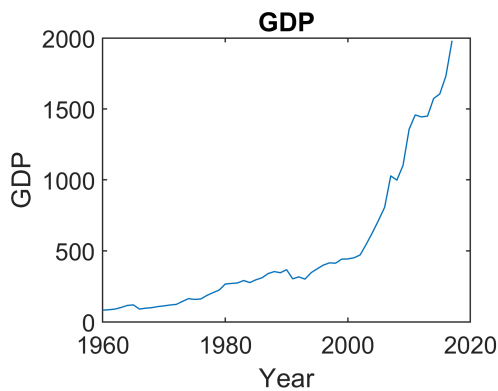
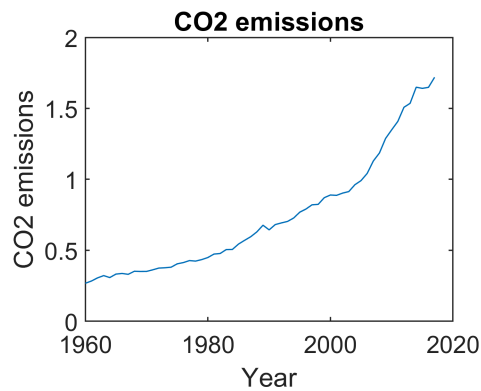
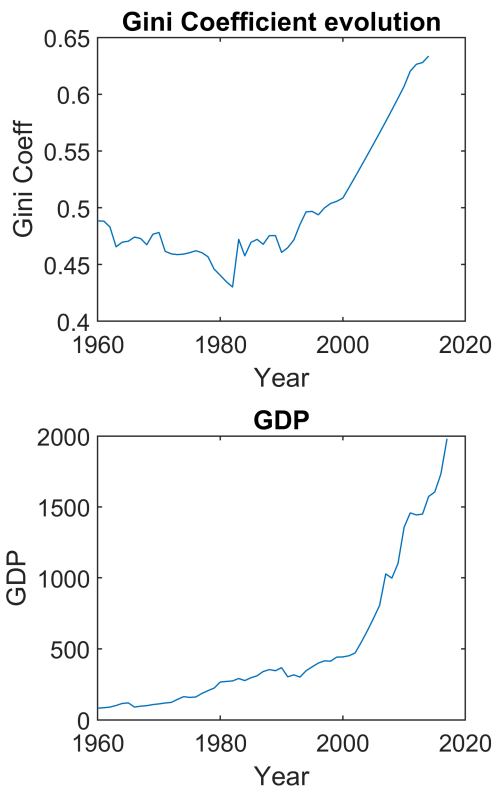


```
clear; close all;
```

## Open data

```
path = "./Data/Gini_index.xlsx";
data = table2array(readtable(path));
time = data(10:end-5,1); gini = data(10:end-5,2);
subplot(221);
plot(time,gini);
title('Gini Coefficient evolution');
xlabel('Year'); ylabel('Gini Coeff');
path = "./Data/india_data.xlsx";
% opts = detectImportOptions(path);
T = readtable(path,"ReadRowNames",true);
data = table2array(T);
% Finding the first NAN value and taking data upto the instance before it
cut_off = length(data(:,1));
for i = 1:length(data(:,1))
    if (sum(isnan(data(i,:))))
        cut_off = i-1;
        break;
    end
end
time2 = data(1:cut_off,1);
Data = data(1:cut_off,2:end);
subplot(222);
plot(time2,Data(:,1));
title('CO2 emissions');
ylabel('CO2 emissions'); xlabel('Year');
subplot(223);
plot(time2,Data(:,2));
title('GDP');
ylabel('GDP'); xlabel('Year');
```



```
GDP = Data(1:end-3,1);
CO2 = Data(1:end-3,2);
y = CO2;
X = [gini GDP];
```

