

KMS states and Tomita-Takesaki Theory

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Chapter 1

Introduction

Chapter 2

Classical and Quantum Mechanics as Probability Theories

2.1 Classical Mechanics

The setting of classical mechanics is usually a locally compact Hausdorff space X . We consider the states (of maximal knowledge) to be the elements of X . Likewise, observables take the form of functions on X . Moreover, given that in principle we could make the value of an observable as precise as we want by improving our knowledge of the state, observables are elements of the set of continuous functions $C(X)$ ¹. Nonetheless the purpose of statistical mechanics is to treat systems in which total knowledge of a state is not practically possible. Instead we consider a probability measure² which assigns to every measurable subset of X a probability of the system's state being in it. We may define the expected value of an observable $f \in C(X)$ through a probability measure P by

$$\langle f \rangle_P = \int f dP. \quad (2.1)$$

Notice that an element $p \in X$ can also be thought as a probability measure by using the Dirac measure δ_p which assigns 1 to a set if it contains p and

¹It doesn't matter whether we consider them real or complex at this stage.

²Along with a σ -algebra which we won't mention explicitly to keep the notation simple but should always be kept in mind.

0 otherwise. Indeed for every element $p \in X$ and observable $f \in C(X)$ we have $\langle f \rangle_{\delta_p} = f(p)$. This motivates us to broaden the definition of states to the probability measures on X . We will call Dirac measures (or equivalently the points in X) pure states.

This definition of state proves to be very helpful for the discussion of ensembles. Whenever the description of the state of a system as a pure state is not feasible, we may consider the set of outcomes Y of measurements we may perform on the system. Every element of Y gives us information of the system in the form of a finite measure. We may define an ensemble as the mapping from Y into the set of finite measures on X . Through normalization of finite measures every ensemble yields a mapping from Y into the set of states and we define the accesible (pure) states of an element $y \in Y$ to be the support of the corresponding state. Although the construction of an ensemble is in general a difficult task, for systems in statistical equilibrium³ there are many standard procedures. In the case of these type of systems we define the partition function $Z : Y \rightarrow \mathbb{R}_0^+$ by assigning to every element y the measure of X given by the ensemble evaluated at y .

Example 2.1.1 *In many physical systems the space of pure states has a natural notion of size which we may represent by giving it the structure of a measure space (X, \mathcal{A}, μ) where \mathcal{A} contains the Borel σ -algebra⁴. We may consider $Y = \mathbb{R}$ to be the set of energy outcomes. If $H : X \rightarrow \mathbb{R}$ is a measurable function taking the interpretation of energy we define the micro-canonical ensemble to be the mapping $y \mapsto \mu_y$ where $\mu_y(\Sigma) = \mu(\Sigma \cap H^{-1}(y))$ for all measurable Σ . The set $H^{-1}(y)$ is the set of accesible states and $\mu_y(\Sigma)$ measures the amount of pure states in Σ which are accesible. Notice that the normalization of μ_y yields a state P_y which assigns a uniform probability measure to X . This is called the equal a priori probabilities postulate. In this ensemble the partition function $Z(y) = \mu_y(X) = \mu(H^{-1}(y))$ is just the amount of accesible states. This ensemble is usually used to describe systems with constant energy and a fixed number of particles.*

Example 2.1.2 *Consider again a measure space (X, \mathcal{A}, μ) but let $Y = \mathbb{R}^+$ be the set of inverse temperatures of the system. If we have an energy function*

³These are systems whose state does not change in time. We refer to the equilibrium as statistical because it may be that the pure state of the system is changing in time but noticing these changes is not feasible for us.

⁴Usually we take a countable set with the counting measure but another example would be a phase space with the Liouville measure.

$H : X \rightarrow \mathbb{R}$ such that $x \rightarrow \exp(-yH(x))$ is integrable for all $y \in Y$ the canonical ensemble assigns to every inverse temperature y a finite measure μ_y by

$$\mu_y(\Sigma) = \int_{\Sigma} e^{-yH(x)} d\mu(x) \quad (2.2)$$

for all measurable sets Σ . This ensemble is usually used to describe systems with a fixed number of particles in thermal equilibrium with a heat bath. Note that we could add to the description of the system the heat bath and we would be able to in principle use the microcanonical ensemble. The difficulty lies in that generally the counting of accesible states is more difficult than the application of the canonical ensemble.

Note that both of the ensembles discussed have images consisting of absolutely continuous measures μ_y with respect to the notion of size μ . The same is true for the induced states P_y . Moreover the Lebesgue-Radon-Nikodým derivative exists and we define the entropy of the ensemble in the state P_y by⁵

$$S(P_y) = - \int_{\text{supp}(P_y)} \log \left(\frac{dP_y}{d\mu} \right) \frac{dP_y}{d\mu} d\mu. \quad (2.3)$$

One can check that in the microcanonical ensemble

$$\frac{dP_y}{d\mu}(x) = \frac{\chi_{H^{-1}(y)}(x)}{Z(y)} \quad (2.4)$$

and in the canonical ensemble

$$\frac{dP_y}{d\mu}(x) = \frac{\exp(-yH(x))}{Z(y)}. \quad (2.5)$$

⁵In general if we start from a decomposable (X, \mathcal{A}, μ) and have an ensemble which yields absolutely continuous measures with respect to μ we can define entropy in this fashion. In particular, if μ comes from the daniel extension of a positive linear functional the space is decomposable.