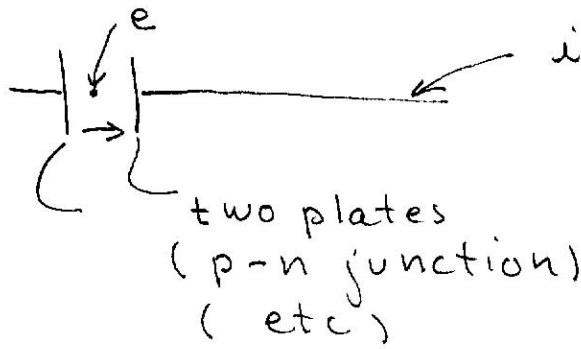


# Shot Noise Formula



$$i(t) = e \sum \delta(t - t_i)$$

↑ Random Arrival time

$$= e \int_{-\infty}^{+\infty} e^{i\omega(t-t_i)} \frac{d\omega}{2\pi}$$

Correlation function

$$\lim_{S \rightarrow \infty} \int_{-S/2}^{+S/2} i(t) i(t+T) \frac{dt}{S}$$

Fourier Transform Gives The Current Spectrum

$$F.T. = \int_{-S/2}^{+S/2} \frac{dt}{S} \int_{-\infty}^{+\infty} [e^{-i\omega T} \int_{-\infty}^{+\infty} I(\omega') e^{i\omega' t} \frac{d\omega'}{2\pi} \int_{-\infty}^{+\infty} I(\omega'') e^{i\omega''(t+T)} \frac{d\omega''}{2\pi} \times dT]$$

Integral over t gives  $2\pi \delta(\omega' + \omega'')$ , then carry out  $\int d\omega''$  leaving

$$F.T. = \int_{-S/2}^{+S/2} \frac{1}{S} [e^{-i\omega T} \int_{-\infty}^{+\infty} \frac{I(\omega') I(-\omega')}{2\pi} e^{-i\omega' T} d\omega'] dT$$

Integral over T gives  $2\pi \delta(\omega + \omega')$ . Thus

$$F.T. (\text{Correlation}) = \frac{1}{S} I(\omega) I^*(\omega); S \text{ is the}$$

time over which the correlation is measured

For single pulse  $I(\omega) = e \sum e^{-i\omega t_i}$ . For random

electron stream  $I(\omega) = e \sum_{i=0}^{\text{Number in } S \text{ sec}} e^{-i\omega t_i}$   $t_i$  - random between  $-S/2$  and  $S/2$

$I(\omega) I^*(\omega) = e^2 \sum_{i=0}^{\text{Number in } S \text{ sec}} = e^2 N$  (cross terms cancel because of randomness. Thus  $F.T. = e^2 \frac{N}{S}$

Now the average power

$$\begin{aligned}
&= \int_{-s/2}^{+s/2} i(t) i(t) dt / S \\
&= \int_{-s/2}^{+s/2} \iint_{-s/2}^{+s/2} I(\omega) I(\omega') e^{i(\omega+\omega')t} \frac{dt}{S} \frac{d\omega}{2\pi} \frac{d\omega'}{2\pi} \\
&= \int_{-s/2}^{+s/2} \int I(\omega) I(\omega') 2\pi \delta(\omega+\omega') \frac{d\omega}{S} d\omega' \\
&= \int \frac{I(\omega) I(-\omega) 2\pi d\omega}{S (2\pi)^2} = \int \frac{I(\omega) I^*(\omega) d\omega}{2\pi S}
\end{aligned}$$

Thus  $\frac{1}{2\pi}$  F.T. (Correlation) = spectral power density.

$= \frac{e^2 N}{S} \cdot \frac{1}{2\pi}$  per unit angular frequency

But  $\frac{e N}{S} = \text{Average Current} = I_{Avg}$

Changing to frequency  $\omega = 2\pi f$ , we obtain the shot noise formula  $\overline{i^2} = 2e I_{Avg} \Delta f$  (where the 2 accounts for negative frequencies)

Note: (We should be calculating

$$\int_{-s/2}^{+s/2} (i(t) - i_{Avg})^2 dt / S$$

The result is the same and the proof is included in one of the extra web pages.