## Analysis using OLS

Stabilize Data Values

```
y = log(y);
X = log(X);
[h1,p1] = adftest(y);
```

h1 = 0, so difference the series

```
dlogy = diff(y);
dlogX = X(1:end-1,:);
[h2,p2] = adftest(X(:,1));
```

h2 = 0, so difference the series

```
dlogX(:,1) = diff(X(:,1));
[h3,p3] = adftest(X(:,2));
```

h3 is not 0, but since other 2 are differenced, need to remove one data point

```
dlogX(:,2) = X(2:end,2);

ols_md1 = fitlm(dlogX,dlogy);
res = ols_md1.Residuals.Raw;
ols_md1
```

```
ols_md1 =
Linear regression model:
    y ~ 1 + x1 + x2
```

Estimated Coefficients:

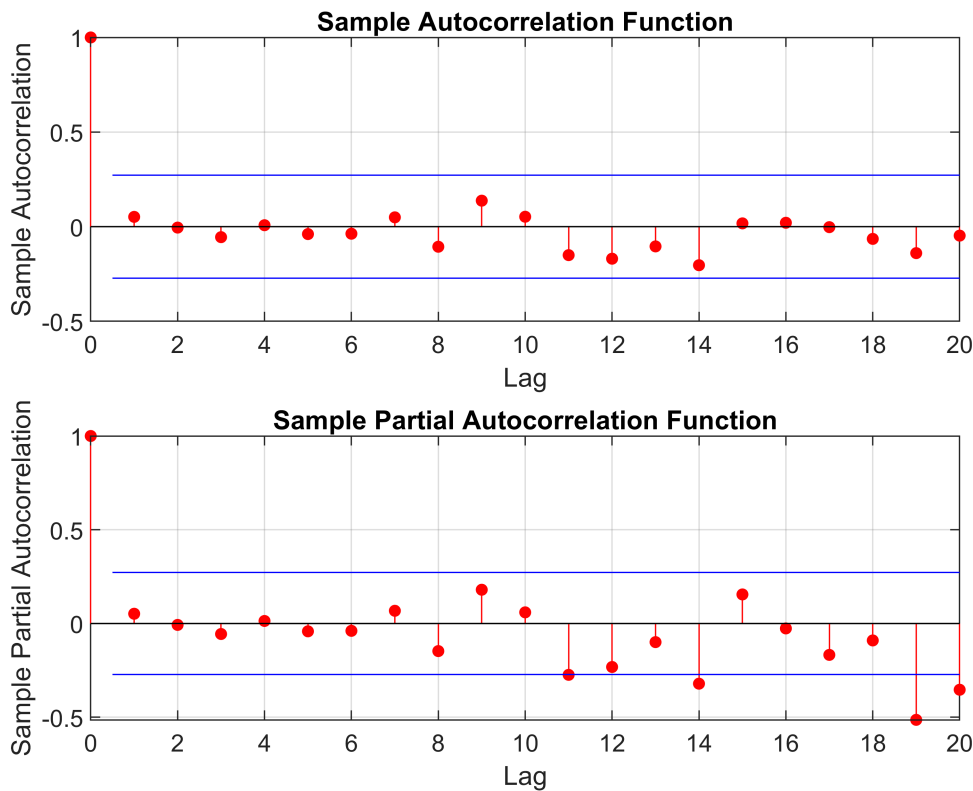
	Estimate	SE	tStat	pValue
(Intercept)	0.065078	0.018354	3.5456	0.00084955
x1	-0.047764	0.64863	-0.073639	0.94159
x2	0.02079	0.025749	0.8074	0.42318

Number of observations: 54, Error degrees of freedom: 51  
Root Mean Squared Error: 0.0867  
R-squared: 0.0137, Adjusted R-Squared: -0.0249  
F-statistic vs. constant model: 0.355, p-value = 0.703

```
fprintf('R-squared = %.4f, Adjusted R-squared = %.4f \n',ols_md1.Rsquared.Ordinary,ols_md1.Rsquared.Adjusted);
```

```
R-squared = 0.0137, Adjusted R-squared = -0.0249
```

```
figure;
subplot(211); autocorr(res);
subplot(212); parcorr(res);
```



Has some PACF at higher lags, but can safely ignore them since our total time is  $\sim 2 \times$  such periods

```
[hres1,pres1] = lbqtest(res);
disp('Whiteness Test for Residuals results');
```

