Spatial Linear Filtering - Why?

- Image processing
- Remove noise from images (e.g. poor transmission (from space), measurement (X-Rays))

Original Image



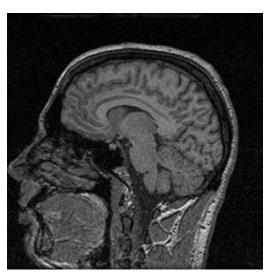
Corrupted Image



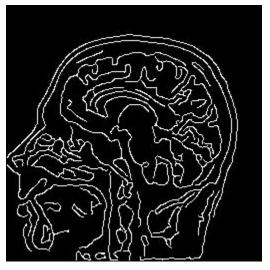
Filtered Image

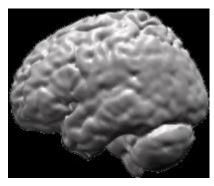


Edge Detection



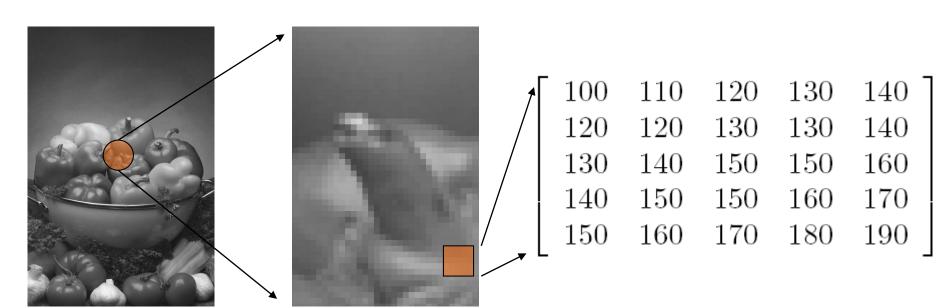
- e.g. could digitally extract brain for display
- could check if machined parts are correct





How does it work?

- Lets use a grey image for ease
- Each pixel has a single value between 0 (black) and 255 (white)



- Now select a filter kernel
- e.g. Gaussian blur (removes noise)
- For each pixel in the image

1	3	1
3	9	3
1	3	1

- Can the kernel be centred over the pixel?
- If so, calculate the sum of the products

```
140
100
     110
          120
                130
     120
          130
                130
                      140
     140
                150
          150
                     160
    150
         150
                      170
                160
                180
     160
          170
                      190
```

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1	3	1
3	9	3
1	3	1

- Can the kernel be centred over the pixel?
- If so, calculate the sum of the products

100	1 3 0	120	130	140
1 3 0	1 9 0	1 3 0	130	140
130	14 0	150	150	160
140	150	150	160	170
150	160	170	180	190

```
Sum=1x100+3x110+1x120+
3x120+9x120+3x130+
1x130+3x140+1x150
= 3080
```

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T 100	110	1 3 0	130	140
120	1 9 0	1 9 0	130	140
130	140	1 3 0	150	160
140	150	150	160	170
150	160	170	180	190

```
Sum=1x110+3x120+1x130+
3x120+9x130+3x130+
1x140+3x150+1x150
= 3260
```

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3	9	3
1	3	1

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100	110	120	130	140
120	120	13 0	1 9 0	140
130	140	150	1 3 0	160
140	150	150	160	170
150	160	170	180	190

- Now select a filter kernel
- e.g. Gaussian blur (removes noise)
- For each pixel in the image

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Γ.	100	110	120	130	140
	120	1 3 0	130	130	140
	13 0	140	130	150	160
	140	1 3 0	150	160	170
'	150	160	170	180	190
_					

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- Can the kernel be centred over the pixel?
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100	110	120	130	140
120	120	130	130	140
130		1 5 0	1.30	160
140	150	13 0	100	170
150	160	170	180	190
L				

Result

*	*	*	*	*
*	3080	3260	3390	*
*	3450	3620	3740	*
*	3720	3870	4060	*
*	*	*	*	*

Cross Correlation

In principle, given a 3×3 subimage of pixels:

$$I_{ij} = \begin{bmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & p_{33} \end{bmatrix}$$

and a 3×3 filter kernal:

$$M_{ij} = \begin{bmatrix} m_{11} & m_{12} & m_{13} \\ m_{21} & m_{22} & m_{23} \\ m_{31} & m_{32} & m_{33} \end{bmatrix}$$

operating on pixel p_{22} , the new value p'_{22} is given by:

$$p'_{22} = m_{11}p_{11} + m_{12}p_{12} + \dots + m_{32}p_{32} + m_{33}p_{33}$$

*What about the edges?

- * indicates filter could not be centred
- 3x3 = 1 pixel edge all the way around
- 5x5 = 2 pixels edge all the way around
- nxm = ? Exercise!

*	*	*	*	*
*	3080	3260	3390	*
*	3450	3620	3740	*
*	3720	3870	4060	*
*	*	*	*	*

*What about the edges?

