

ECE 595: Machine Learning I

Lecture 8.1: Hand-Crafted and Deep Features - Convolution

Spring 2020

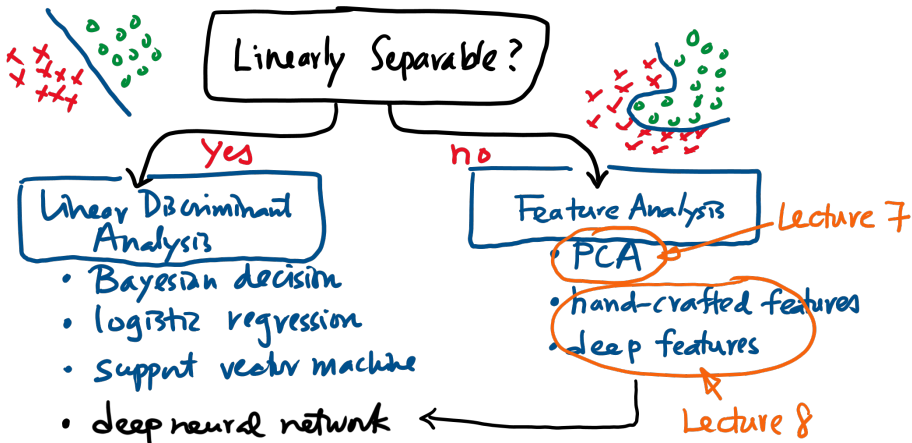
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Overview

Supervised Learning for Classification



Outline

Feature Analysis

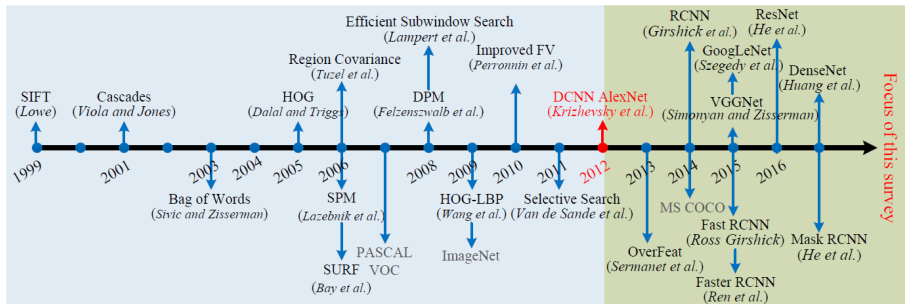
- Lecture 7 Principal Component Analysis (PCA)
- Lecture 8 Hand-Crafted and Deep Features

This Lecture

- Little History of Feature Extractions
- Convolution
 - What is convolution (if you don't know what it is yet)?
 - Some interesting facts about convolution
- SIFT and HOG
 - Gaussian derivatives
 - Pyramid
 - Histogram of oriented gradients
- Deep Features
 - What are they?
 - How to use them?

A Rough History of Feature Extraction

Deep Learning for Generic Object Detection: A Survey, <https://arxiv.org/pdf/1809.02165.pdf>



- PCA: Statistical analysis. Content agnostic.
- SIFT: Image specific. Non-training.
- Deep Features: Image specific. Training.

Convolution

Convolution (2D) is between two functions f and h :

- An **input** function $f(\mathbf{x})$, indexed by spatial coordinate $\mathbf{x} = [x_1, x_2]^T$
- A **filter** $h(\mathbf{x})$

The **output** is of the convolution is

$$\begin{aligned}g(\mathbf{x}) &= f(\mathbf{x}) * h(\mathbf{x}) \\ &= \int f(\mathbf{x} - \boldsymbol{\xi})h(\boldsymbol{\xi})d\boldsymbol{\xi}\end{aligned}$$

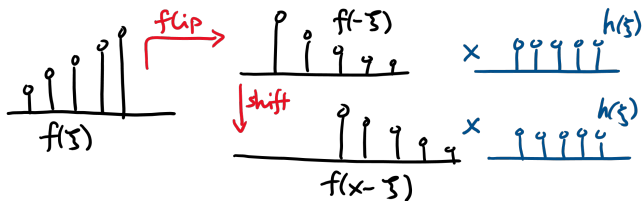
Do not be confused with **correlation**:

$$\begin{aligned}g(\mathbf{x}) &= f(\mathbf{x}) \circledast h(\mathbf{x}) \\ &= \int f(\mathbf{x} + \boldsymbol{\xi})h(\boldsymbol{\xi})d\boldsymbol{\xi}\end{aligned}$$

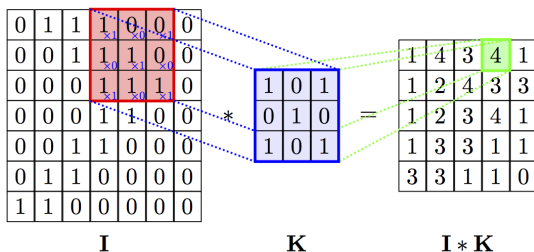
Convolution **flips** the filter, whereas correlation does not.

Pictorial Illustration

A convolution operation always involves 3 steps: flip-shift-add.



Most tutorials you see on the internet are correlations.



