

## Homework - October 1, 2024

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### Problem 1: Bernoulli Distribution

Consider tossing a coin for a few times, where the outcomes follow a Bernoulli distribution parametrized by  $\pi$ . In other words, the probability of having a head is  $\pi$ , whereas the probability of having a tail is  $1 - \pi$ .

Suppose we have  $N_1$  heads and  $N_2$  tails. Give the formula for the likelihood of  $\pi$ . Show that the negative log-likelihood is convex in  $\pi$ .

Recall that the likelihood function  $\mathcal{L}$  is the same as the probability density function. So for a Bernoulli distribution we have

$$L(\pi) = \pi^{N_1}(1 - \pi)^{N_2}$$

Which is the joint probability of observing the exact number of heads and tails we obtained in our trial.

We can show the negative-log-likelihood function is convex through the definition of concavity or more simply through a second derivative test:

$$\begin{aligned} -\log L &= \mathcal{L} = -(N_1 \log(\pi) + N_2 \log(1 - \pi)) \\ \frac{d\mathcal{L}}{d\pi} &= -\frac{N_1}{\pi} + \frac{N_2}{1 - \pi} \\ \frac{d^2\mathcal{L}}{d\pi^2} &= \frac{N_1}{\pi^2} + \frac{N_2}{(1 - \pi)^2} \end{aligned}$$

Both terms are positive in the range  $\pi \in (0, 1)$ , so the function is indeed convex. We can also prove this using the definition of convexity.

### Problem 2: Constraints

Consider the training objective  $J = \|Xw - t\|^2$  with a constraint  $\|w\|^2 \leq C$  for some constant  $C$ . How would the hypothesis class capacity, overfitting/underfitting and bias/variance vary according to  $C$ ?

	Larger $C$	Smaller $C$
Model Capacity	Larger	Smaller
Overfitting/Underfitting	Overfitting	Underfitting
Bias	Low	High
Variance	High	Low