Pulse arrival time uncertainty propagation

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This code was developed by Miodrag Bolic for the book PERVASIVE CARDIAC AND RESPIRATORY MONITORING DEVICES.

Acknowledgements:

Uncertainty propagations is based on the code: Joe Klebba (2021). Uncertainty Propagation Functions (https://www.mathworks.com/matlabcentral/fileexchange/89812-uncertainty-propagation-functions), MATLAB Central File Exchange. Retrieved May 19, 2021

Fitting PAT vs systolic data

Fitting is done based on regression SBP=a*In(PAT)+b

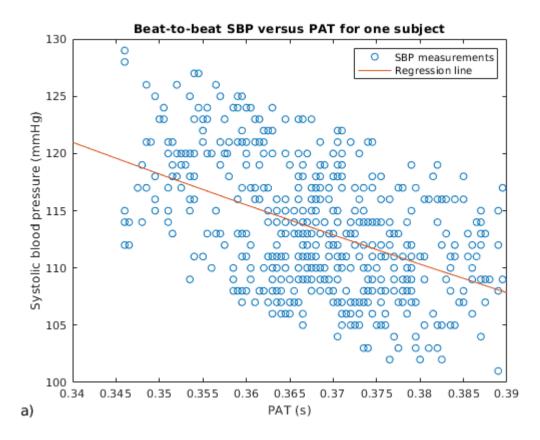
Let us assume that the physiological signals used to obtain PAT values are sampled at 250 Hz which means that the sampling period is 4 ms. We will present the error in timing using uniform distribution in the range from -0.002s to 0.002s.

```
close all
clear_all_but('SAVE_FLAG')
fprintf('\nExample: PAT\n')
```

Example: PAT

```
load('patsbp.mat')
fo = fitoptions('Method','NonlinearLeastSquares',...
                'Lower',[-100,0],...
                'Upper',[-80,inf],...
                'StartPoint',[-90 18]);
linearfittype = fittype({'log(x)','1'},'options',fo);
fit_res = fit(patsbp.PAT,patsbp.SBP,linearfittype)
fit_res =
    Linear model:
    fit_res(x) = a*log(x) + b
    Coefficients (with 95% confidence bounds):
     a =
           -95.74 (-111.3, -80.2)
              17.7 (2.168, 33.23)
      b =
ci = confint(fit_res, 0.67);
std_a=abs(ci(1,1)-fit_res.a) % obtain standard deviation for parameter a
std_a = 7.7091
std_b=abs(ci(2,2)-fit_res.b) % obtain standard deviation for parameter b
std_b = 7.7073
figure
plot(patsbp.PAT,patsbp.SBP,'o')
xlabel("PAT (s)")
ylabel("Systolic blood pressure (mmHg)")
hold on
i=0.34:0.01:0.39;
plot(i,fit_res(i))
legend({'SBP measurements', 'Regression line'})
title('Beat-to-beat SBP versus PAT for one subject')
```

annonation_save('a)', "Fig8.6a.jpg", SAVE_FLAG);



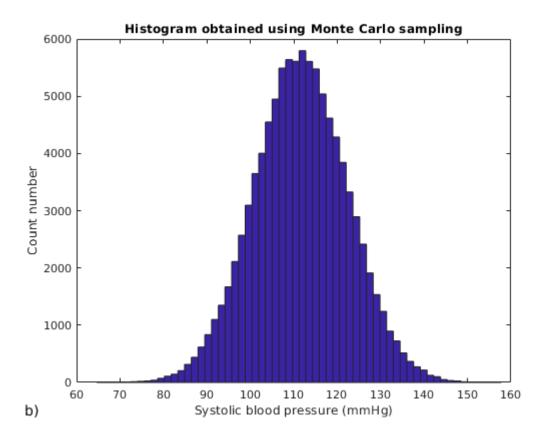
Defining the uncertainties of the parameters