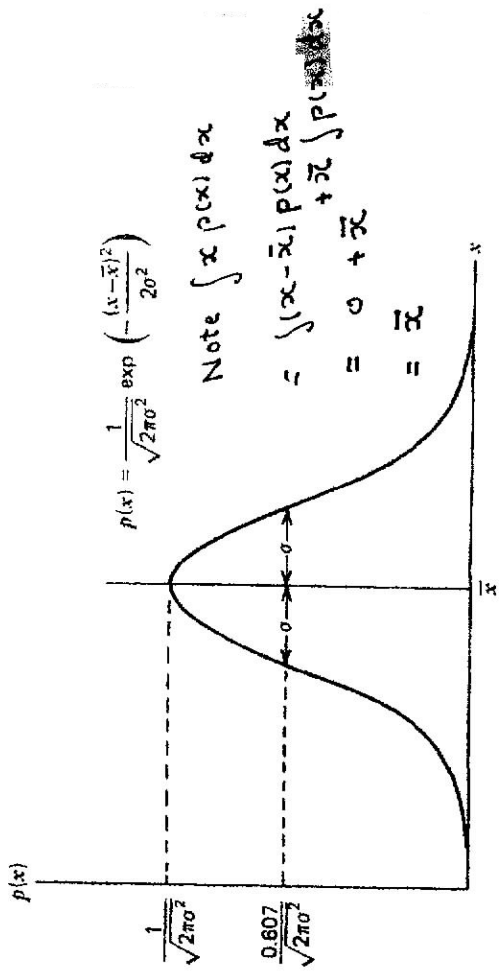


FIGURE 4.6. Gaussian probability density function.

obtained whenever a large number of independent random causes produces a cumulative effect.\* This is the case with thermal noise which is the superposition of the effects of a very large number of electrical charges in spontaneous and random motion within a conductor. Similarly, the shot noise (Section 2.6) follows the gaussian distribution since it, too, is generated by a large number of electrons emitted from the cathode at random times.

The gaussian† probability density is defined as

$$p(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\bar{x})^2}{2\sigma^2}\right) \quad (4-28)$$

and is plotted in Fig. 4.6. The terms  $\bar{x}$  and  $\sigma$  are the mean and the standard deviation defined in Section 4.4. The curve has an even symmetry about  $\bar{x}$  which implies

$$P(X \leq \bar{x}) = P(X > \bar{x}) = 0.5 \quad (4-29)$$

Since thermal noise has a zero average value,‡ we shall from now on set  $\bar{x} = 0$ , and use

$$p(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{x^2}{2\sigma^2}\right) \quad (4-30)$$

\*More formally, this follows from the central limit theorem of statistics, which states that if there are  $N$  independent random variables, and if we designate  $M$  as their sum, then the distribution of  $M$  approaches the gaussian distribution as  $M$  becomes very large [3].

†The equation was first derived by DeMoivre in 1733, but named in honor of K. F. Gauss (1777-1855) who used it while studying the statistics of errors.

‡The shot noise does have a dc component—the anode current of the tube. This is subtracted out in the analysis.

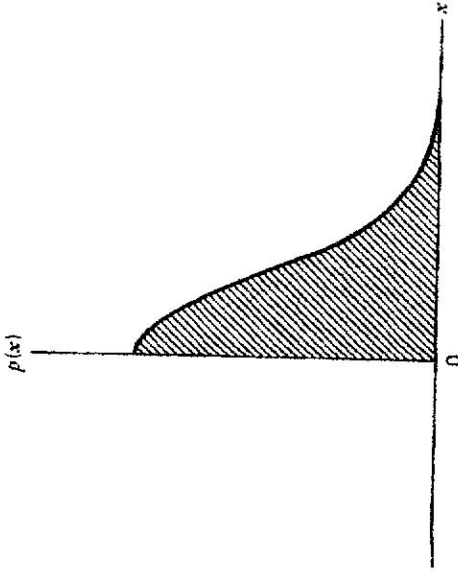


FIGURE 4.7. Idealized representation of rectified gaussian noise.

