

Lecture 3

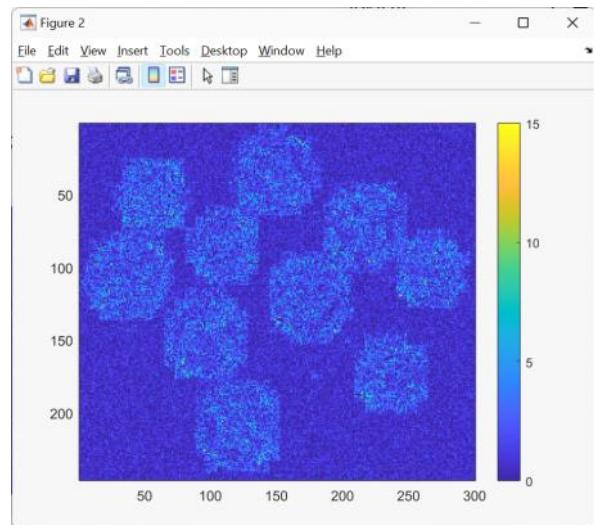
Use JPEG Compressor to compress digital images (cont'd):

```

>> I = imread('coins.png');
>> imwrite(I, 'bitstream_q90.jpg', 'quality', 90);
>> Z = imread('bitstream_q90.jpg');
>> imshow(Z)

>> Zd = double(Z);
>> Id = double(I);
>> diff = Zd - Id;
>> min(diff(:))
ans =
    -13
>> max(diff(:))
ans =
    15
>> psnr(I, Z)
ans =
    42.4284

```

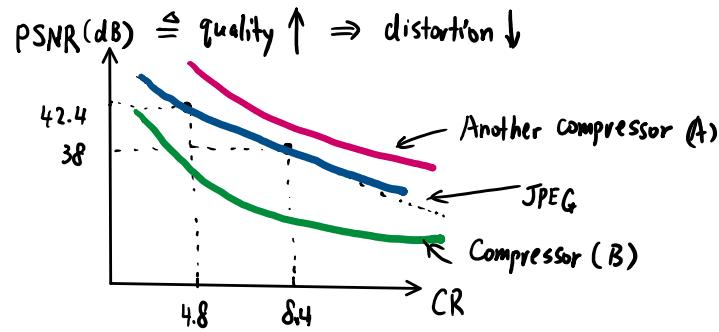
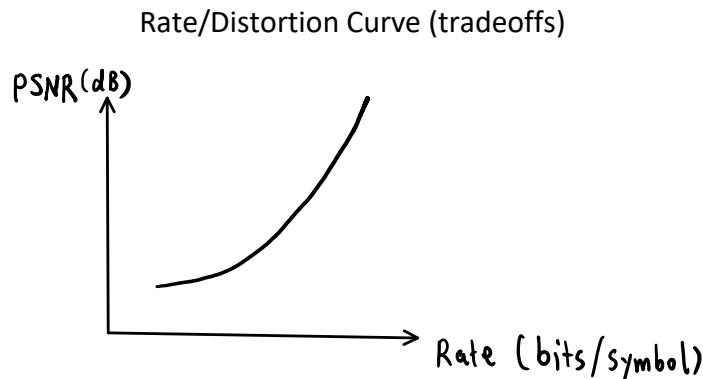


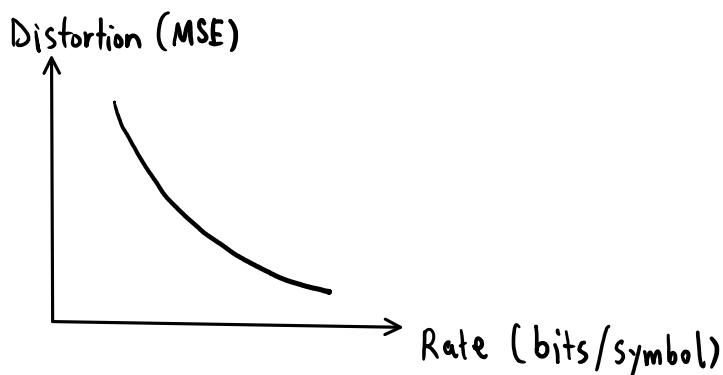
```

>> figure;
>> imagesc(abs(diff)); colorbar

```

15,269 bytes bitstream_q90.jpg => CR = 73800/15269 = 4.8333:1





- Lossless Image Compression

```

>> imwrite(I, 'bitstream_lossless.jpg', 'mode', 'lossless');
>> J = imread('bitstream_lossless.jpg');
>> isequal(I,J)
ans =
logical
1 => Lossless Compression

```

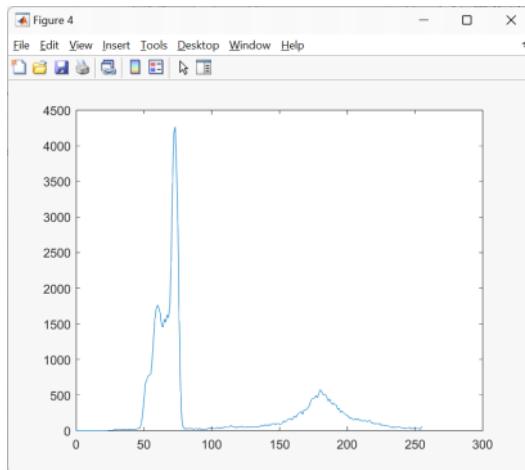
$$39,186 \text{ bitstream_lossless.jpg} \Rightarrow CR = 73800/39186 = 1.8833$$

- Histogram

```

>> [c, b] = imhist(I);
>> figure; plot(b, c)

```



```

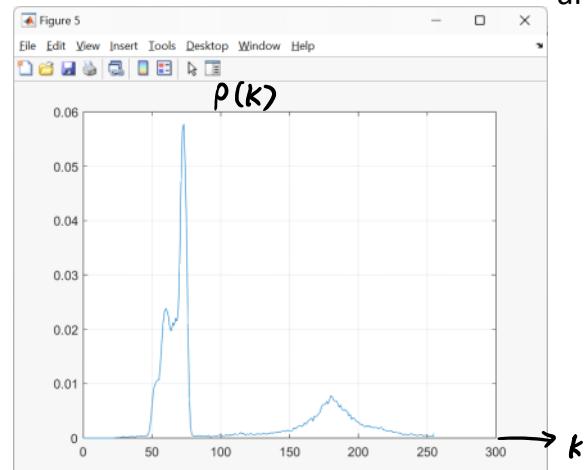
>> p = c/73800;
>> figure; plot(b, p);

```

```

>> sum(p)
ans =
1

```



- Entropy

Entropy is a statistical measure of randomness that can be used to characterize the texture of the input image.

Entropy

Entropy is defined as $-\sum(p.*\log_2(p))$, where p contains the normalized histogram counts returned from `imhist`.

$$\text{Entropy} = \underbrace{\sum_{k=0}^{255}}_{-} p(k) \cdot \log_2 p(k)$$

```
>> entropy(l)
ans =
6.3162 bits/symbol      ⇒ CR =  $\frac{8 \text{ bits/symbol}}{6.3162 \text{ bits/symbol}} = 8/6.3162 = 1.2666 : 1$ 
```

```
>> K = randi(255, 512, 512);
>> figure; imshow(uint8(K))
>> entropy(uint8(K))
```

```
ans =
```

```
7.9937
```

