Assignment 1 for #70240413 "Statistical Machine Learning"

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1 Mathematics Basics

Choose one problem from the 1.1 and 1.2. A bonus would be given if you finished the both.

1.1 Calculus

The gamma function is defined by (assuming x > 0)

$$\Gamma(x) = \int_0^\infty u^{x-1} e^{-u} du. \tag{1}$$

- (1) Prove that $\Gamma(x+1) = x\Gamma(x)$.
- (2) Also show that

$$\int_0^1 u^{a-1} (1-u)^{b-1} du = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}.$$
 (2)

Solution: For Question (1), we can prove it by Using integration by parts, the steps are as follows:

$$\Gamma(x+1) = \int_0^\infty u^x e^{-u} \, du$$

$$= \left[-u^x e^{-u} \right]_0^\infty + \int_0^\infty x u^{x-1} e^{-u} \, du$$

$$= \lim_{u \to \infty} (-u^x e^{-u}) - (0e^{-0}) + x \int_0^\infty u^{x-1} e^{-u} \, du$$

$$= x \int_0^\infty u^{x-1} e^{-u} \, du$$

$$= x \Gamma(x)$$
(3)

As we know, when $u \to \infty$, $-u^x e^{-u} \to 0$, so the equation is proved.

Solution: For Question (2), we know that the left of the equation is a Beta function. From the definitions, we can express the equation which we want to prove as:

$$\Gamma(a+b)B(a,b) = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)} \tag{4}$$

It's a double integral, the expansion formula is as follows:

$$\Gamma(a+b)B(a,b) = \int_0^\infty u^{a+b-1}e^{-u}du \int_0^1 v^{a-1}(1-v)^{b-1}dv$$

$$= \int_0^\infty \int_0^1 (uv)^{a-1}[u(1-v)]^{b-1}ue^{-u}du dv$$
(5)

Then we do a transformation w = uv, z = u(1-v). The inverse transformation is u = w+z, v = w/(w+z), the corresponding ranges of them are $w \in (0, \infty)$ and $u \in (0, \infty)$. The absolute value of the Jacobian is

$$\left| \nabla \frac{\partial(u,v)}{\partial(w,z)} \right| = \frac{1}{(w+z)} \tag{6}$$

Next, we use the changed of variables to do a double integral, the equation above becomes:

$$\int_{0}^{\infty} \int_{0}^{\infty} w^{a-1} z^{b-1} (w+z) e^{-(w+z)} \frac{1}{w+z} dw dz
= \int_{0}^{\infty} \int_{0}^{\infty} w^{a-1} z^{b-1} e^{-(w+z)} dw dz
= \int_{0}^{\infty} w^{a-1} e^{-w} dw \int_{0}^{\infty} z^{b-1} e^{-z} dz
= \Gamma(a) \Gamma(b)$$
(7)

Finally the equation is proved.

1.2 Optimization

Choose one problem from the following 1.3 and 1.4. A bonus would be given if you finished the both.

1.3 Stochastic Process

We toss a fair coin for a number of times and use H(head) and T(tail) to denote the two sides of the coin. Please compute the expected number of tosses we need to observe a first time occurrence of the following consecutive pattern

$$H, \underbrace{T, T, ..., T}_{k}. \tag{8}$$

Solution: we asume that E is the expection of the consecutive pattern $H, \underbrace{T, T, ..., T}_{k}$, and E_{T}^{k} is the expection of $\underbrace{T, T, ..., T}_{k}$, then we have

$$\begin{cases}
E = 1 + \frac{1}{2}E + \frac{1}{2}E_T^k, \\
E_T^k = E_T^{k-1} + 1 + \frac{1}{2}E_T^k + \frac{1}{2} \times 0,
\end{cases} \tag{9}$$

which

1.4 Probability

Suppose $p \sim Beta(p|\alpha, \beta)$ and $x|p \sim Bernoulli(x|p)$. Show that $p|x \sim Beta(p|\alpha + x, \beta + 1 - x)$, which implies that the Beta distribution can serve as a conjugate prior to the Bernoulli distribution.

