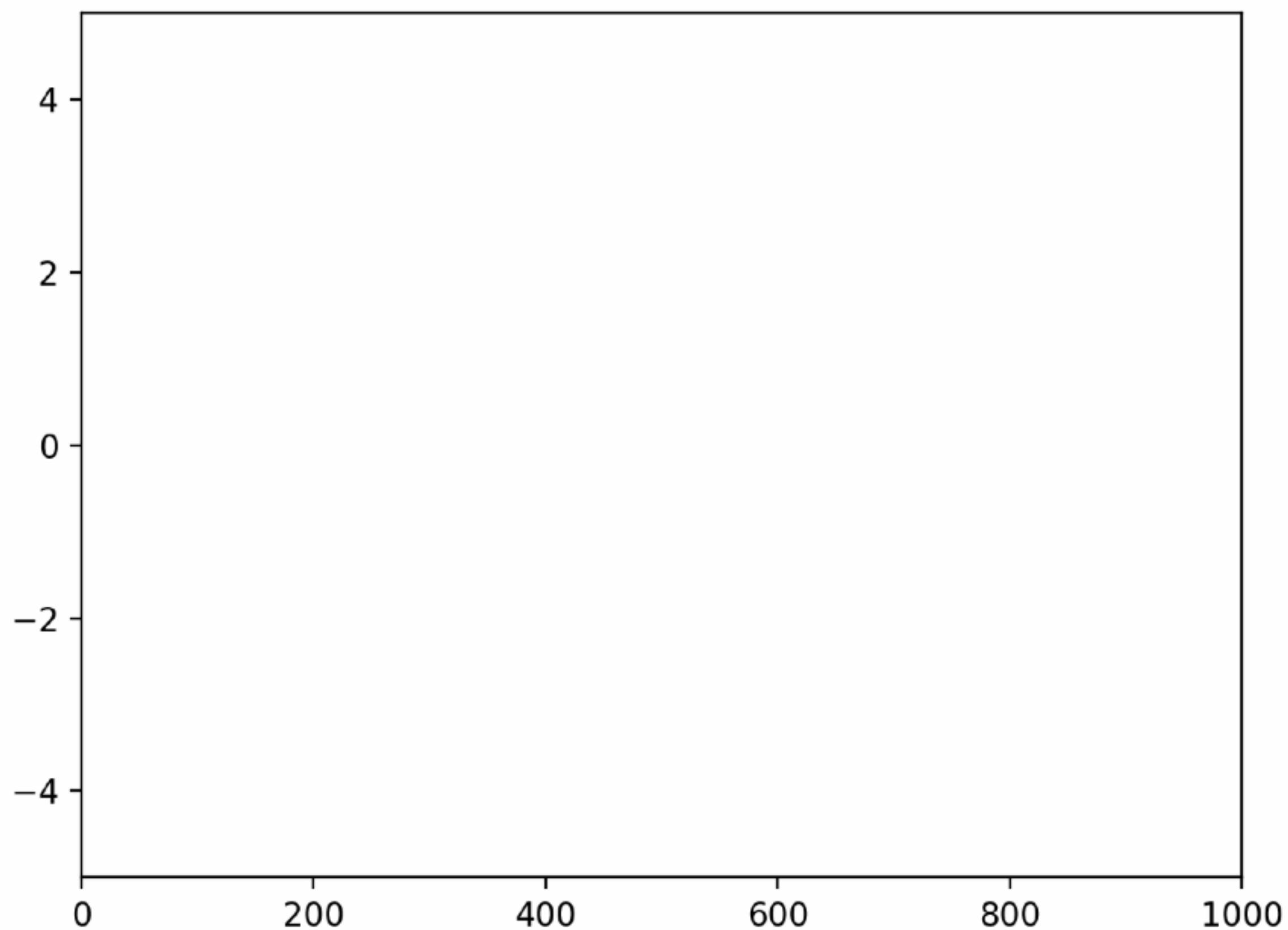
