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## Problem Set #3

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Due December 4th

Please type up your answers (Latex, Mathematica, Word, etc.) and email them as a pdf file to [scharfenakere@umkc.edu](mailto:scharfenakere@umkc.edu).

### ANALYTICAL

1. Find the maximum entropy distributions over the following sets with the described constraints.
  - a) The whole real line constrained to have a given value of its variance.
  - b) The nonnegative real half-line constrained to have a given geometric and arithmetic mean.
  - c) The whole real line constrained to have a given mean and absolute mean.
2. The entropy difference between two distributions is measured by the Kullback-Leibler divergence. Given data  $x$  in the form of a vector of binned frequencies of  $n$  observations,  $\tilde{f}[x_k] = \{\frac{x_1}{n}, \dots, \frac{x_K}{n}\}$ , we can always view this data as a sample of a multinomial model with predicted frequencies  $\hat{f}[x_k] = \{p_1, \dots, p_K\}$ . The likelihood function for a sample  $x = \{x_1, \dots, x_K\}$  (where  $x_j$  is the number of observations in bin  $j$ ) from a multinomial distribution, writing  $n = \sum_{k=1}^K x_k$  for the sample size, is:

$$P[x|p] = \frac{n!}{x_1! \dots x_K!} p_1^{x_1} \dots p_K^{x_K} \quad (0.1)$$

Using Stirling's approximation,  $\log[x!] \approx x \log[x] - x$ , prove that

$$\begin{aligned}
P[x|p] &= e^{-n \sum_{k=1}^K \tilde{f}[x_k] \log \left[ \frac{\tilde{f}[x_k]}{\hat{f}[x_k]} \right]} \\
&= e^{-n \text{KL}[\tilde{f}[x_k], \hat{f}[x_k]]}
\end{aligned} \tag{0.2}$$

That is, show that the log of the multinomial likelihood is equivalent to the Kullback-Leibler divergence between the empirical and theoretical bin frequencies.

3. Shannon's theorem is a useful and intuitive point of departure for modeling the uncertainty in an economic agents' decision environment. It quantifies the amount of choice variability for a given decision environment and is an increasing function of the number of choices available. Because Shannon entropy is a measure of uncertainty it can provide a meaningful probabilistic description of a decision environment by being incorporated into the behavior of agents as a behavioral constraint. Assuming the typical agent conditions their discrete choices  $a$  on perceived signals  $x$  from their social environment, we can summarize individual behavior as responding to a payoff function  $v[a, x]$ . The typical agent's mixed strategy in this scenario is defined by the conditional probability distribution  $p[a|x] : A \times X \rightarrow (0, 1)$  over the actions  $a \in A$  that maximizes the expected payoff,  $\sum_a p[a|x] v[a, x]$ . The agent's expected payoff maximization program subject to a constraint on the minimum entropy of the mixed strategy distribution is:

$$\begin{aligned}
&\text{Max}_{\{p[a|x] \geq 0\}} \sum_a p[a|x] v[a, x] \\
&\text{subject to } \sum_a p[a|x] = 1 \\
&\quad - \sum_a p[a|x] \text{Log}[p[a|x]] \geq H_{min}
\end{aligned} \tag{0.3}$$

- a) Find the maximum entropy distribution over the action set conditional on  $x$ , that solve the programming problem for  $p[a|x]$ .
- b) If the payoff function is equal to the outcome  $v[a, x] = x$  and the action set consists of only two actions  $a = \{a_1, a_2\}$ , what is the MAXENT distribution for  $p[a|x]$ ?
- c) For this MAXENT distribution determine what happens in limit when the shadow price of reducing the entropy constraint goes to zero and infinity.