As an exercise, let us apply (4-22) to (4-30). The answer, after some algebra is quite as expected. The variance of the gaussian distribution is  $\sigma^2$ , which is why the symbol is already included in (4-28). If (4-18) is tried on (4-30), with  $\bar{x} = 0$ . Again, this is expected and not very informative. Let us, however, put gaussian-noise voltage through a half-wave linear detector. By definition means that the output contains only those noise peaks which are, say, in positive direction (see Fig. 4.7). We now have for gaussian noise:

$$\begin{aligned} \text{Average} &= \frac{1}{\sqrt{2\pi\sigma^2}} \int_0^{\infty} x \exp\left(-\frac{x^2}{2\sigma^2}\right) dx \\ &= \frac{\Gamma(1)}{\sigma\sqrt{2\pi} 2(1/2\sigma^2)} = \frac{\sigma}{\sqrt{2\pi}} \end{aligned} \quad (4-31)$$

Since the rms value of a gaussian waveform is  $\sigma$ , the ratio of the rms to average value of thermal noise is

$$(\text{Ratio})_{\text{noise}} = \frac{\text{rms}}{\text{average}} = \sqrt{2\pi} = 2.507 \quad (4-32)$$

For a pure sine wave,  $V \sin \theta$ , the rms value is  $V/\sqrt{2}$ , and the average value (for half-wave rectification) is

$$\frac{1}{2\pi} \int_0^{\pi} V \sin \theta d\theta = \frac{V}{\pi} \quad (4-33)$$

Hence

$$(\text{Ratio})_{\text{sine wave}} = \frac{\pi}{\sqrt{2}} = 2.221 \quad (4-34)$$

which is different by  $20 \log(2.507/2.221) = 1.05$  dB. The significance of this will be seen in Chapter 11.

Returning to (4-30), we note that (4-15) and (4-16) cannot be applied directly because the gaussian function (4-30) is not integrable in closed form. The expression for the gaussian PDF is, however, related to the error function defined as

$$\operatorname{erf}(u) \equiv \frac{2}{\sqrt{\pi}} \int_0^u e^{-\lambda^2} d\lambda \quad (4-35)$$

and to the complementary error function

$$\operatorname{erfc}(u) \equiv \frac{2}{\sqrt{\pi}} \int_u^\infty e^{-\lambda^2} d\lambda = 1 - \operatorname{erf}(u) \quad (4-36)$$

Using these functions it can be shown [4] that the probability  $P(X \leq x)$  is given by

$$\begin{aligned} P(X \leq x) &= 1 - \frac{1}{2} \operatorname{erfc}\left(\frac{x}{\sqrt{2}\sigma}\right) \\ &= \frac{1}{2} \left[ 1 + \operatorname{erf}\left(\frac{x}{\sqrt{2}\sigma}\right) \right] \end{aligned} \quad (4-37)$$

for  $x \geq 0$ . For  $x \leq 0$ , the proper expression is

$$P(X \leq x) = \frac{1}{2} \operatorname{erfc}\left(\frac{|x|}{\sqrt{2}\sigma}\right) \quad (4-38)$$

A short table of the  $\operatorname{erfc}(x)$  is included in Appendix A.

A frequent practical problem is to calculate the probability of an outcome falling within a given range centered around the mean (Fig. 4.8). In particular,

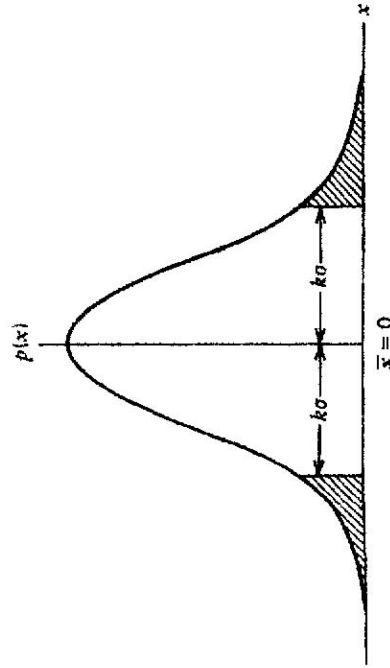


FIGURE 4.8. Definition of  $\pm k\sigma$  limits.

