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GalvanizeU

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Warm-up discussion

How did the lab go?

What did you learn about issues with MC?

Learning objectives

- Review the difference between a PDF and an RV. (Lecture)
- Understand the two cases in which Importance Sampling is useful. (Lecture+Lab)
- Know what the proposal distribution is for Importance Sampling. (Lecture+Lab)
- Be able to describe the Importance Sampling procedure. (Lecture+Lab)

Recall Random Variables

Recall that an RV is made of two components:

p(x): the probability density function

f(x): the outcome function

So, the expectation is defined as:

$$E[X] = \int_a^b f(x)p(x)dx$$

We will use this separation of concerns in Importance Sampling.

Querying revisited

With Monte Carlo integration, we can approximate E[X].

But MCI relies on drawing from X.

What if we can't even do that?

- \triangleright E[X]: difficult
- ▶ $a \sim X$: difficult
- ightharpoonup P(X=a): easy

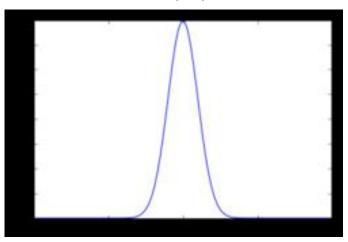
Importance Sampling to the rescue.

Big idea:

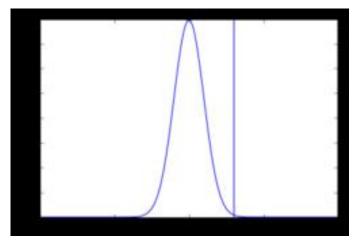
We use an easy-to-sample distribution to approximate a difficult-to-sample one.

Intuitive example

Let's say we have $X \sim \mathcal{N}(0,1)...$



And we want to know P(X >= 3):



Discuss: what will MCI do?

General procedure

We can posit an easy approximating distribution, call it q.

Then...

We sample from q, then...

We *correct* the samples from q towards our original distribution.

How do we correct the samples?

Let's draw a sample from q: $a \sim q$

That sample naturally has a probability in both p and q:

p(a)

q(a)

(Remember, we can use the CDF of both distributions.)

Then, we weight a by the ratio

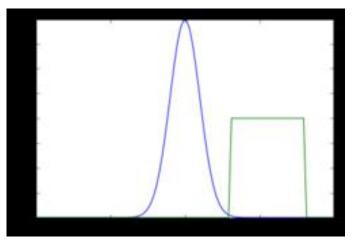
$$\frac{p(a)}{q(a)}$$

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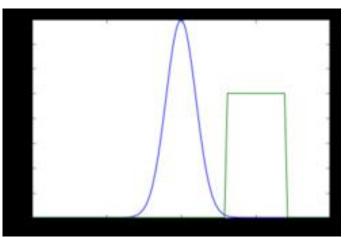
We call that ratio the *importance weight* of a:

$$w(a) = \frac{p(a)}{q(a)}$$

Example: $p = \mathcal{N}(0,1), q = \mathcal{U}(3,8)$



Example: $p = \mathcal{N}(0,1), q = \mathcal{U}(3,6)$



Example:
$$p = \mathcal{N}(0,1), q = \mathcal{U}(3,10)$$

