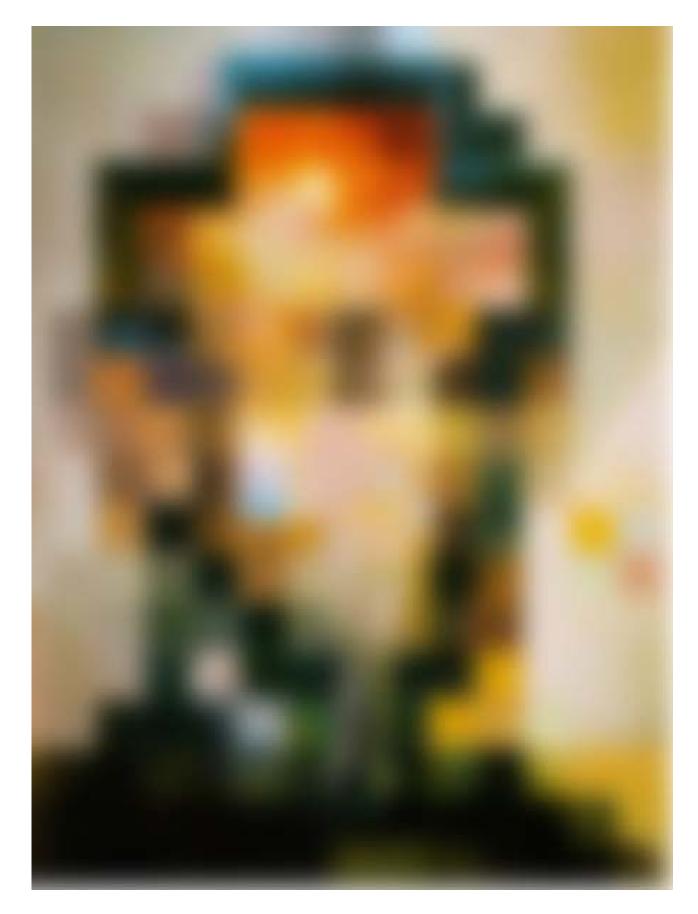
# **Image Filtering**



Salvador Dali, "Gala Contemplating the Mediterranean Sea, which at 30 meters becomes the portrait of Abraham Lincoln", 1976



Salvador Dali, "Gala Contemplating the Mediterranean Sea, which at 30 meters becomes the portrait of Abraham Lincoln", 1976

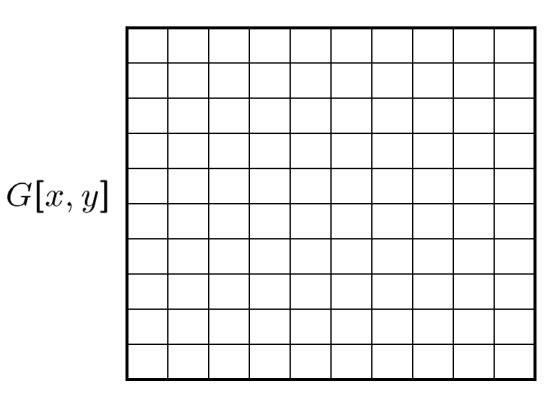
# **Filtering noise**

How can we "smooth" away noise in an image?

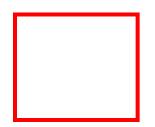
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	100	130	110	120	110	0	0
0	0	0	110	90	100	90	100	0	0
0	0	0	130	100	90	130	110	0	0
0	0	0	120	100	130	110	120	0	0
0	0	0	90	110	80	120	100	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

## **Mean filtering**

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0



F[x, y]





## **Mean filtering**

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

F[x, y]

	0	10	20	30	30	30	20	10	
	0	20	40	60	60	60	40	20	
	0	30	60	90	90	90	60	30	
	0	30	50	80	80	90	60	30	
l	0	30	50	80	80	90	60	30	
	0	20	30	50	50	60	40	20	
	10	20	30	30	30	30	20	10	
	10	10	10	0	0	0	0	0	

