

Signal to Noise Ratio

The use of **SNR** (Signal to Noise Ratio) is a confusing topic, since

- There is a range (about 10) definitions.
- Can a single number really classify **how good an image is**.
- In practice image quality is very strongly image dependant.

For signal independent additive noise,

$$f(i, j) = s(i, j) + n(i, j)$$

Signal & Noise variances are:

$$\begin{aligned}\sigma_s^2 &= \langle |s(i, j) - \langle s(i, j) \rangle|^2 \rangle \\ \sigma_n^2 &= \langle |n(i, j)|^2 \rangle\end{aligned}$$

Define SNR by

$$\text{SNR} = \frac{\sigma_s}{\sigma_n}$$

Signal to Noise Ratio I

Noting that the signal and the noise are uncorrelated, we have

$$\sigma_f^2 = \sigma_s^2 + \sigma_n^2$$

so we can write the SNR as:

$$\text{SNR} = \sqrt{\frac{\sigma_f^2}{\sigma_n^2} - 1}$$

To calculate SNR, need 2 of σ_s , σ_f , or σ_n .

Calculation of SNR for Single Image

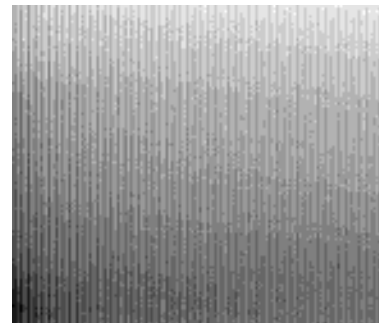
From single image can **only** find σ_f .

If **region** of image with **NO** signal can estimate σ_n from that region.

This method works for an image that contains a large region of “water” or “sky” where there is no signal.



Whole Image



Piece of Sky

Sky region shows typical CCD array fixed pattern noise.

Calculated values are $\sigma_f^2 = 5287$ and $\sigma_n^2 = 1.85$, so that

$$\text{SNR} \approx 53.4$$

Calculate SNR for Multiple Images

Assume that we have **two** realisation of same scene, so:

$$\begin{aligned} f(i, j) &= s(i, j) + n(i, j) \\ g(i, j) &= s(i, j) + m(i, j) \end{aligned}$$

which is equivalent to two image of the same scene take at different times.

The noise in each realisation have the same PDF, so that

$$\begin{aligned} \langle n(i, j) \rangle &= \langle m(i, j) \rangle = 0 \\ \langle |n(i, j)|^2 \rangle &= \langle |m(i, j)|^2 \rangle = \sigma_n^2 \end{aligned}$$

The two noise realisations were measured at *different* times, so they are uncorrelated, so:

$$\langle n(i, j) m(i, j) \rangle = 0$$

In both cases the noise is uncorrelated with the signal, so that

$$\langle s(i, j) n(i, j) \rangle = \langle s(i, j) m(i, j) \rangle = 0$$

Calculate SNR for Multiple Images I

If we then define the *Normalised* Correlation between $f(i, j)$ and $g(i, j)$ as:

$$r = \frac{\langle (fg - \langle f \rangle \langle g \rangle) \rangle}{[\langle |f - \langle f \rangle|^2 \rangle \langle |g - \langle g \rangle|^2 \rangle]^{1/2}}$$

Then by expanding, collecting terms, and cancelling all zero terms, *it-can-be-shown* that

$$r = \frac{\sigma_s^2}{\sigma_s^2 + \sigma_n^2}$$

to we can form the SNR from

$$\text{SNR} = \sqrt{\frac{r}{1-r}}$$

So allowing direct calculation of the SNR independent of the type of the signal.

Calculate SNR for Multiple Images II

If there is **more than two realisations** available a better estimate for SNR can be found by forming the normalised correlation between pairs of images and averaging.

In practice this measure of SNR **looks about right** for most images. For examples:

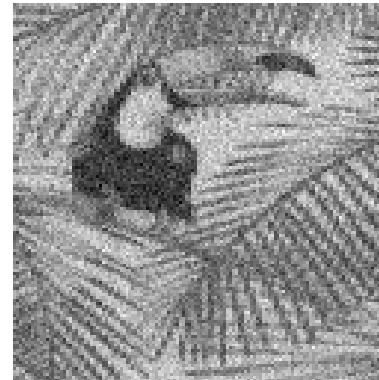
SNR	>	20	Little visible noise
SNR	\approx	10	Some noise visible
SNR	\approx	4	Noise clearly visible
SNR	\approx	2	Image severely degraded
SNR	\approx	1	Is there an image?

Digital Simulation

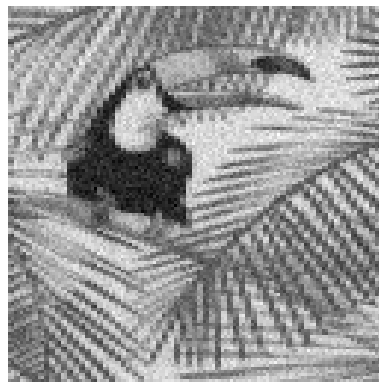
Digitally simulated “noisy” images:



SNR = 1



SNR = 2



SNR = 4



SNR = 8

Images formed by addition of Gaussian random noise.