

CS4495/6495

# Introduction to Computer Vision

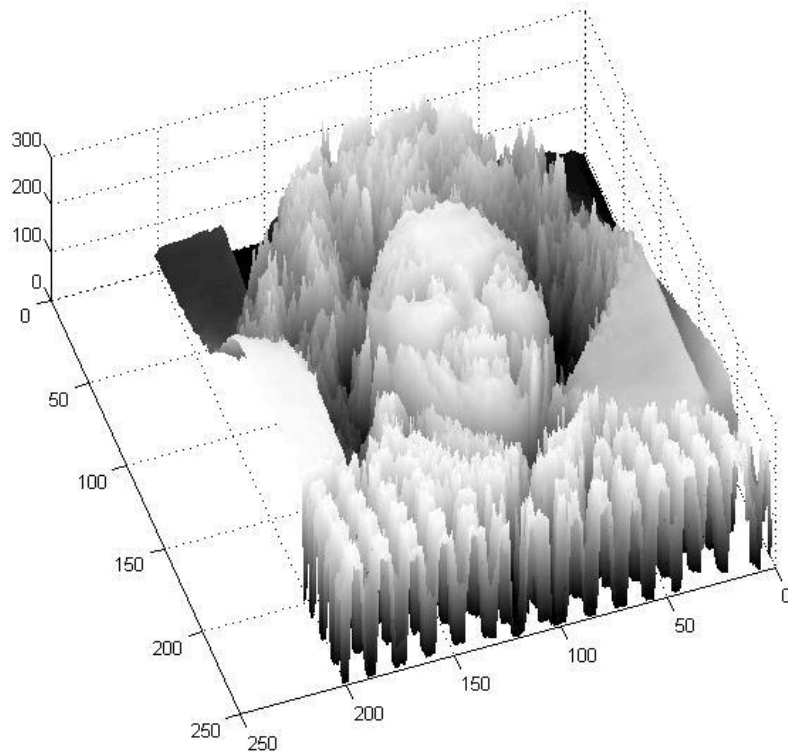
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2A-L1 *Images as functions*

