Equation 7.6 gives the probability per unit energy for finding an electron in a given energy state E at a given temperature T:

$$P(T,E) = \frac{1}{1 + \exp\left\{\left(E - E_{\mathrm{F}}\right)/kT\right\}},$$

where $E_{
m F}$ is the Fermi energy and k is the Stefan-Boltzmann constant and equals $8.62 imes 10^{-5}\,{
m eV}~{
m K}^{-1}.$

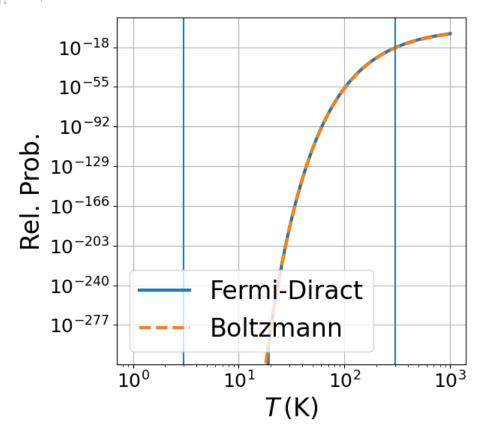
The bandgap energy for silicon is the difference in energy between the top of the valence band and the bottom of the conduction band $E_{\rm G}=E-E_{\rm F}=1.12\,{\rm eV}$ (Table 7.4).

By taking the ratio of P at the two temperatures, we can estimate the relative probabilities:

$$P(E=E_{
m G})/P(E=E_{
m F}) = rac{1}{1+\exp\left\{(E-E_{
m F})/kT
ight\}} \, .$$

Let's make a plot to compare the prediction from Fermi-Dirac statistics (Equation 7.6) to that of Boltzmann (Equation 7.3).

```
In [111]: %matplotlib inline
         import matplotlib.pyplot as plt
         import numpy as np
         StefanBoltzmann_const = 8.62e-5 # eV/K
         def FermiDirac probability(T. E=1.12):
             return 2./(1. + np.exp(E/StefanBoltzmann const/T))
         def Boltzmann probability(T, E=1.12, gi=1, gi=1);
             return gi/gj*np.exp(-E/StefanBoltzmann_const/T)
         temperatures = np.linspace(1., 1000., 1000)
         FD prob = FermiDirac probability(temperatures)
         B_prob = Boltzmann_probability(temperatures)
         fig = plt.figure(figsize=(6, 6))
         ax = fig.add subplot(111)
         ax.loglog(temperatures, FD_prob, lw=3, label="Fermi-Diract")
         ax.loglog(temperatures, B_prob, lw=3, ls='--', label="Boltzmann")
         ax.grid(True)
         ax.legend(loc='best', fontsize=24)
         ax.tick params(labelsize=18)
         ax.set_xlabel(r'$T\,\left( {\rm K} \right)$', fontsize=24)
         ax.set_ylabel("Rel. Prob.", fontsize=24)
         ax.axvline(3.)
         ax.axvline(300.)
```



You can see that, even for $T o 1000\,\mathrm{K}$, there is a very small probability to occupy the conduction band.

7.5

We're told to assume the electrical conductivity depends on the number density of electrons, which we can estimate as a function of bandgap energy $E_{\rm G}$ and temperature T using Equation 7.9:

$$n_N = A T^{3/2} e^{-rac{E_{
m G}}{kT}}.$$

For silicon, $E_{
m G}=1.12\,{
m eV}$ and for germanium, $E_{
m G}=0.67\,{
m eV}$ (Table 7.4).

So we're asked to compare n_N for $T=40\,\mathrm{K}$ and for $T=40\,\mathrm{K}+1\,\mathrm{K}$ for these two semiconductors. This is a very small difference $\Delta T/T=1/40$, so let's Taylor-expand n_N about small ΔT :

$$\begin{split} (T+\Delta T)^{3/2} &= T^{3/2} \bigg(1+\frac{\Delta T}{T}\bigg)^{3/2} \approx T^{3/2} \left(1+\frac{3\Delta T}{2T}\right) \\ &\exp\bigg(-\frac{E}{k\left(T+\Delta T\right)}\bigg) = \exp\bigg(-\frac{E}{kT}\bigg(1+\frac{\Delta T}{T}\bigg)^{-1}\bigg) \approx \exp\bigg(-\frac{E}{kT}\bigg(1-\frac{\Delta T}{T}\bigg)\bigg) = \exp\bigg(-\frac{E}{kT}\bigg) \, \exp\bigg(\frac{E}{kT}\frac{\Delta T}{T}\bigg) \\ &\approx \exp\bigg(-\frac{E}{kT}\bigg) \, \bigg(1+\frac{E}{kT}\frac{\Delta T}{T}\bigg) \\ &\Rightarrow T^{3/2} e^{-\frac{E}{kT}} \, \bigg(1+\bigg(\frac{3}{2}+\frac{E}{kT}\bigg)\,\frac{\Delta T}{T}\bigg) \, . \end{split}$$