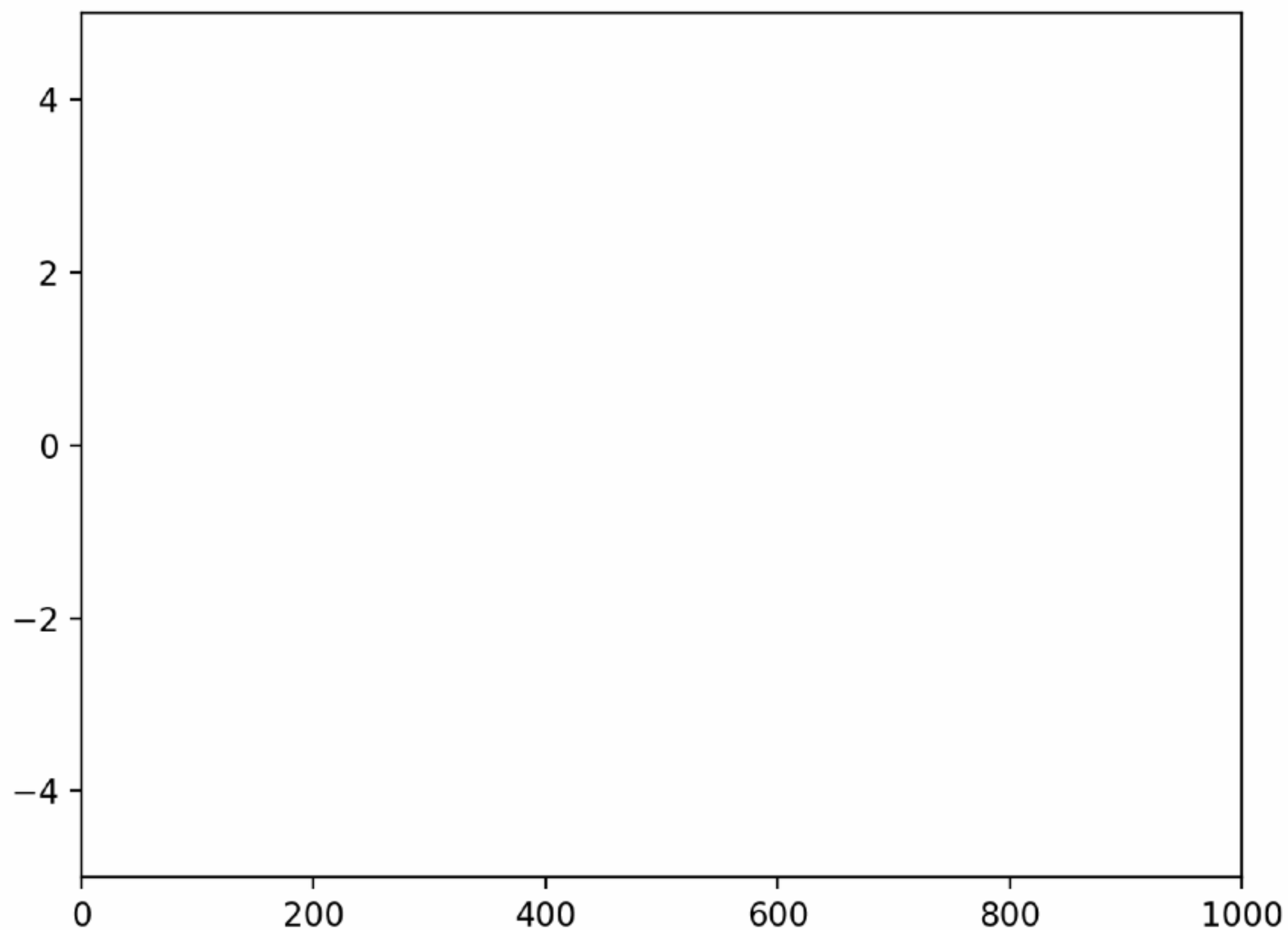