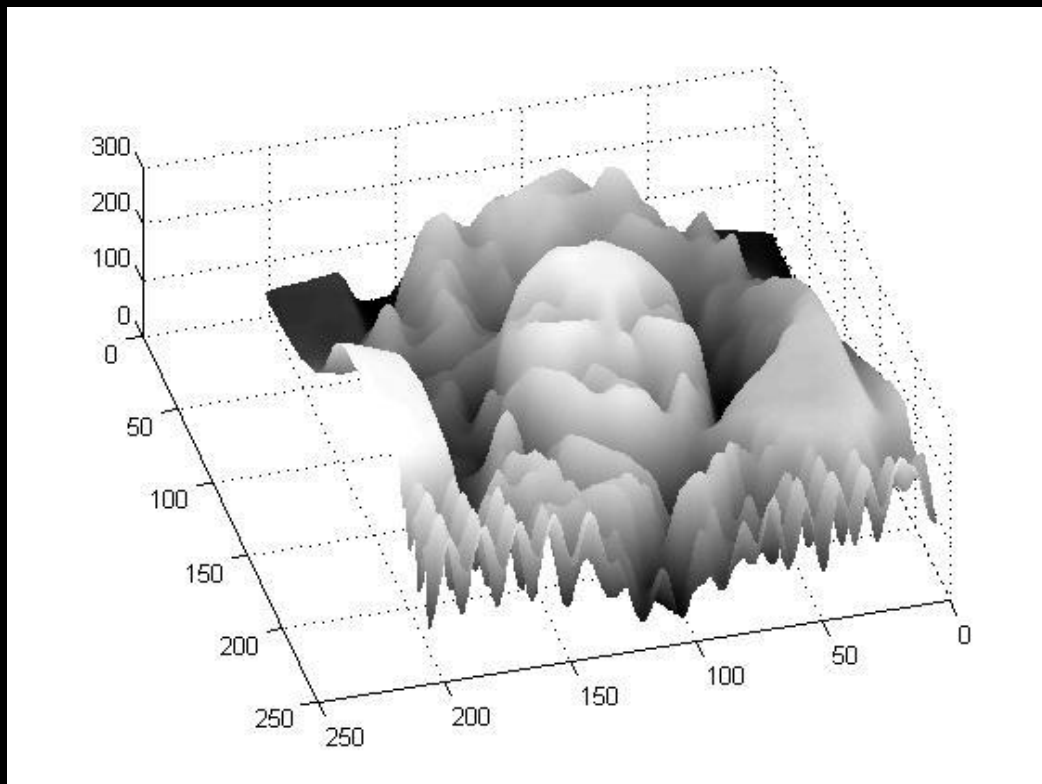
# Images as functions



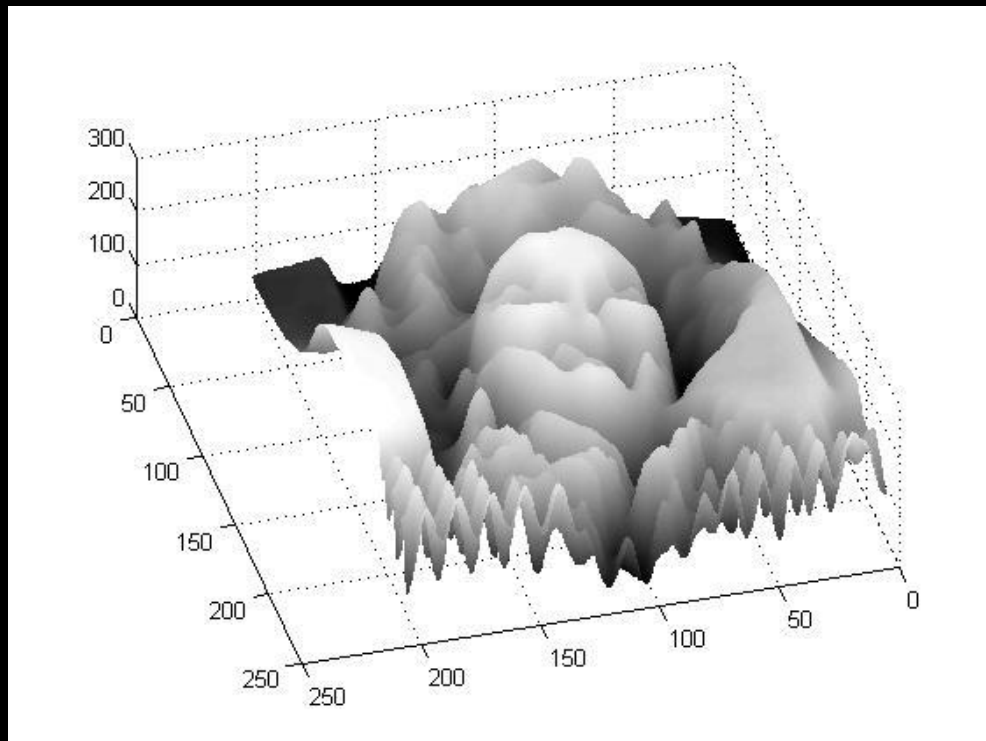
# Images as functions



# Images as functions



# Images as functions



# Quiz

An image can be thought of as:

- a) A 2-dimensional array of numbers ranging from some minimum to some maximum
- b) A function  $I$  of  $x$  and  $y$ :  $I(x, y)$
- c) Something generated by a camera.
- d) All of the above.

# Images as functions

We think of an image as a *function*,  $f$  or  $I$ , from  $R^2$  to  $R$ :

$f(x, y)$  gives the intensity or value at position  $(x, y)$

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Practically define the image over a rectangle, with a finite range:

$$f: [a, b] \times [c, d] \rightarrow [\min, \max]$$



# Color images as functions

A color image is just three functions “stacked” together. We can write this as a “vector-valued” function:

$$f(x, y) = \begin{bmatrix} r(x, y) \\ g(x, y) \\ b(x, y) \end{bmatrix}$$

# The real Phyllis

```
>> pd(40:60,30:40)
```

```
ans =
```

152	122	99	83	122	120	154	150	123	141	112
102	140	109	114	125	124	69	134	123	141	132
138	160	135	109	104	89	91	145	128	102	154
101	147	165	87	93	97	110	145	157	124	141
58	68	96	115	80	98	137	160	145	168	166
57	127	62	92	145	127	93	121	168	221	157
69	108	74	71	156	119	106	140	156	161	158
116	132	101	60	134	159	110	125	153	145	123
109	119	130	113	80	176	121	108	111	152	133
135	77	102	134	127	136	154	130	139	120	160
175	127	112	145	153	125	160	126	103	94	166
205	187	151	87	128	154	124	174	96	129	142
206	211	207	171	153	146	173	194	125	129	164
214	205	235	200	170	162	151	151	183	152	107
225	199	211	203	125	145	154	181	201	184	137
207	203	172	169	170	127	116	95	197	187	138
171	208	150	157	184	153	109	119	148	182	138
111	170	150	116	128	170	144	132	119	176	132
101	172	168	130	112	131	116	136	129	137	121
103	167	164	131	104	106	96	111	106	103	139
92	136	146	138	92	63	73	101	120	126	134