TABLE 4.2. Probability of an Outcome Falling Within  $\pm k\sigma$

| $k$ | $P_{\pm k\sigma}$ |
|-----|-------------------|
| 0.5 | 0.383             |
| 1.0 | 0.683             |
| 1.5 | 0.866             |
| 2.0 | 0.955             |
| 2.5 | 0.988             |
| 3.0 | 0.997             |
| 3.5 | 0.9995            |
| 4.0 | 0.99994           |

a commonly encountered range is  $\pm k\sigma$ , where  $k$  is some positive integer. Thus seek the probability  $P(-k\sigma \leq X \leq +k\sigma)$ . This is calculated most easily by utilizing the symmetry of the gaussian PDF (see Fig. 4.6) and excluding two areas outside the  $\pm k\sigma$  limits:

$$\begin{aligned} P_{\pm k\sigma} &= P(-k\sigma \leq X \leq +k\sigma) \\ &= 1 - 2P(X > k\sigma) = 1 - 2 \int_{k\sigma}^\infty \frac{\exp(-x^2/2\sigma^2)}{\sqrt{2\pi\sigma^2}} dx \end{aligned} \quad (4-39)$$

Now make a change in variable,  $u = x/\sqrt{2}\sigma$  and  $du = dx/\sqrt{2}\sigma$ . Then

$$P_{\pm k\sigma} = 1 - 2 \int_{k/\sqrt{2}}^\infty \frac{2e^{-u^2}}{\sqrt{\pi}} du = 1 - \operatorname{erfc}\left(\frac{k}{\sqrt{2}}\right) \quad (4-40)$$

Table 4.2 shows  $P_{\pm k\sigma}$  for several values of  $k$ . We can see from the table that the familiar  $\pm 3\sigma$  limit does indeed include nearly all of the possible outcomes.

#### 4.6. PEAK FACTOR

A useful application of the preceding material is the evaluation of the peak factor for gaussian noise. For a sinusoidal wave,  $V_p \cos \omega t$ , the peak factor is defined as the ratio

$$\frac{\text{peak value}}{\text{rms value}} = \frac{V_p}{V_p/\sqrt{2}} = \sqrt{2} \quad (4-41)$$

is (in decibels)  $20 \log \sqrt{2} = 3$  dB.

# Example Binomial Distribution For

$m = 6$   $p = \frac{1}{2}$  (Random Walk.  $\leftrightarrow$  Phase Fluctuations in a Laser)

$$P_m(n) = \binom{m}{n} p^n (1-p)^{n'}$$

As discussed in class  $n$  = number of steps to the right  
 $n'$  = number of steps to the left.

Let  $N = n - n'$  be the net no of steps to the right after  $m$  steps.

Also have  $m = n + n'$

Thus eliminating  $n$  and  $n'$  above we obtain

$$P_m(N) = \binom{m}{\frac{N+m}{2}} (p)^{\frac{N+m}{2}} (1-p)^{\frac{m-N}{2}} = \frac{m!}{\left(\frac{N+m}{2}\right)! \left(\frac{m-N}{2}\right)!} p^{\frac{N+m}{2}} (1-p)^{\frac{m-N}{2}}$$

