## Homework 2: problems about exponential families

## Rajdeep Brahma

Due: February 9th at 11:59 PM

This homework set will cover problems concerning exponential family theory. All derivations must be typed. Screenshots of work done with pen and paper will not be accepted.

**Problem 1**: Verify that displayed equation 7 in the exponential family notes holds for the binomial distribution, the Poisson distribution, and the normal distribution with both  $\mu$  and  $\sigma^2$  unknown.

Solution: First, we consider the binomial distribution. We have  $c(\theta)=\operatorname{n}\log(1+e^{\theta})$  where  $\theta=\log(\frac{p}{1-p})$ . Now  $\nabla c(\theta)=n\frac{e^{\theta}}{1+e^{\theta}}=np$  and  $\nabla^2 c(\theta)=n\frac{e^{\theta}}{(1+e^{\theta})^2}=np(1-p)$ . So, this matches the expectation and variance of the binomial distribution. Next, we consider the poisson distribution. We have  $c(\theta)=e^{\theta}$  where  $\theta=\log(\lambda)$ . Now  $\nabla c(\theta)=e^{\theta}=\lambda$  and  $\nabla^2 c(\theta)=e^{\theta}=\lambda$ . So, this matches the expectation and variance of the poisson distribution. Finally, we consider the normal distribution. We have  $c(\theta)=-\frac{\theta_1^2}{4\theta_2}+\frac{1}{2}\log(-\frac{1}{2\theta_2})$  where  $\theta=(\theta_1,\theta_2)=(\frac{\mu}{\sigma^2},-\frac{1}{2\sigma^2})$ . Now  $\nabla c(\theta)=(-\frac{\theta_1}{2\theta_2},\frac{\theta_1^2}{4\theta_2^2}-\frac{1}{2\theta_2})=(\mu,\mu^2+\sigma^2)$  and  $\nabla^2 c(\theta)=\begin{pmatrix} -\frac{1}{2\theta_2}&\frac{\theta_1}{2\theta_2^2}\\ \frac{\theta_1}{2\theta_2^2}&\frac{1}{2\theta_2^2}-\frac{\theta_1^2}{2\theta_2^3} \end{pmatrix}=\begin{pmatrix} \frac{\sigma^2}{2\mu\sigma^2}&\frac{2\mu\sigma^2}{2\sigma^4+4\mu^2\sigma^2} \end{pmatrix}$ . So, this matches the expectation and variance of  $(X,X^2)$  where  $X\sim N(\mu,\sigma^2)$ .

**Problem 2**: This problem concerns the proof of Theorem 3 in the exponential family notes. Do the following:

- part a: Show that the second derivative of the map h is equal to  $-\nabla^2 c(\theta)$  and justify that this matrix is negative definite when the exponential family model is identifiable.
- part a: Finish the proof of Theorem 3.

Note that part a will be referenced later in this course. Hence, it is treated as its own sub-problem. Solution:

$$h(\theta) = \langle \mu, \theta \rangle - c(\theta).$$

. Hence  $\nabla h(\theta) = \mu - \nabla c(\theta)$  and  $\nabla^2 h(\theta) = -\nabla^2 c(\theta) = -\mathrm{Var}_{\theta}(Y)$ . This is a negative definite matrix because the exponential family model is identifiable implying that Y is not concentrated on a hyperplane. To finish the proof of Theorem 3 we need to prove 2 more results. First one of them is "cumulant functions are infinitely differentiable and are therefore continuously differentiable". This is because  $c(\theta) = \log\left(\int \exp\left(\langle Y(w),\theta\rangle\right)\lambda(dw)\right)$ . Since  $\log()$  and  $\exp()$  functions are infinitely differentiable,  $c(\theta)$  is too. The second result that needs to be proved is " $g^{-1}(\theta)$  is infinitely differentiable". This is because  $\frac{\partial A^{-1}}{\partial x} = -A^{-1}\frac{\partial A}{\partial x}A^{-1}$ . Hence, matrix inversion is infinitely differentiable and so is  $g^{-1}(\theta)$  since  $g(\theta) = \nabla c(\theta)$  is infinitely differentiable.

**Problem 3**: Let y be a regular full exponential family with canonical parameter  $\theta$ . Verify that y is sub-exponential.

**Problem 4**: In the notes it was claimed that the scalar products of  $\sum_{i=1}^{n} \{y_i - \nabla c(\theta)\}$  are also sub-exponential (see the "Finite sample concentration of MLE" section in the exponential family notes). Show that this is in fact true when the observations  $y_i$  are iid realizations from a regular full exponential family.

**Problem 5**: Derive the MLEs of the canonical parameters of the binomial distribution and the normal distribution with both  $\mu$  and  $\sigma^2$  unknown.

**Problem 6**: Derive the asymptotic distribution for  $\hat{\tau}$ , the MLE of the submodel mean value parameter vector. Hint: use the Delta method.

**Problem 7**: Prove Lemma 1 in the exponential family notes.