

# Notes

## Computer Vision

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2018 January

### 1 Week 1

Gaussian Filter:  $G(x, y) = \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}}$

Edge Detection:

- Prewitt Operator:  $[[1,1,1]]$
- Sobel operator:  $[[1,2,1]]$

### 2 Week 2

histogram equalization: increase contrast

1. compute cdf at k (normalized)
2. multiply by L-1 (L, number of levels; 256), then floor
3. this is new intensity

KNN Distances:

1. L2 Distance (Euclidean distance):  $\sqrt{\sum (I_1 - I_2)^2}$
2. L1 Distance (Manhattan distance):  $\sum |I_1 - I_2|$

L1 is diamond shaped, L2 is circular

KNN computationally expensive and unable to differentiate position and intensity shift.

### 3 Week 3

for given image len x width:

- input  $x = D \times 1$  vector (flatten image)
- weight  $W = k \times D$  array
- bias  $b = k \times 1$  vector

- score  $s = k \times 1$  vector

$$s = Wx + b$$

$$s = Wx \text{ where } W=[W,b], x=[x,1]$$

- $x=(D+1) \times 1$  vector
- $W=k \times (D+1)$  array
- $s=k \times 1$  vector

Softmax classifier

probability of class  $m$  can be calculated with  $\text{softmax}(f) = \frac{e^{f_m}}{\sum_{j=1}^K e^{f_j}}$

cross-entropy loss for  $i$ th training sample:  $L_i = -\log \frac{e^{f_{y_i}}}{\sum_{j=1}^K e^{f_j}}$

assumption of cnn:

- locality of pixel dependencies
- stationary of image statistics
- translation invariance
- same filter for whole image