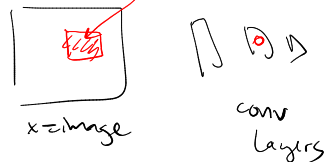
I. Visualization/Analysis (of a trained network)



At the neuron level

- pick one neuron: activities on the training set \rightarrow stats
 ex: classificⁿ task

- what it sees \rightarrow receptive field



histogram of activities



neuron: discriminative for these classes

- what does it react to?

\rightarrow display input patterns that maximize the activity
 \hookrightarrow from the training set

\rightarrow compute the pattern that would " " " "

By gradient descent:
$$\frac{\partial x_i}{\partial t} = \eta \frac{\partial \text{activity}(x_i)}{\partial x}$$
 (\uparrow gradient)

\hookrightarrow if you apply this to the full input, looking at output neurons:

e.g. classificⁿ task:



$$\frac{\partial p(\text{class } c)}{\partial \text{image } x} = \text{sensitivity of the prediction (for class } c) \text{ to the input } x$$

image variation δx

$$x' = x + \eta \frac{\partial p(\text{other class})}{\partial x} \Rightarrow \text{change completely the prediction}$$
 (not much)

\hookrightarrow adversarial examples (2014)

\hookrightarrow adversarial attacks

\hookrightarrow due to data dimension

robustifying

\hookrightarrow train with $x \in \text{training set} \mapsto l$
 $x + \delta x$ associated adversarial attack $\mapsto l$

\hookrightarrow smooth function:
$$\frac{\partial F(x)}{\partial x} = 0$$

$$\hookrightarrow \left\| \frac{\partial F}{\partial x} \right\|^2$$



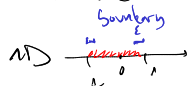
high-dim Data space

- no dense sampling (requires 10^{th} samples)

- all points are on the boundary

\hookrightarrow measure concentration

\hookrightarrow look at uniform distribⁿ in the unit-ball



proportion of points closest to the boundary

$\xrightarrow{\text{dim} \rightarrow \infty} 1$

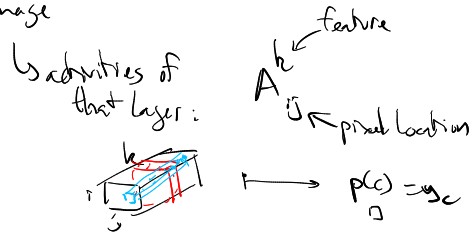
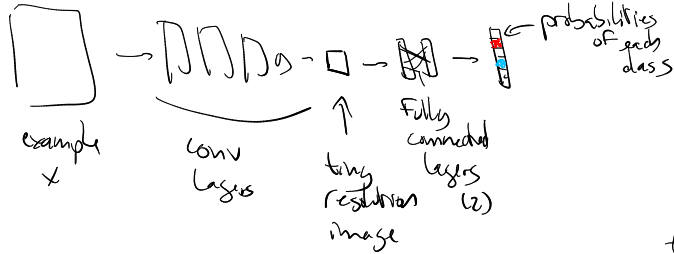
- does it have an impact? $\frac{\partial F}{\partial a}(x)$ → find which neurons influence the output the most
 ↳ activity of that neuron

⇒ Reasoning at the neuron level: compare to human brain

The case of CNN

- which parts of the image are responsible for the decision?

grad-CAM: Class Activation Maps → class probabilities



- importance of feature k for class c:

$$\alpha_k^c = \frac{1}{\#pixels} \sum_{ij} \frac{\partial y_c}{\partial A_{ij}^k}$$

- importance of one pixel (i, j):

$$\sum_k \alpha_k^c A_{ij}^k \in \mathbb{R}$$

$$- \text{ReLU}(\sum_k \alpha_k^c A_{ij}^k)$$