G[x, y]

#### **Cross-correlation filtering**

Let's write this down as an equation. Assume the averaging window is (2k+1)x(2k+1):

$$G[i,j] = \frac{1}{(2k+1)^2} \sum_{u=-k}^{k} \sum_{v=-k}^{k} F[i+u,j+v]$$

We can generalize this idea by allowing different weights for different neighboring pixels:

$$G[i,j] = \sum_{u=-k}^{k} \sum_{v=-k}^{k} H[u,v]F[i+u,j+v]$$

This is called a **cross-correlation** operation and written:

$$G = H \otimes F$$

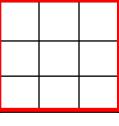
H is called the "filter," "kernel," or "mask."

The above allows negative filter indices. When you implement need to use: H[u+k,v+k] instead of H[u,v]

#### Mean kernel

What's the kernel for a 3x3 mean filter?

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
	F[x, y]								



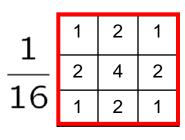
H[u, v]

When can taking an un weighted mean be bad idea?

# **Gaussian filtering**

A Gaussian kernel gives less weight to pixels further from the center of the window

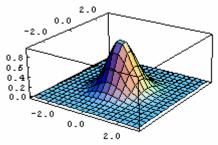
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	90	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0



H[u, v]

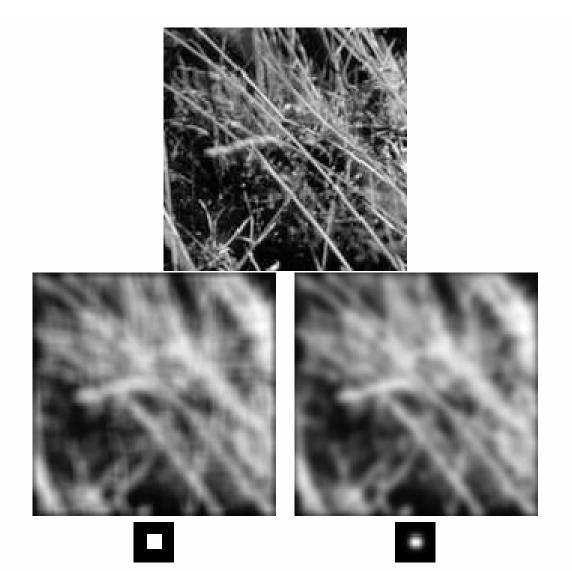
This kernel is an approximation of a Gaussian function:

$$h(u,v) = \frac{1}{2\pi\sigma^2} e^{-\frac{u^2 + v^2}{\sigma^2}}$$



What happens if you increase  $\sigma$  ?