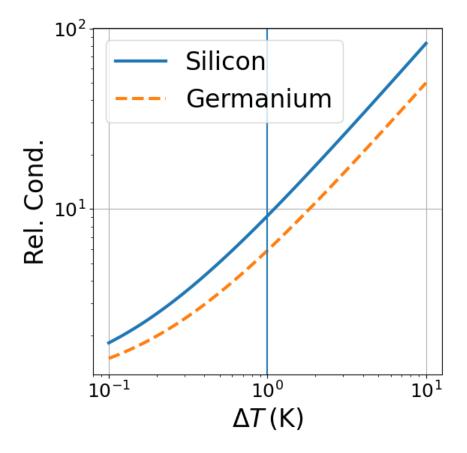
The term on the left outside the parentheses is just number density of electrons at temperature T, so taking the ratio gives

$$rac{n_N(T+\Delta T)}{n_N(T)}pprox \left(1+\left(rac{3}{2}+rac{E}{kT}
ight)rac{\Delta T}{T}
ight).$$

Again, we can plot this expression.

```
In [20]: %matplotlib inline
         import matplotlib.pyplot as plt
         import numpy as np
         StefanBoltzmann_const = 8.62e-5 # eV/K
         def relative_conductivity(Delta_T, E, T=40.):
             return (1. + (3./2 + E/StefanBoltzmann_const/T)*Delta_T/T)
         Delta_T = 10**np.linspace(-1, 1, 100)
         E_Si = 1.12
         E_{Ge} = 0.67
         Si_prob = relative_conductivity(Delta_T, E_Si)
         Ge_prob = relative_conductivity(Delta_T, E_Ge)
         fig = plt.figure(figsize=(6, 6))
         ax = fig.add_subplot(111)
         ax.loglog(Delta_T, Si_prob, lw=3, label="Silicon")
         ax.loglog(Delta_T, Ge_prob, lw=3, ls='--', label="Germanium")
         ax.grid(True)
         ax.legend(loc='best', fontsize=24)
         ax.tick params(labelsize=18)
         ax.set_xlabel(r'$\Delta T\,\left( {\rm K} \right)$', fontsize=24)
         ax.set_ylabel("Rel. Cond.", fontsize=24)
         ax.axvline(1.)
```

Out[20]: <matplotlib.lines.Line2D at 0x7fefdac7dbe0>



7.6

The quantum efficiency q is a measure of how readily photons are converted into current by the CCD, and it depends on the surface reflectivity R, the absorption coefficient α , and the layer thickness z.

Equation 7.11 tells us how R depends on the index of refraction:

$$R=igg(rac{n_1-n_2}{n_1+n_2}igg)^2,$$

where n_1 is the index for air (≈ 1) and n_2 is the index for silicon (we're told $n_2 = 4$ at $500 \, \mathrm{nm}$).

With these numbers, we get

$$R = \left(\frac{1-4}{1+4}\right)^2 = 0.36,$$

which means 36\% of the incident energy is reflected, leaving 74\% to be absorbed the CCD.

Next, we need to determine the absorption coefficient for silicon at 500 nm. Figure 7.10 gives that information. The figure suggests $\alpha = 10^4 \, \mathrm{cm}^{-1}$. (Unfortunately, Chromey has a typo in the figure caption. You can see a similar plot with the right units here - https://www.pveducation.org/pvcdrom/materials/optical-properties-of-silicon.)

So we can estimate the required thickness z to achieve a given quantum efficiency q using

$$q = (1-R) \, e^{-lpha z} \Rightarrow z = rac{\ln\left(rac{1-R}{q}
ight)}{lpha} = rac{\ln\left(rac{0.76}{0.4}
ight)}{\left(10^4\,\mathrm{cm}^{-1}
ight)} pprox [600\,\mathrm{nm}].$$

8.1

We're told we start out with a quantum efficiency of 40\% for our detector.

Referring to Problem 7.6 above, we can cast the quantum efficiency as

$$q = (1 - R) e^{-\alpha z}.$$

We are told that the reflectivity of the detector is reduced from 30\% to 5\%. Since nothing else about the detector has changed, we can calculate the ratio of the new to the old efficieny as

$$\frac{q'}{q} = \frac{1 - R'}{1 - R},$$

where the primed quantities represent the new values.

Solving for q' gives

$$q' = q\left(\frac{1-R'}{1-R}\right) = (0.4)\left(\frac{1-0.05}{1-0.3}\right) = \boxed{0.54}.$$

8.2

Equation 8.5 tells us

$$ext{DQE} = rac{\left(ext{SNR}_{ ext{out}}
ight)^2}{\left(ext{SNR}_{ ext{perfect}}
ight)^2} = rac{\left(ext{SNR}_{ ext{out}}
ight)^2}{N_{ ext{in}}}.$$

We are told that the measurement involves 10^4 photons, which will take as $N_{\rm in}$. So next we'll need the signal-to-noise ratio for the voltages, ${\rm SNR_{out}}$.