- Reduce size of image but could be counter intuitive for users.
- Put a single colour border or pad the image beforehand.
- Use old pixel values (not sensible for e.g. edge detectors)
- Use known neighbours colour.
- Use some fudge (e.g. have some filters reduced in size so they can fit against the edges when needed)

*	*	*	*	*
*	3080	3260	3390	*
*	3450	3620	3740	*
*	3720	3870	4060	*
*	*	*	*	*
	$\overline{}$	1		

Aren't those numbers too big?

- Yes, they are no longer in the range 0-255
- We could have negative numbers (for a high pass filter)
- So we need to map the new numbers onto our range 0-255
- First calculate the max and min values
- max=4060, min=3080
 Then for each intermediate value I, calculate
 I'=((I-min)*255)/(max-min)

*	*	*	*	*
e e	0	47	81	*
*	96	141	172	*
*	167	206	255	*
*	*	*	*	*

Normalisation

- The last step:
- I'=((I-min)*255)/(max-min)
- Is called Normalisation
- As used, it maps pixels values to the range 0..255. Pixels could be mapped to other ranges (e.g. You might have a 12 bit display, or you might want to map back to the same range the image currently uses)

And colour?





Calculate sum of products for each colour channel independently Normalise as before, but replace red with new red, etc.

1140	130	150	128	130
1320	190	130	120	120
1120	130	150	140	145
147	168	165	162	160
150	170	167	167	165
4400	420	400	400	440
1120	1 3 0	130	108	110
13 0	195	837	80	78
90	3 5	82	82	81
80	82	85	86	89
92	95	100	101	102
14	3 5	15	12	13
3 2	91	13)	12	12
12	13 3	15	14	14
20	21	22	23	26
19	18	16	20	21

What about the filters?

1	1	1
1	1	1
1	1	1

-1	-1	-1
-1	8	-1
-1	-1	-1

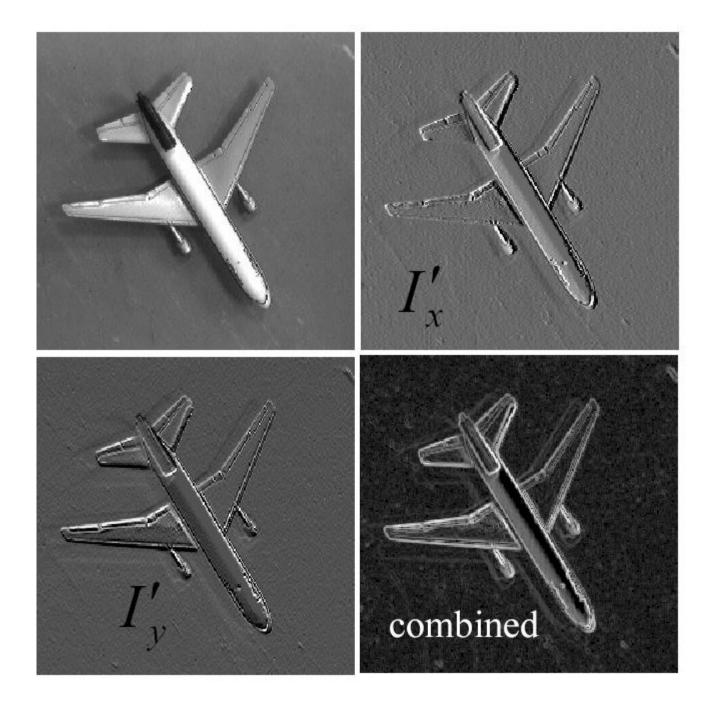
•	Low pass 3x3 box filter (nxm box
	filter has size nxm with all 1's).
	The bigger the filter, the larger the
	blur (and time - think about that)

 High pass filter (like sharpen in Photoshop

-1	0	1
-2	0	2
-1	0	1

1	2	1
0	0	0
-1	-2	-1

 Sobel edge detectors – Xdirection and Y-direction



Low Pass Filter - Gaussian

1	4	7	10	7	4	1
4	12	26	33	26	12	4
7	26	55	71	55	26	7
10	33	71	91	71	33	10
7	26	55	71	55	26	7
4	12	26	33	26	12	4
1	4	7	10	7	4	1

7x7

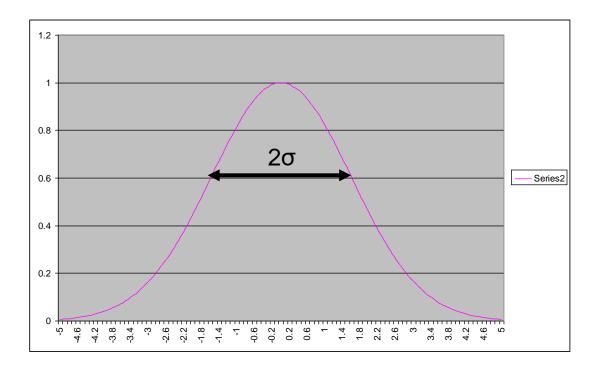
3x3

1	3	1
3	9	3
1	3	1

Where do these numbers come from?

Gaussian

- 1 dimensional
- $G(x)=e^{-(x^2/2\sigma^2)}$
- In Excel power(exp(1), $-(x*x/2*\sigma*\sigma)$)



Gaussian

- 2D $G(x,y)=e^{-((x^2+y^2)/2\sigma^2)}$