Solution: Consider calculating the posterior p|x, and we know the likelihood function x|p and the prior p, here we use Bayes' theorem:

$$P(p|x) = \frac{P(x|p)P(p)}{P(x)}$$

$$= \frac{P(x|p)P(p)}{\int P(x|p')P(p')dp'}$$
(10)

From the definition, $P(p) \sim Beta(p|\alpha,\beta)$ and $P(x|p) \sim Bernoulli(x|p)$, and the Beta function is

$$Beta(p|\alpha,\beta) = \frac{1}{B(\alpha,\beta)} p^{\alpha-1} (1-p)^{\beta-1}$$
(11)

so P(p|x) should be

$$P(p|x) = \frac{P(x|p)P(p)}{\int P(x|p')P(p')dp'}$$
(12)

Differentiation allows for the calculation of the slope of the tangent of a curve at any given point as shown in Figure 1.

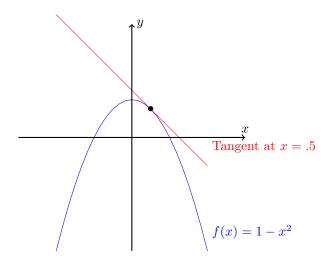


Figure 1: The plot of $f(x) = 1 - x^2$ with a tangent at x = .5.

Differentiation is now a technique taught to mathematics students throughout the world. In this document I will discuss some aspects of differentiation.

2 Exploring the derivative using Sage

The definition of the limit of f(x) at x = a denoted as f'(a) is:

$$f'(a) = \lim_{h \to 0} \frac{f(a+h) - f(a)}{h} \tag{13}$$

The following code can be used in sage to give the above limit:

def illustrate(f, a):

```
Function to take a function and illustrate the limiting definition of a derivative at a given po
"""

lst = []
for h in srange( 01 3 01):
```

```
for h in srange(.01, 3, .01):
    lst.append([h,(f(a+h)-f(a))/h])
return list_plot(lst, axes_labels=['$x$','$\\frac{f(%.02f+h)-f(%.02f)}{h}$' % (a,a)])
```

Figure 2: The derivative of $f(x) = 1 - x^2$ at x = .5 converging to -1 as $h \to 0$.

If we want to plot the tangent at a point α to a function we can use the following:

```
y = ax + b (definition of a straight line)

f'(a)x + b (definition of the derivative)

f'(a)x + f(a) - f'(a)a (we know that the line intersects f at (a, f(a))
```

We can combine this with the approach of the previous piece of code to see how the tangential line converges as the limiting definition of the derivative converges:

```
def convergetangentialline(f, a, x1, x2, nbrofplots=50, epsilon=.1):
    """
    Function to make a tangential line converge
    """
    clrs = rainbow(nbrofplots)
    k = 0
    h = epsilon
    p = plot(f, x, x1, x2)
    while k < nbrofplots:
        tangent(x) = fdash(f, a, h) * x + f(a) - fdash(f, a, h) * a
        p += plot(tangent(x), x, x1, x2, color=clrs[k])
        h += epsilon
        k += 1
    return p</pre>
```

The plot shown in Figure 3 shows how the lines shown converge to the actual tangent to $1 - x^2$ as x = 2 (the red line is the 'closest' curve).

Figure 3: Lines converging to the tangent curve as $h \to 0$.

Note here that the last plot is given using the **real** definition of the derivative and not the approximation.

3 Conclusions

In this report I have explored the limiting definition of the limit showing how as $h \to 0$ we can visualise the derivative of a function. The code involved https://sage.maths.cf.ac.uk/home/pub/18/ uses the differentiation capabilities of Sage but also the plotting abilities.

There are various other aspects that could be explored such as symbolic differentiation rules. For example:

$$\frac{dx^n}{dx} = (n+1)x^n \text{ if } x \neq -1$$

Furthermore it is interesting to not that there exists some functions that **are not** differentiable at a point such as the function $f(x) = \sin(1/x)$ which is not differentiable at x = 0. A plot of this function is shown in Figure 4.

Figure 4: None differentiable function at x = 0.