Our average voltage (the signal) is

$$S = \frac{(113 + 120 + 115) \text{ mV}}{3} = 116 \text{ mV}.$$

And the noise will be the standard deviation of the voltage measurements:

$$\sigma = 3 \, \mathrm{mV}$$
.

So now we can write our DQE as

$$\mathrm{DQE} = rac{\left(rac{116 \mathrm{\ mV}}{3 \mathrm{\ mV}}
ight)^2}{10^4} pprox 0.15.$$

8.3

The detector in this problem has a quantum efficiency of q. That means that, for every $N_{\rm in}$ photon that strikes the surface of the detector, $N_{\rm detect}=qN_{\rm in}$ are actually detected (and $(1-q)N_{\rm in}$ are NOT detected).

The detector is also said to have a quantum yield of y with an associated uncertainty $\sigma(y)$. That means that each of the $N_{\rm detect}=qN_{\rm in}$ induces $N_{\rm events}=yN_{\rm detect}$ events that are counted up by the detector.

So, as described by Equation 2.31 back in Chapter 2, the uncertainty $\sigma(N_{\rm events})$ associated with the events is given by

$$\sigma^2(N_{
m events}) = \left(rac{\partial N_{
m events}}{\partial N_{
m detect}}
ight)^2 \sigma^2\left(N_{
m detect}
ight) + \left(rac{\partial N_{
m events}}{\partial y}
ight)^2 \sigma^2\left(y
ight) = y^2 N_{
m detect} + N_{
m detect}^2 \sigma^2(y),$$

taking Poisson error bars for $N_{
m detect}$ (i.e., $\sigma^2\left(N_{
m detect}\right)=N_{
m detect}$).

To calculate DQE, we need $(SNR)^2_{\rm out}$ and $(SNR)^2_{\rm perfect}$. The latter is easy:

$$\left(SNR\right)_{\mathrm{perfect}}^2 = N_{\mathrm{in}}.$$

The former is

$$(SNR)_{ ext{out}}^2 = rac{y^2 N_{ ext{detect}}^2}{y^2 N_{ ext{detect}} + N_{ ext{detect}}^2 \sigma^2(y)} = rac{N_{ ext{detect}}^2}{N_{ ext{detect}} + N_{ ext{detect}}^2 \left(rac{\sigma(y)}{y}
ight)^2}$$

Now to calculate DQE:

$$ext{DQE} = rac{N_{ ext{detect}}^2}{N_{ ext{detect}} + N_{ ext{detect}}^2 \left(rac{\sigma(y)}{y}
ight)^2}}{N_{ ext{in}}}$$

But remember $N_{
m detect}=qN_{
m in}$, so

$$ext{DQE} = rac{rac{q^2 N_{
m in}^2}{q N_{
m in} + q^2 N_{
m in}^2 \left(rac{\sigma(y)}{y}
ight)^2}}{N_{
m in}} = rac{q^2 N_{
m in}}{q N_{
m in} + q^2 N_{
m in}^2 \left(rac{\sigma(y)}{y}
ight)^2} = \left[rac{q}{1 + q N_{
m in} \left(rac{\sigma(y)}{y}
ight)^2}
ight]$$

So you can see that, if $\sigma(y)=0$, then $\mathrm{DQE}=q$. However, since all numbers are positive, if $\sigma(y)>0$, then $\mathrm{DQE}< q$.

8.4

We'll need to recall the definition of QE (Equation 8.1):

$$ext{QE} = rac{N_{ ext{detect}}}{N_{ ext{in}}}.$$

We're told that QE=0.9, so if 1000 photons are incident in the 1-second exposure, we'll actually detect $(0.9) \times 1000 = 900$ of them.

Then recall the definition of DQE (Equation 8.5):

$$ext{DQE} = rac{\left(ext{SNR}
ight)_{ ext{out}}^2}{\left(ext{SNR}
ight)_{ ext{perfect}}^2}.$$

For our detector we have a signal S=900 photons and sources of noise including the dark current $(1 \text{ photon s}^{-1}) \times (1 \text{ s}) = 1 \text{ photon}$ and a read noise of 3 photons.

Therefore, our out SNR is

$${\rm (SNR)}_{\rm out} = rac{900}{\sqrt{(900+1+3)+1+3}} pprox 30.$$

A perfect detector would have

$$(SNR)_{perfect} = \sqrt{1000} \approx 32.$$

So

$$DQE = \frac{30^2}{32^2} = \boxed{0.89}.$$

For the 4-second exposure, we have all the same numbers except now the dark current is $(1 \text{ photon s}^{-1}) \times (400 \text{ s}) = 400 \text{ photon}$.

So

$${
m (SNR)}_{
m out} = rac{900}{\sqrt{(900+400+3)+400+3}} pprox 22.$$

And

$$DQE = \frac{22^2}{32^2} = \boxed{0.47}.$$