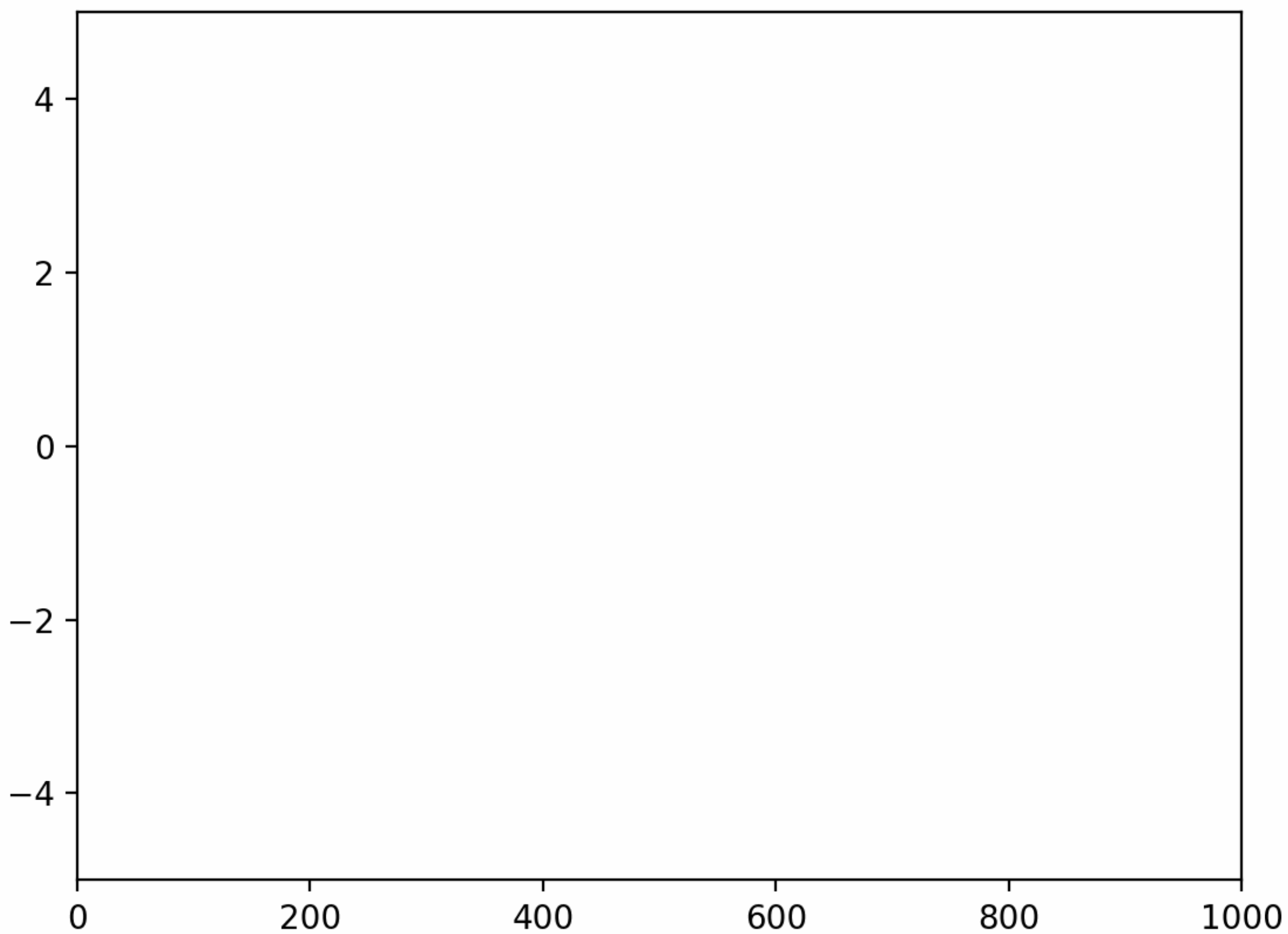
UcB working



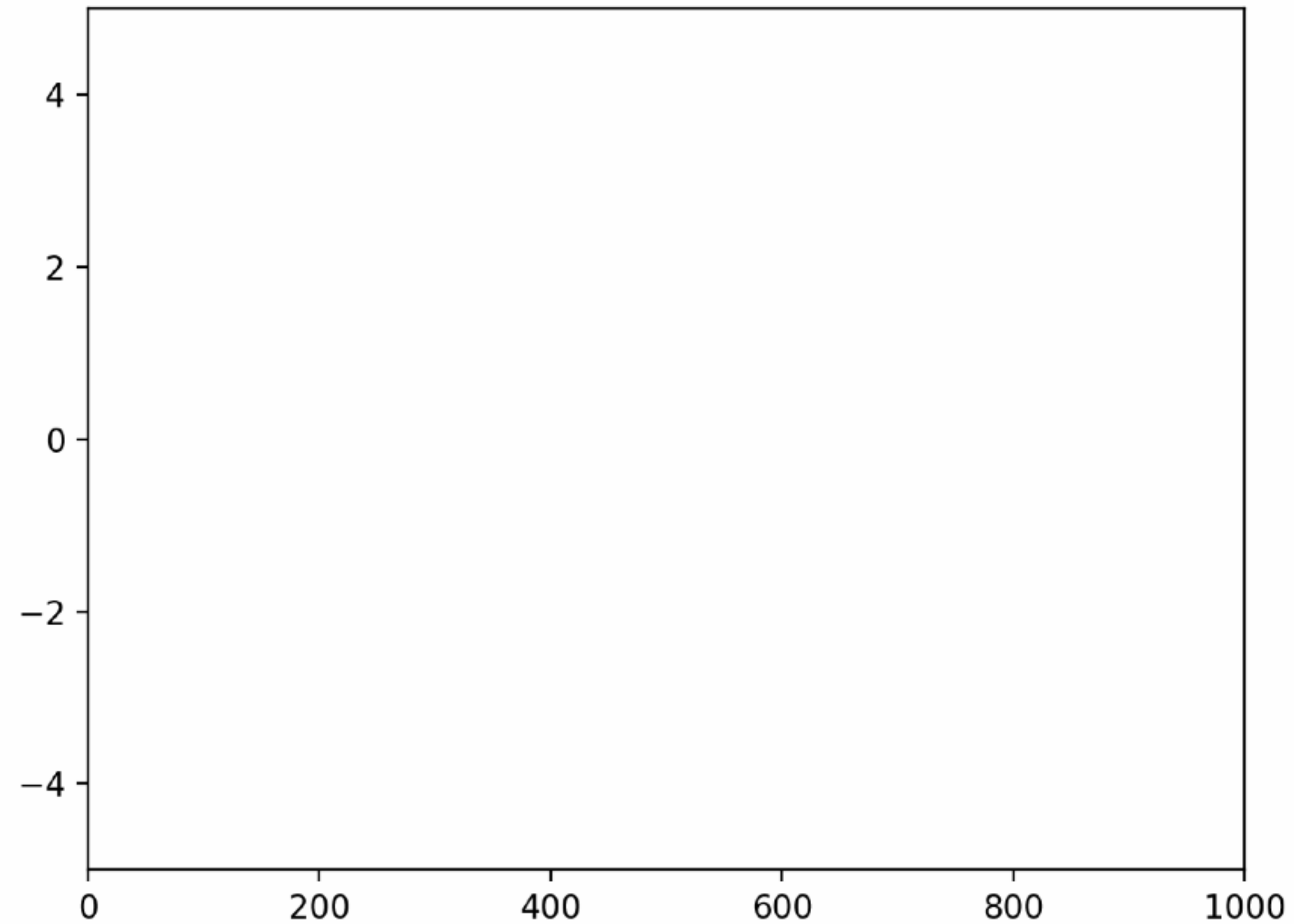
UCB working for two coins with probability

of head 0.9 (red) and 0.5 (blue)



