

CS 5135/6035 Learning Probabilistic Models

Exercise Questions for Lecture 19: Accept-Reject Method; Monte Carlo Integration

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Questions

1. Draw samples from a density $f(x) \propto x^{\alpha-1}(\frac{1-x}{1+x})^{\beta-1}$ ($\alpha = 2.7$ and $\beta = 6.3$) using Accept-reject method using a Beta distribution with ($\alpha = 2$ and $\beta = 2$) as a candidate.
[15 points]
 - a. List the *three criteria* you verified to ensure that the candidate density is suitable. Provide results from your verification.
[Hint: To test compatibility in terms of support, use visualization. Use Julia code to determine the value of M .]
 - b. Write psuedo code for the Accept-reject method in this context.
 - c. Write Julia code to generate samples from $f(x)$.
 - d. Plot the histogram for samples generated from target and candidate densities.
 - e. What fraction of the total samples were accepted?
2. Draw samples from a density $f(x) \propto x^{\alpha-1}(\frac{1-x}{1+x})^{\beta-1}$ ($\alpha = 2.7$ and $\beta = 6.3$) using Accept-reject method using a Beta distribution with ($\alpha = 2$ and $\beta = 5$) as a candidate.
[6 points]
 - a. List the *three criteria* you verified to ensure that the candidate density is suitable. Provide results from your verification.
[Hint: To test compatibility in terms of support, use visualization. Use Julia code to determine the value of M .]
 - b. Write Julia code to generate samples from $f(x)$.
 - c. What fraction of the total samples were accepted?
3. Between the candidate distributions used in (1) and (2), which of them results in a tighter envelope?
[4 points]
 - a. How do you determine this using the value of M in both cases?
 - b. How do determine this using the *fraction of accepted samples*?
 - c. What is the advantage of using a tighter envelope?

Bonus question

1. Use Accept-reject method to generate samples from a Poisson distribution with $\lambda = 2$ using a Uniform distribution as a candidate.
 - a. List the three criteria and the results from your verification.
 - b. Write psuedo code for the Accept-reject method in this context.
 - c. Write Julia code to generate samples.
 - d. Compare histogram of these samples with that of random samples generated using `'rand(Poisson(2),10000)'`

Sample code

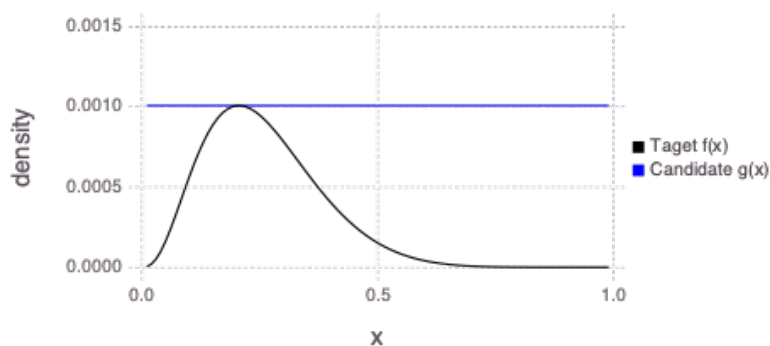
1. Julia code for visualizaing target and candidate densities

```
x = collect(0.01:0.01:0.99);
d = Uniform(0,1); # candidate density

## Distributions.Uniform{Float64}(a=0.0, b=1.0)
f(x) = x.^2.7 .* (((1.-x)./(1.+x)).^6.3);
g(x) = pdf(d,x);
M = maximum(f.(x)./g.(x))

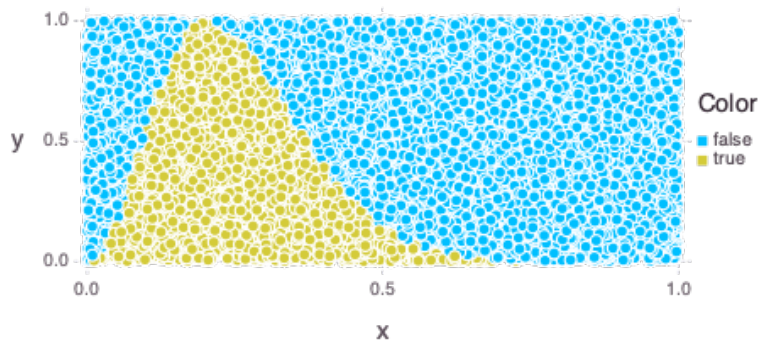
## 0.0010080798592428059

myplot = plot(
    layer(x=x,y=f.(x),Geom.line,Theme(default_color=colorant"black")),
    layer(x=x,y=M.*g.(x),Geom.line,Theme(default_color=colorant"blue")),
    Guide.ylabel("density"),Guide.xlabel("x"),
    Guide.manual_color_key("", ["Target f(x)", "Candidate g(x)"], ["black","blue"]));
draw(PNG("./figs/viz_f_g.png", 5inch, 2.5inch), myplot);
```



2. Accept-reject method

```
n = 10000;
y = rand(d,n);
u = rand(Uniform(0,1),n);
x_samples = y[u.<f.(y)./(M*g.(y))];
myplot = plot(x=y,y=u,color = u.<f.(y)/M,Geom.point);
draw(PNG("./figs/accept_reject_samples.png", 5inch, 2.5inch), myplot);
```



3. Plotting histograms of samples

```

samples1 = y;
hist1 = [fit(Histogram,samples1,x).weights; 0]./96500;
samples2 = x_samples;
hist2 = [fit(Histogram,samples2,x).weights; 0]./99000;

myplot = plot(
    layer(x=x,y=f.(x),Geom.line,Theme(default_color=colorant"black")),
    layer(x=x,y=M.*g.(x),Geom.line,Theme(default_color=colorant"blue")),
    layer(x=x,y=hist2, Geom.bar,
        Theme(default_color=colorant"SteelBlue")),
    layer(x=x,y=hist1, Geom.bar,
        Theme(default_color=colorant"orange")),
    Coord.Cartesian(xmin=0, xmax=1),
    Guide.ylabel("density"),Guide.xlabel("x"),
    Guide.manual_color_key("", ["Taget f(x) ", "Candidate Mg(x)", "Samples g(x)",
        "Samples f(x)"], ["black","blue","orange","SteelBlue"]));
draw(PNG("./figs/hist_target_candidate_samples.png", 5inch, 2.5inch), myplot);

```