# Digital images

In computer vision we typically operate on digital (discrete) images:

*Sample* the 2D space on a regular grid

*Quantize* each sample (round to “nearest integer”)

# Digital images

Image thus represented as a *matrix* of integer values.

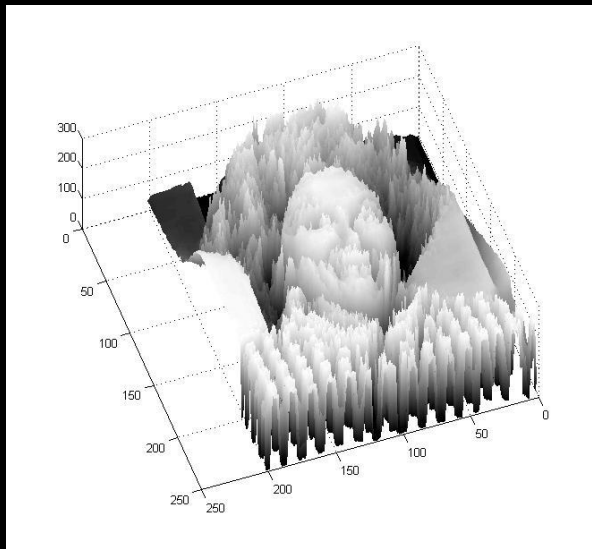
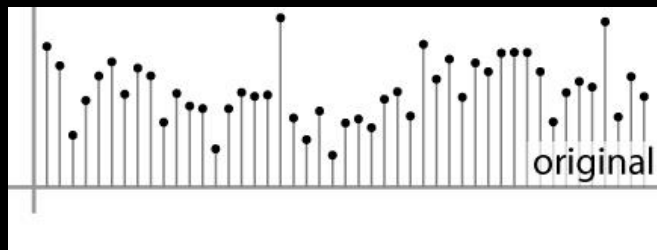
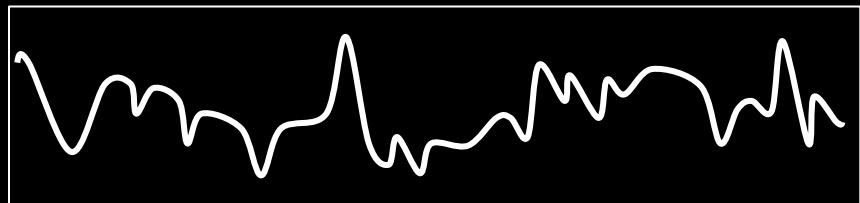


Diagram illustrating the transformation of a 3D surface plot into a 2D matrix. A green arrow points from the 3D plot to the matrix. The matrix is labeled with  $i$  (row index) and  $j$  (column index).

62	79	23	119	120	105	4	0
10	10	9	62	12	78	34	0
10	58	197	46	46	0	0	48
176	135	5	188	191	68	0	49
2	1	1	29	26	37	0	77
0	89	144	147	187	102	62	208
255	252	0	166	123	62	0	31
166	63	127	17	1	0	99	30