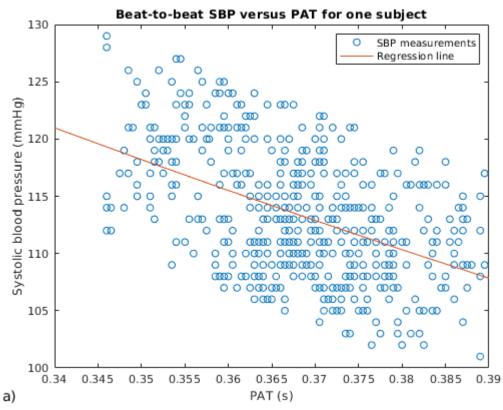
```
PAT=0.375;
b=fit_res.b; ub=std_b; % these values are obtained after fitting
a=fit_res.a; ua=std_a; % these values are obtained after fitting
f = @(a,b,x)a.*log(x)+b;
n=100000;
PAT_data=PAT*ones(n,1)+unifrnd(-0.002,0.002,n,1)-unifrnd(-0.002,0.002,n,1);
```

Performing uncertainty propagation

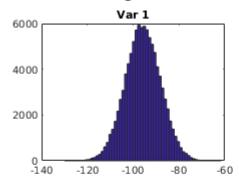
```
[CI,valMC]=propUncertMC(f,{{a ua};{b,ub};{'Custom',PAT_data}},n,'mean','varHist',60,'he
ylabel("Count number")
xlabel("Systolic blood pressure (mmHg)")

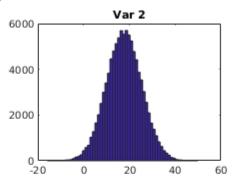
title('Histogram obtained using Monte Carlo sampling')
annonation_save('b)',"Fig8.6b.jpg", SAVE_FLAG);
```

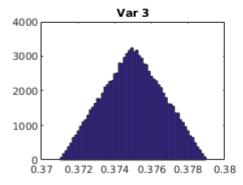




Histogram Of MC Samples For Each Variable







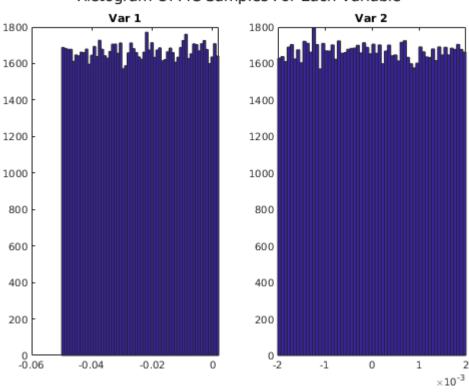
<u>Excersize</u>: repeat uncertainty quantification for the case when uncertainties in estimating coefficients are negligible: ub=ua=0.0001. What is the standard deviation and 95% confidence intervals in this case.

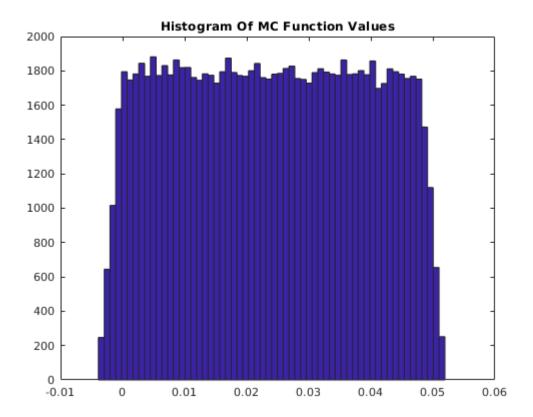
Averaging the estimated over time

```
T=15; %averaging over T pulsees
uncertMC_T=uncertMC/sqrt(15)
uncertMC_T = 2.7682
```

Example of calculating uncertainty subtraction of two uniform random variable

Histogram Of MC Samples For Each Variable





uncertMC = (CI(2)-CI(1))/2 % this is a triangular distribution and therefore this way

uncertMC = 0.0176

valMC

valMC = 0.0240