Numbers for  $m = 6$

Prob Dist<sup>n</sup> of  $P_m(N)$  Net No to Right

Probability of  $P_m(N)$  No to the right

Notes

| $n$ | $n'$ | $n - n' = N$ | $P_m(N)$                                       | Probability  | Notes  |
|-----|------|--------------|--|--|--|
| 6   | 0    | 6            | $P_6(6) = \frac{6!}{(6!)^2} = \frac{1}{64}$    | $P_6(0) = \left(\frac{1}{2}\right)^6 = \frac{1}{64}$ | $2 \left( \sum P_m(N) \right)_{2,4,6} + P_m(0) = 1$  |
| 5   | 1    | 4            | $P_6(4) = \frac{6!}{(4!)^2} = \frac{15}{64}$   | $P_6(1) = \frac{6}{64}$                              |  |
| 4   | 2    | 2            | $P_6(2) = \frac{6!}{(2!)^2} = \frac{15}{64}$   | $P_6(2) = \frac{15}{64}$                             | $\sum_{n=0}^6 P_m(n) = 1$  |
| 3   | 3    | 0            | $P_6(0) = \frac{6!}{(3!)^2} = \frac{20}{64}$   | $P_6(3) = \frac{20}{64}$                             |  |
| 2   | 4    | -2           | $P_6(2) = \frac{6!}{(4!)(2!)} = \frac{15}{64}$ | $P_6(4) = \frac{15}{64}$                             | Review Questions!  |
| 1   | 5    | -4           | $P_6(4) = \frac{6!}{(5!)(1!)} = \frac{6}{64}$  | $P_6(5) = \frac{6}{64}$                              |  |
| 0   | 6    | -6           | $P_6(6) = \frac{6!}{(6!)^2} = \frac{1}{64}$    | $P_6(6) = \frac{1}{64}$                              | a) What is $\bar{N} = \sum P(N)N$ ?<br>b) What is $\bar{n}$ ?<br>c) What are $(n - \bar{n})$ and $(N - \bar{N})$ ? |

$$\ln P(N) = m! - \ln \left( \frac{m-N}{2} \right)! - \left( \frac{m+N}{2} \right)! - m \ln 2$$

Use Stirling's formula

Example of Central Limit Theorem

$$\begin{aligned} \ln P(N) &= \left( m + \frac{1}{2} \right) \ln m - m + \frac{1}{2} \ln(2\pi) \\ &\quad - \left( \frac{m-N}{2} + \frac{1}{2} \right) \ln \left( \frac{m-N}{2} \right) + \frac{m-N}{2} - \frac{1}{2} \ln(2\pi) \\ &\quad - \left( \frac{m+N}{2} + \frac{1}{2} \right) \ln \left( \frac{m+N}{2} \right) + \frac{m+N}{2} - \frac{1}{2} \ln(2\pi) \\ &\quad - m \ln 2 \end{aligned}$$

$$\ln \left( \frac{m-N}{2} \right) = \ln \frac{m}{2} \left( 1 - \frac{N}{m} \right) \approx \ln \frac{m}{2} + \left( -\frac{N}{m} - \frac{1}{2} \left( \frac{N}{m} \right)^2 \right)$$

$$\ln \left( \frac{m+N}{2} \right) = \ln \frac{m}{2} \left( 1 + \frac{N}{m} \right) \approx \ln \frac{m}{2} + \left( \frac{N}{m} - \frac{1}{2} \left( \frac{N}{m} \right)^2 \right)$$

Thus

$$\begin{aligned} \ln P(N) &= + \frac{1}{2} \ln m + m \ln 2 - \ln m + \ln 2 - m \ln 2 \\ &\quad - \frac{N^2}{m} + \frac{1}{2} \frac{N^2}{m} + \frac{1}{2} \left( \frac{N}{m} \right)^2 - \frac{1}{2} \ln(2\pi) \end{aligned}$$

higher order

Taking Antilog

$$P(N) = \frac{2}{\sqrt{2\pi m}} e^{-\frac{1}{2} \frac{N^2}{m}} \quad N > 0$$

(If  $N$  can be negative as well) (steps to left; divide by 2) No. of steps right

To Relate this to diffusion let  $m = qt$   
 where  $q$  = no of steps/unit time ;  $x = Nl$   
 where  $l$  = distance between steps. Let  $D = \frac{ql^2}{2}$   
 the Diffusion coefficient

$$P(N) \rightarrow \frac{dP}{dx} dx = \frac{1}{\sqrt{\pi D t}} e^{-\frac{x^2}{4Dt}} l$$

This satisfies the diffusion equation  $D \frac{d^2 P}{dx^2} = + \frac{dP}{dt}$