**2D**



**1D**

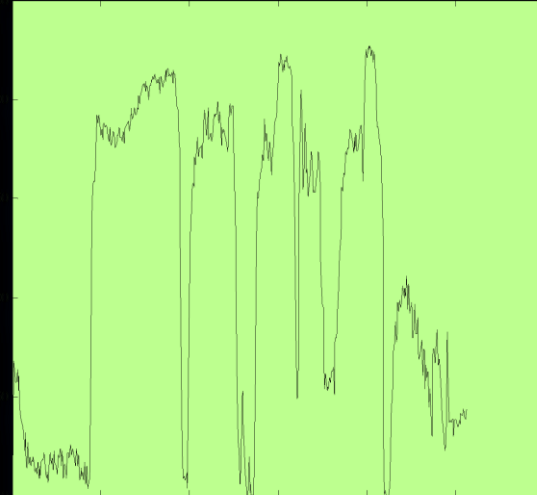
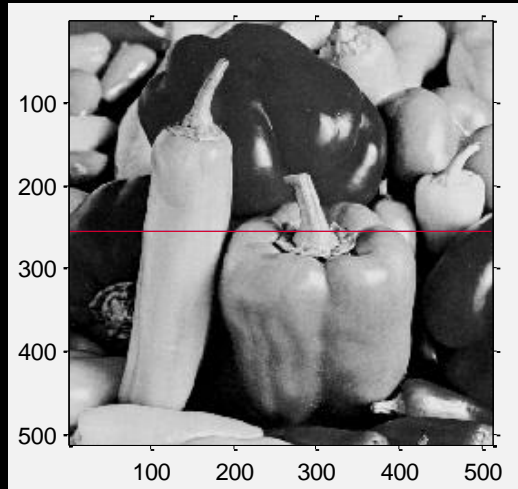
# Matlab – images are matrices

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```
>> im = imread('peppers.png'); % semicolon or many numbers  
>> imgreen = im(:, :, 2);
```

# Matlab – images are matrices

```
>> im = imread('peppers.png'); % semicolon or many numbers  
>> imgreen = im(:, :, 2);  
>> imshow(imgreen)  
>> line([1 512], [256 256], 'color', 'r')  
>> plot(imgreen(256, :));
```





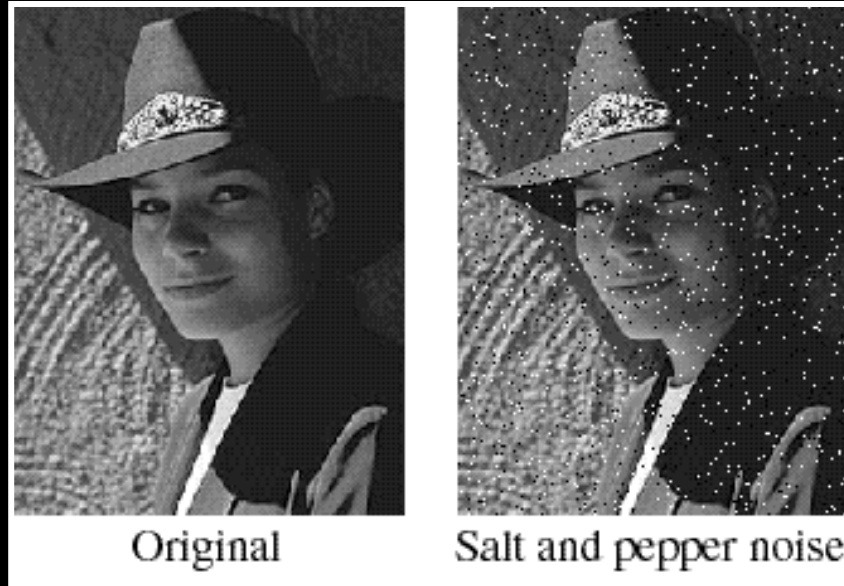
# Noise in images

- Noise is just another function that is combined with the original function to get a new – guess what – function

$$\vec{I}^{\text{c}}(x, y) = \vec{I}(x, y) + \vec{N}(x, y)$$

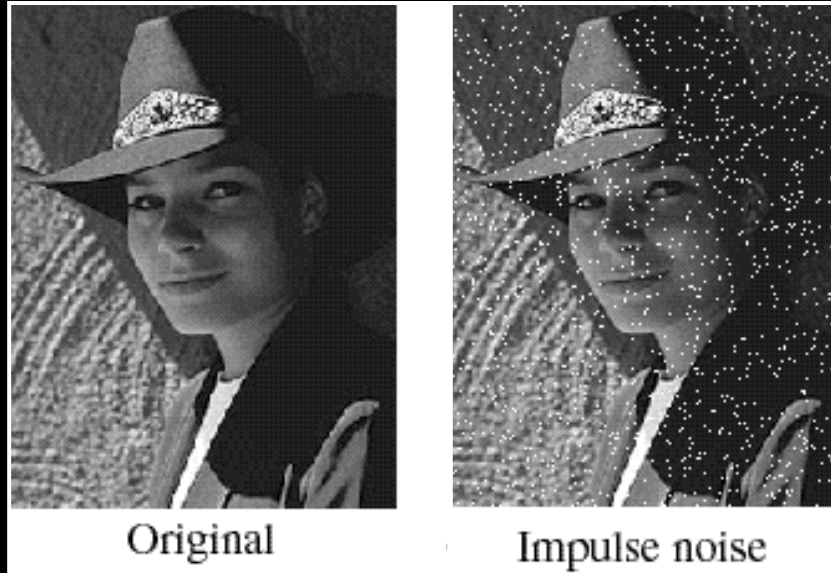
# Common Types of Noise

**Salt and pepper noise:** random occurrences of black and white pixels



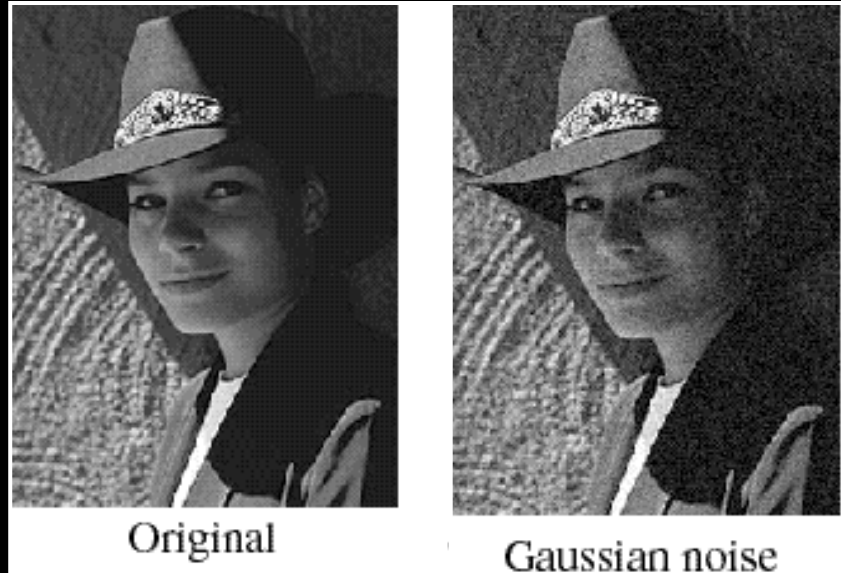
# Common Types of Noise

**Impulse noise:** random occurrences of white pixels



# Common Types of Noise

***Gaussian noise:*** variations in intensity drawn from a Gaussian normal distribution



# Gaussian noise

