```
clear; clc; close all;
p=0.8; n=30;
theta = linspace(0,1,n); lower=zeros(10000,3); upper=zeros(10000,3);
coverage=zeros(10000,3); Intlength=zeros(10000,3);
for i=1:10000
R = binornd(1,p,1,n);
%Frequentist, Using the Wald Interval
mean1=sum(R)/length(R);
lower(i,1) = mean1 - 1.96*sqrt(mean1*(1-mean1)/length(R));
upper(i,1) = mean1 + 1.96*sqrt(mean1*(1-mean1)/length(R));
%Uniform- Prior Bayesian, it is actually beta(1,1)distribution
[mean2, sigma2] = betastat(1+sum(R), 1+length(R)-sum(R));
lower(i, 2) = mean2 - 1.96*sqrt(sigma2/length(R));
upper(i,2) = mean2 + 1.96*sqrt(sigma2/length(R));
%Beta(8,2)-Prior Bayesian
[mean3, sigma3] = betastat(8+sum(R), 2+length(R) - sum(R));
lower(i,3) = mean3 - 1.96*sqrt(sigma3/length(R));
upper(i,3) = mean3 + 1.96*sqrt(sigma3/length(R));
for k=1:3
    Intlength(i,k) = upper(i,k) - lower(i,k);
    if lower(i,k) <= 0.8 && upper(i,k) >= 0.8
        coverage(i, k)=1;
    end
end
end
%%%OUTPUT
CoverageFrequentist=sum(coverage(:,1),1)/size(coverage,1)
CoverageUniform=sum(coverage(:,2),1)/size(coverage,1)
CoverageBeta82=sum(coverage(:,3),1)/size(coverage,1)
IntervalLengthFrequentist=mean(Intlength(:,1))
IntervalLengthUniform=mean(Intlength(:,2))
IntervalLengthBeta82=mean(Intlength(:,3))
```

N=30	Frequentist	Uniform	Beta(8,2)
Coverage Rate	0.9447	0.3448	0.1709
Avg. Interval Length	0.2785	0.0505	0.0441
N=5	Frequentist	Uniform	Beta(8,2)
Coverage Rate	0.6649	0.7380	0.9398
Avg. Interval Length	0.5117	0.2661	0.1724

Comments:

- 1. We can see, when n=30, frequentists has bigger coverage rate but bigger interval length, which means they are mostly distributed around the true p in a big range; Bayesians has smaller coverage rate but smaller interval length, which means they are mostly concentrated in smaller range, but the peak seldom falls near true p.
- 2. When n=5, compared n=30, frequentists gets both bigger confidence interval and less coverage rate; Bayesians, however, has bigger coverage rate and smaller confidence interval, apparently better than the frequentist prior estimate.
- 3. In fact, the random Bernoulli generates p=0.5 on average, which redirect the posterior p from 0.8 to 0.5. Under such circumstances, the coverage rate defined in the problem is no longer very meaningful, as the prior estimate changes.