

Use only your course notes, integral tables or Mathematica.

### 1. Linear Gaussian Systems. (12 points)

Let  $x$  be a Gaussian distributed scalar random variable with mean zero and variance  $\sigma_x^2$ . Let  $y$  be related to  $x$  by

$$y = Ax + \mu + \epsilon \quad (1)$$

where  $A$  and  $\mu$  are constants, and  $\epsilon$  is Gaussian noise with mean zero and variance  $\sigma_\epsilon^2$ . Assume  $x$  and  $\epsilon$  are statistically independent. For example,  $x$  might be a hidden continuous state variable, and  $y$  an observable linearly related to  $x$  with additive noise  $\epsilon$ . More concretely, suppose  $x$  is the Celsius temperature of some object (drawn from a population of objects with zero-mean Gaussian temperature distribution) and  $y$  is the reading of a noisy, Fahrenheit thermometer (with  $A$  and  $\mu$  the scale change and offset between the Celsius and Fahrenheit systems). We are going to estimate  $x$  and its distribution by using Bayes theorem to calculate  $p(x|y)$ .

(a) From the definition of  $y$  in terms of  $x$  and  $\epsilon$ , show that

$$E[y|x] = E[Ax + \mu + \epsilon|x] = Ax + \mu \quad (2)$$

You need *not* write  $p(y|x)$  explicitly to evaluate this, but you can. If you don't write the density, use words to make a clear argument for the result.

(b) Also use the definition of  $y$  to show that

$$\text{var}(y|x) = E[(y - E[y|x])^2|x] = E[(Ax + \mu + \epsilon - E[y|x])^2|x] = \sigma_\epsilon^2 \quad (3)$$

Again, need *not* write  $p(y|x)$  explicitly to show this, but you can.

(c) Show that

$$\mu_y = E[y] = \mu \quad (4)$$

$$\text{var}(y) = E[(y - \mu_y)^2] = A^2 \sigma_x^2 + \sigma_\epsilon^2 \quad (5)$$

using the fact that  $\epsilon$  and  $x$  are statistically independent (and therefore uncorrelated).

(d) Finally, use the fact that linear functions of Gaussian variables are Gaussian, and equations (2)-(5) to write  $p(y|x)$ ,  $p(y)$  and then, via Bayes' theorem the posterior distribution  $p(x|y)$

$$p(x|y) = \frac{p(y|x)p(x)}{p(y)} \quad .$$

Your answer should **explicitly** express the conditional mean  $E[x|y]$  and the conditional variance  $\text{var}(x|y)$  in terms of  $A$ ,  $\mu$ ,  $\sigma_x^2$ , and  $\sigma_\epsilon^2$ .

Hint: You should complete the square to write  $p(x|y)$  and extract the conditional mean and variance.

## 2. The 80/20 Rule: Heavy-Tailed Densities (12 points)

Read ‘Million Dollar Murray’ in the class Blackboard site Documents → Other Notes and Readings.

The Pareto principle (also called the 80/20 rule) states that for many phenomena, 80% of the effects come from 20% of the causes. The notion that a few sources together carry a huge proportion of the effects has attracted a lot of attention in recent years, although the ratios are not often 80/20. For example, the most wealthy 1% of US households collectively hold 40% of the country’s wealth (Christopher Ingraham, Washington Post web page, Dec. 6, 2017).

In both social and physical phenomena, heavy-tailed distributions — densities with tails that drop off as a power law, rather than exponentially — occur widely and give rise to such skewed proportions. Such distributions contrast strongly those with tails that decay exponentially, which are appropriate for phenomena with cases that cluster around the peak of the density (e.g. Gaussian and exponential densities).

The Pareto Type I density is

$$p(x) = \alpha x_m \left( \frac{1}{x} \right)^{\alpha+1} \quad (6)$$

with support on  $x_m \leq x < \infty$ , and parameter restricted to  $\alpha > 1$ . Let’s use the Pareto Type I to explore wealth allocation for heavy-tailed distributions.

Let  $x$  denote the net worth of individuals in a population modeled by a Pareto type I density. For convenience, set the minimum net worth to  $x_m = 1$ .

- (a) (2 points) Sketch or plot a Pareto type I density with  $x_m = 1$  and  $\alpha > 1$ . Show that the proportion of the population with net worth greater than or equal to  $x_0 > x_m = 1$  is

$$f_{\geq} = \int_{x_0}^{\infty} p(x) dx = \left( \frac{1}{x_0} \right)^{\alpha} \quad (7)$$

- (b) (2 points) Show that the *average* wealth in the entire population is

$$\text{wealth} = E[x] = \int_1^{\infty} x p(x) dx = \frac{\alpha}{\alpha - 1} . \quad (8)$$

- (c) (2 points) Show that the average wealth among individuals with net worth greater than or equal to  $x_0$  is

$$\text{wealth}(x_0) = \int_{x_0}^{\infty} x p(x) dx = \frac{\alpha}{\alpha - 1} \left( \frac{1}{x_0} \right)^{\alpha} x_0 \quad (9)$$

- (d) (2 points) Conclude that the fraction of the total wealth held by those with net worth greater than  $x_0$  is

$$f_{\geq} = \frac{\text{wealth}(x_0)}{\text{wealth}} = \left( \frac{1}{x_0} \right)^{\alpha-1} . \quad (10)$$

- (e) (4 points) To write the proportion of wealth held by the rich in terms of their proportion of the population, solve Equation (7) for  $x_0(f_0)$  and substitute into Equation (10). Plot the resulting  $f_{\geq}$  vs  $f_0$  for values of  $\alpha$  in the range  $1.05 \leq \alpha \leq 100$ . Make plots for  $0 \leq f_0 \leq 0.2$  as well as for  $0 \leq f_0 \leq 1.0$ . Interpret the results in words. What value of  $\alpha$  corresponds to Ingraham’s report that  $f_0 = 0.01$  yields  $f_{\geq} = 40\%$ ?