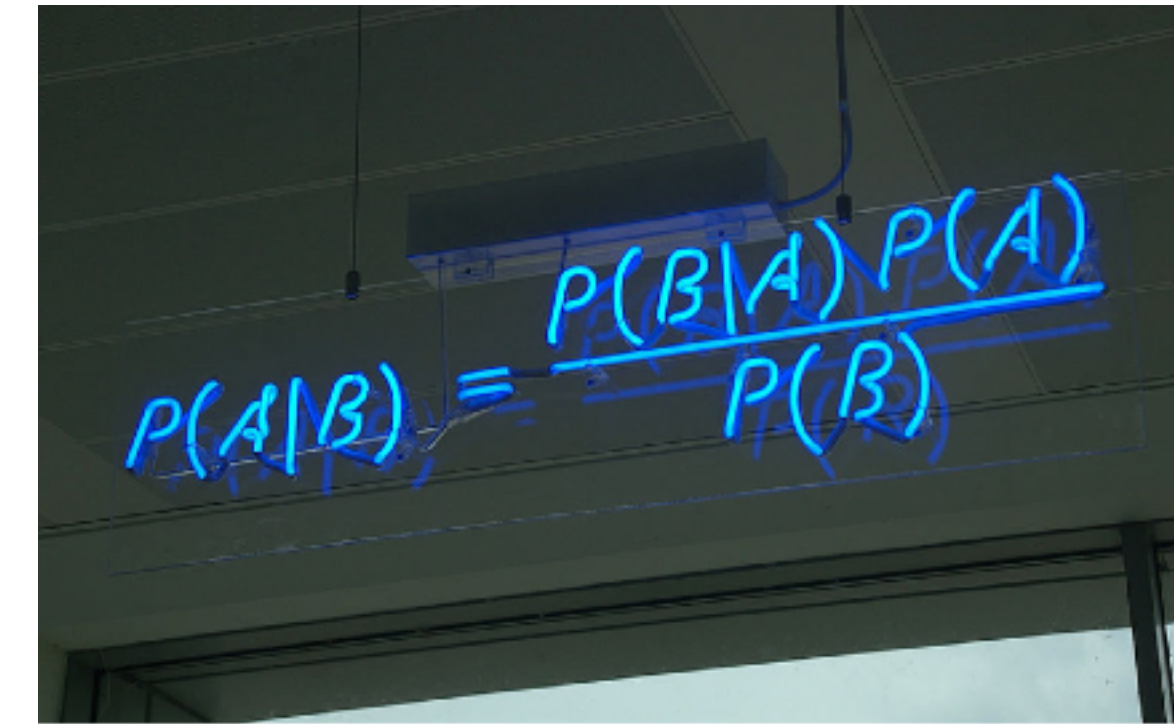


UCB working



UCB working for two coins with probability
of head 0.9 (**red**) and 0.5 (**blue**)

Bayesian Inference



A photograph of a chalkboard with the formula for Bayes' theorem written in blue chalk: $P(A|B) = \frac{P(B|A)P(A)}{P(B)}$.

- Suppose you have data sampled from a parametric distribution with parameter θ .
- Our goal is to estimate θ , given the data samples.

$$\mathbb{P}(\theta | \text{data}) = \frac{\mathbb{P}(\text{data} | \theta) \times \mathbb{P}(\theta)}{\mathbb{P}(\text{data})}$$

- **Posterior distribution**: Represents what you know after having seen the data.
- **Likelihood**: How likely is the data generated from the distribution with parameter θ
- **Prior**: Represents what you know before seeing the data.
- $\mathbb{P}(\text{data}) = \int \mathbb{P}(\text{data} | \theta) \mathbb{P}(\theta) d\theta$: This is a normalisation factor, which is independent of θ . In general, this quantity is very difficult to compute.