

## P653 HW2

Due Sept 8, 2005

### Problem 1. Information Theoretic Definition of Entropy

Entropy is a measure of the uncertainty in an ensemble. Here we will use some simple information theoretic ideas (due to Shannon) to calculate the entropy for a generic ensemble – even one which is not in equilibrium.

Our setup is that we know that our system is in one of  $N$  states, and that each state  $i$  occurs in the ensemble with probability  $p_i$ .

Suppose we measure that the system is in state  $i$ . Let us assume that we can quantify the amount that we learn from this measurement. Call this  $k_i$ . Clearly  $k_i$  is larger if the probability  $p_i$  is smaller. We will define  $k_i$  so that it is only a function of the probability  $p_i$ , ie

$$k_i = k(p_i).$$

We will define the entropy of information  $S_I$  (for the remainder of this question we drop the  $I$ ) to be the average amount of information gained from a measurement:

$$S = \sum_i p_i k(p_i).$$

**1.1.** Entropy, as a state variable is extensive. Suppose we build a larger system out of two independent subsystems  $A$  and  $B$ . The entropy of the whole should be

$$S = S_A + S_B,$$

where  $S_A$  and  $S_B$  are the entropies of the parts.

Use this relationship to derive a functional relationship that  $k$  must obey.

[Hint 1: Recall that independence means that the probability of  $A$  being in state  $i$  and  $B$  being in state  $j$  is  $p_{ij} = p_i + p_j$ .]

[Hint 2: You will need to use that  $\sum_i p_i = 1$ .]

**1.2.** Show that this functional relationship is obeyed by

$$k(x) = -k \log(x),$$

for any constant  $k$ .

Prove that these are the only solutions to this relationship. [Feel free to assume that  $k(x)$  is differentiable, and you may use standard uniqueness theorems.]

**1.3.** We need that the entropy will be maximized in equilibrium. What constraint does that place on the sign of  $k$ ?

As the notation might suggest, we will take the constant  $k$  to be Boltzmann's constant. We have therefore arrived at the definition

$$S_I = -k_b \sum_i p_i \log p_i,$$

where the  $I$  stands for information.

**1.4.** Show that the thermodynamic entropy in the microcanonical ensemble is equal to the information entropy.

**1.5.** Show that the thermodynamic entropy in the canonical ensemble is equal to the information entropy.

**Problem 2. Maximum Entropy Principle** Consider the set of all possible states  $i$  of a system. Each of these state contains a given number of particles  $N_i$  and has energy  $E_i$ . We define an ensemble of these states by specifying the probability  $p_i$  for the system to be in a given state.

Find the  $p_i$ 's for the ensemble which maximizes the entropy  $S_I$  subject to the constraint that the expectation value of the energy and particle number are fixed,

$$\langle E \rangle = \sum_i p_i E_i$$

$$\langle N \rangle = \sum_i p_i N_i$$

[Don't forget that the probabilities are constrained by  $\sum_i p_i = 1$ .]

### Problem 3. Microscopic origin of dissipation

Here we look at a simple model of a dissipative classical system, namely an oscillator coupled to an oscillator bath. For a quantum treatment, see Feynman and Vernon, Ann. Phys. **24**, 118 (1963).

Consider a harmonic oscillator with position  $X$ , momentum  $P$ , and frequency  $\Omega$ , linearly coupled to a bath of other oscillators with positions  $x_i$ , momenta  $p_i$  and frequencies  $\omega_i$ . This system obeys a Hamiltonian

$$H = \frac{P^2}{2M} + \frac{1}{2}M\Omega^2 X^2 + \sum_i \left( \frac{p_i^2}{2m_i} + \frac{1}{2}m_i\omega_i^2 x_i^2 \right) + \sum_i \lambda_i x_i X.$$

**3.1.** Write the equations of motion for  $x_i$ ,  $p_i$ ,  $X$ , and  $P$ .

**3.2.** Fourier transform these equations of motion, using the convention

$$A(t) = \int \frac{d\omega}{2\pi} e^{-i\omega t} A(\omega).$$

**3.3.** Eliminate  $p_i(\omega)$  to write  $x_i(\omega)$  as a function of  $X(\omega)$ .

**3.4.** Eliminate  $x_i(\omega)$  and  $P$  from the equations for  $X$ , to arrive at an equation of the form,

$$[\omega^2 - \Omega^2 - F(\omega)]X = 0.$$

What is  $F(\omega)$ ?

**3.5.** It is convenient to introduce a "spectral density"

$$J(E) = \sum_i \frac{\lambda_i^2}{m_i \omega_i} 2\pi \delta(E - \omega_i)$$

which encodes all relevant information about the oscillator bath. Verify that

$$F(\omega) = \frac{1}{2M} \int \frac{dE}{2\pi} \frac{1}{\omega - E} [J(E) - J(-E)].$$

**3.6.** In the limit of a large bath, one can assume that  $J(E)$  is smooth. We will concentrate on an *ohmic* bath, where  $J(E) = \alpha E$  for small  $E$ . To be concrete, we will take

$$J(E) = \begin{cases} \alpha E & |E| < E_c \\ 0 & |E| > E_c \end{cases}$$

Furthermore, we will assume that  $E_c$  is the largest energy in the problem. Explicitly calculate  $F(\omega)$ , neglecting terms of order  $\omega/E_c$ . It may be useful to recall that

$$\frac{1}{x - y} = \frac{P}{x - y} - i\pi\delta(x - y),$$

though this example has been cooked up so that one does not have to explicitly use this formula.

**3.7.** Substituting this expression for  $F$  into equation of motion for  $X$  should give the equations of motion for a damped harmonic oscillator. What is the frequency  $\bar{\Omega}$  of this oscillator? Note that in most experimental realizations of this system there is no way to measure the bare frequency  $\Omega$ . Only the renormalized frequency  $\bar{\Omega}$  is physical.