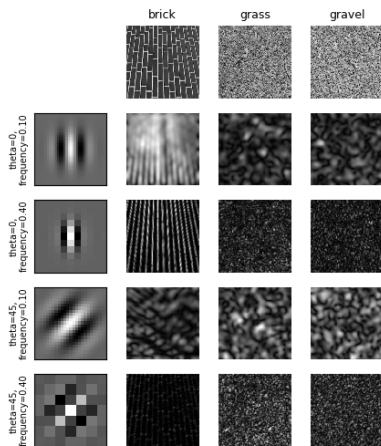
A Few Things You Need to Know about Convolution

- The core of convolution is the concept of **linear shift invariant** (LSI).
- A system \mathcal{T} is LSI if
 - $\mathcal{T}(af_1(\mathbf{x}) + bf_2(\mathbf{x})) = a\mathcal{T}(f_1(\mathbf{x})) + b\mathcal{T}(f_2(\mathbf{x}))$
 - Let $g(\mathbf{x}) = \mathcal{T}(f(\mathbf{x}))$. Then for any ξ , $f(\mathbf{x} + \xi) \mapsto g(\mathbf{x} + \xi)$.
- Convolution is the only operation that allows LSI.
- **Eigen-functions** of a convolution operation are the Fourier series.
- The flip operation is necessary to define the Fourier series.
- This can be dated back to Pierre-Simon Laplace (1749-1827) and Joseph Fourier (1768-1830), with 200 years of work in real / functional analysis.
- Convolution with large filters are always implemented by **Fast Fourier Transforms**.
- Convolution can be performed at the **speed of light!** Put a mask at the focal plane of the lens. It will give you the convolution of the mask and the image (in the Fourier domain).

<https://www.youtube.com/watch?v=4Eg0Tbk601s>

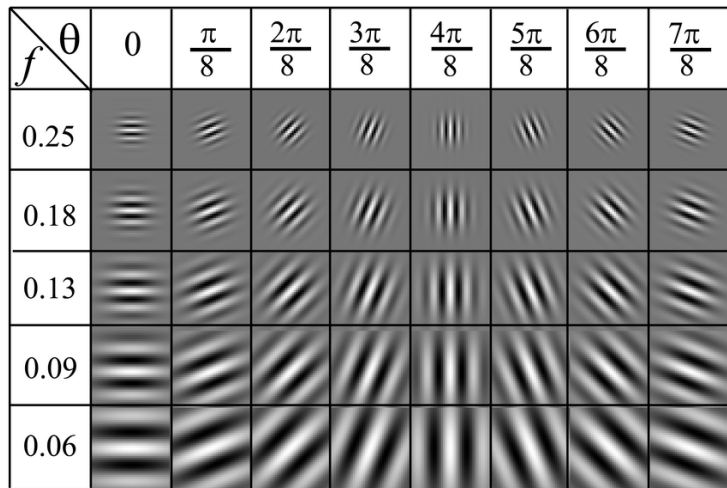
Effect of Convolution / Correlation

Image responses for Gabor filter kernels



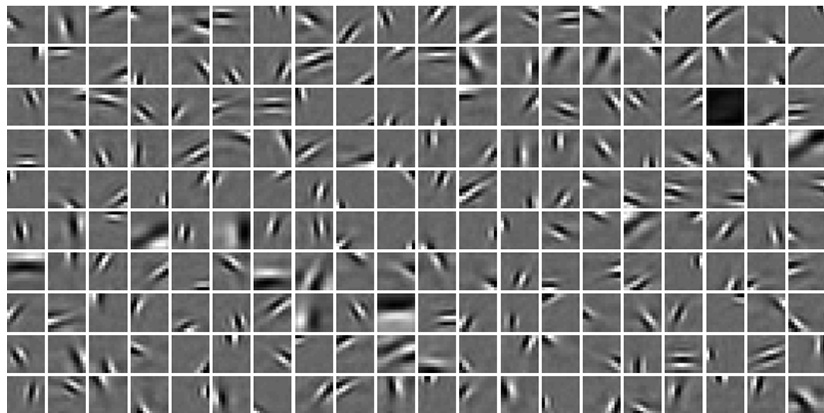
Examples of Filters

Gabor Filter



Examples of Filters

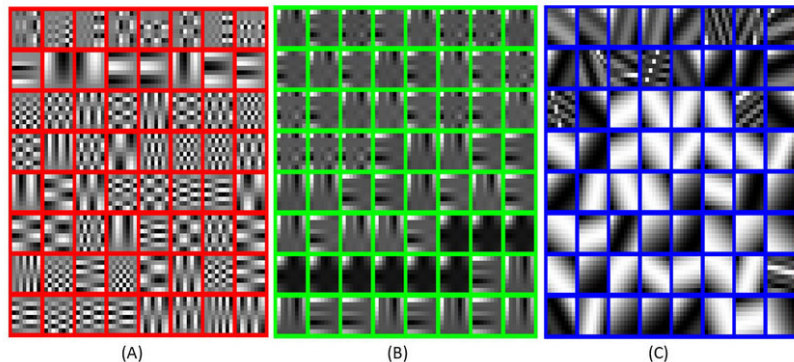
Another Gabor Filter



<https://www.quora.com/How-are-Gabor-filters-implemented-in-visual-area-V1-in-the-brain>

Examples of Filters

KSVD Filters



https://www.researchgate.net/figure/Basis-functions-used-by-a-KSVD-The-KSVD-based-dictionary-elements-or-atoms-mostly_fig6_336133323