```
>> noise = randn(size(im)).*sigma;
```

```
>> output = im + noise;
```

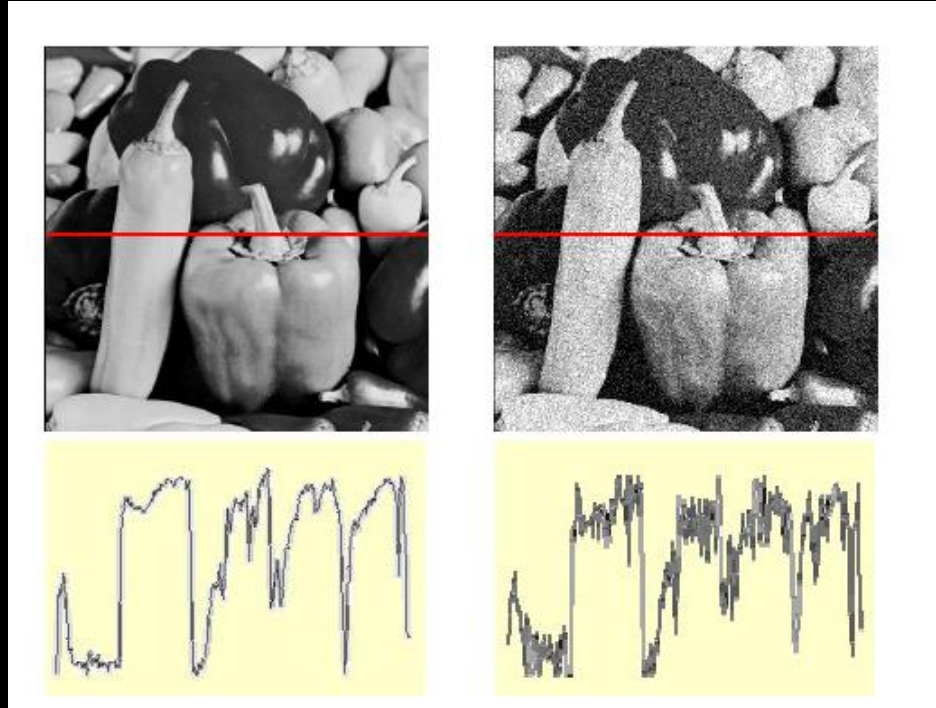


Fig: M. Hebert

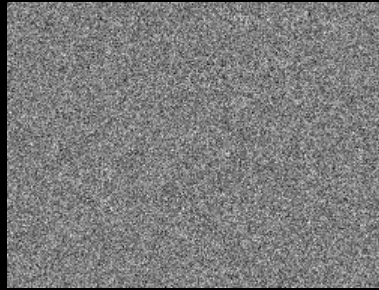
# Quiz: Effect of $\sigma$ on Gaussian noise

*Noise images:* Images showing noise values generated with different sigma

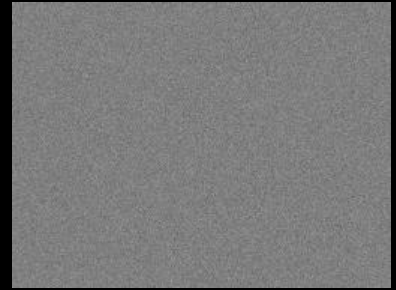
$\sigma = 2, 8, 32, 64$

Guess sigma for each noise image