## Mean vs. Gaussian filtering

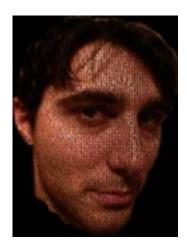


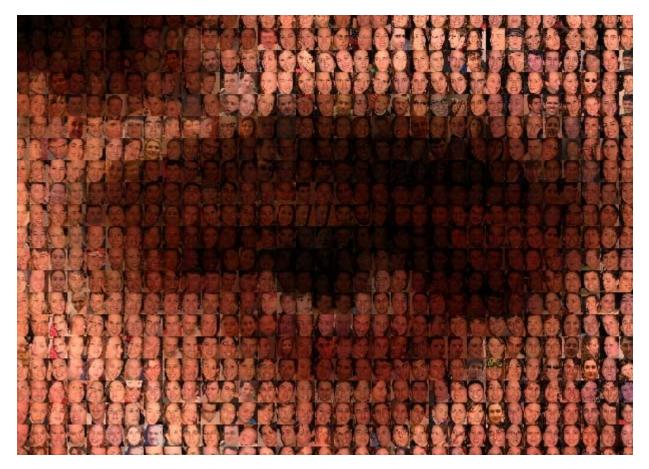
#### **Pixelation Fun**



\$	
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\$RM!!!~~~~~!!~~	~~ ~ `~~!!!M\$\$\$\$\$\$\$\$\$\$\$\$\$
\$\$\$\$\$\$\$\$\$\$\$\$\$\$\$\$M?!!!!~~~~~~	<~~~!!M\$\$\$\$\$\$\$\$\$\$\$\$\$\$
\$\$\$\$\$\$\$\$\$\$\$\$\$\$\$M!!~~~	`~~~!!!!M\$\$\$\$\$\$\$\$\$\$\$\$\$\$
\$\$\$\$\$\$\$\$\$\$\$\$\$\$RRM!~~~	~!!!M\$\$\$\$\$\$\$\$\$\$\$\$
\$\$\$\$\$\$\$\$\$\$\$\$\$MM!!!~~~`	~!!M\$\$R\$\$\$M\$R\$\$\$\$\$\$\$\$
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\$\$\$\$\$\$\$\$8M\$\$\$X!~:	~!!\$\$\$R\$\$XM\$\$\$\$\$\$\$\$
\$\$\$\$\$\$\$\$\$\$\$\$\$R!~	~~!M\$\$\$MR\$M?\$\$\$\$\$\$\$
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## **Pixelation Fun**





http://www.salle.url.edu/~ftorre/

#### Convolution

A **convolution** operation is a cross-correlation where the filter is flipped both horizontally and vertically before being applied to the image:

$$G[i,j] = \sum_{u=-k}^{k} \sum_{v=-k}^{k} H[u,v]F[i-u,j-v]$$

It is written:  $G = H \star F$ 

Suppose H is a Gaussian or mean kernel. How does convolution differ from cross-correlation?

# **Median filters**

A **Median Filter** operates over a window by selecting the median intensity in the window.

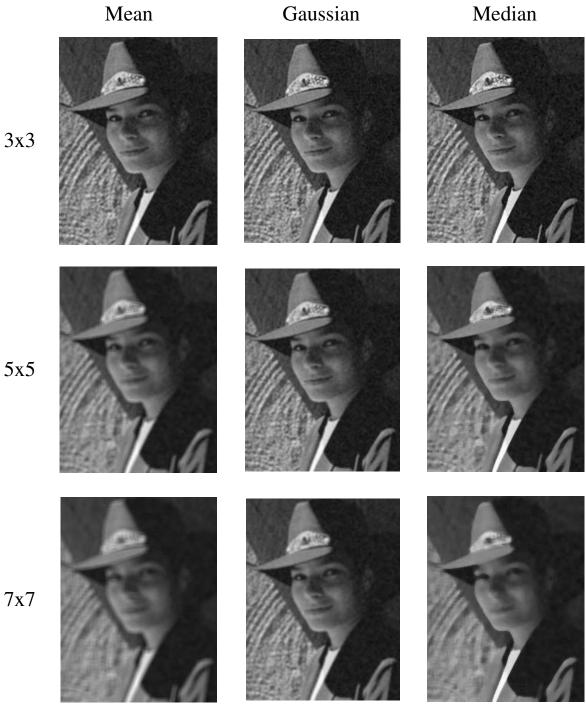
What advantage does a median filter have over a mean filter?

Is a median filter a kind of convolution?

#### **Comparison: salt and pepper noise**

Mean Gaussian Median 3x3 5x5 7x7

## **Comparison: Gaussian noise**



# **Unsharp Masking**

So, what does blurring take away?





# **Unsharp Masking (MATLAB)**

```
Imrgb = imread('file.jpg');
```

```
im = im2double(rgb2gray(imrgb));
```

```
g= fspecial('gaussian', 25,4);
```

```
imblur = conv2(im,g,'same');
```

```
imagesc([im imblur])
```

```
imagesc([im im+.4*(im-imblur)])
```