```
noise = randn(size(im)).*sigma
```



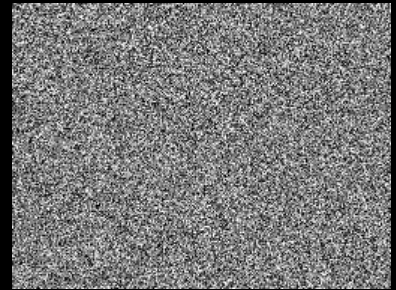
sigma =



sigma =



sigma =



sigma =

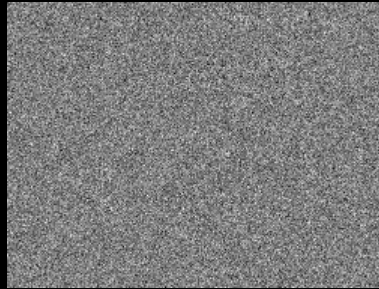
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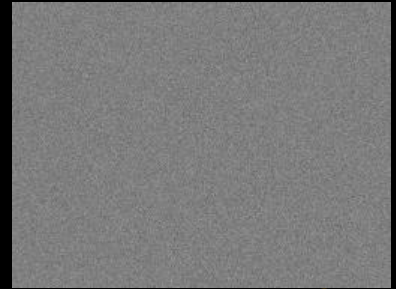
$\sigma = 2, 8, 32, 64$

Guess sigma for each noise image

```
noise = randn(size(im)).*sigma
```



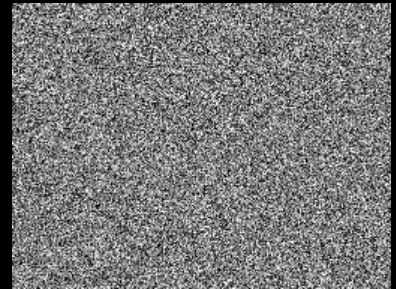
sigma = 32



sigma = 8



sigma = 2



sigma = 64

## Values of $\sigma$ to use

- A  $\sigma$  of 1.0 would be tiny if the range is [0 255] but huge if pixels went from [0.0 1.0].
- Matlab can do either and you need to be very careful - if in doubt convert to doubles.



# Displaying images in Matlab

Look at the Matlab function `imshow()`

```
imshow(im, [LOW HIGH])
```

will display the image *im* with value LOW as black and HIGH as white.

# Displaying images in Matlab

Look at the Matlab function `imshow()`

```
imshow(im, [])
```

will display the image *im* with the based on the range of pixel values in *im*.

# Quiz

When adding noise to images as arithmetic operators we have to worry about:

- a) The speed of the addition operation
- b) The magnitude of noise compared to the range of the image
- c) Whether we add the noise to the image or the image to the noise (the order of operation)
- d) None of the above