Whiteness Test for Residuals results

```
disp(hres1);disp(pres1);
```

0

0.8872

Hence, Residuals are white

```
[h_adres1,p_adres1] = adtest(res);
disp('Gaussianity Test for Residuals results');
```

Gaussianity Test for Residuals results

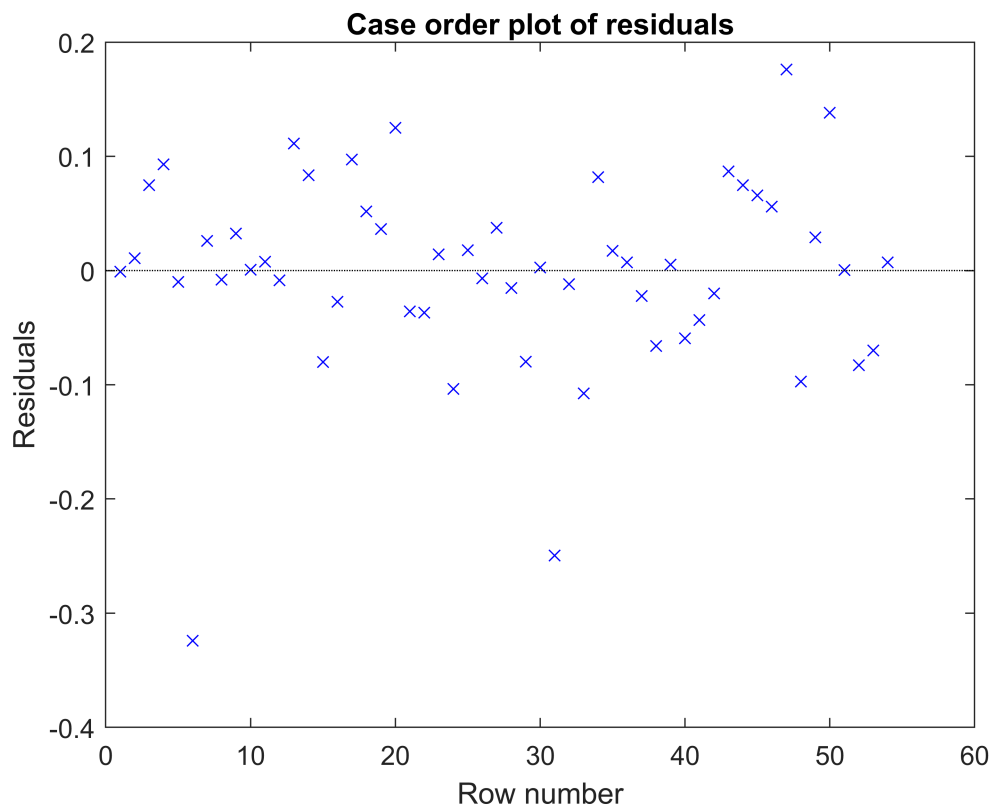
```
disp(h_adres1);disp(p_adres1);
```

1

0.0084

Gaussianity rejected! Residuals are not Gaussian

```
figure;
plotResiduals(ols_md1,'caseorder');
```



Exhibits some heteroskedasticity

## Feasible Generalized least squares

```
[coeff,se,EstCoeffCov] = fgls(dlogX,dlogy,'innovMdl','HC0','display','final');
```

OLS Estimates:

	Coeff	SE
Const	0.0651	0.0184
x1	-0.0478	0.6486
x2	0.0208	0.0257

FGLS Estimates:

	Coeff	SE
Const	0.0662	0.0009
x1	-0.0974	0.0544
x2	0.0216	0.0009

## OLS Analysis without differencing

```
ols_md12 = fitlm(X,y);
res2 = ols_md12.Residuals.Raw;
ols_md12
```

ols\_md12 =  
Linear regression model:

$y \sim 1 + x1 + x2$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	6.4462	0.25198	25.582	3.9611e-31
x1	-0.14444	0.40652	-0.3553	0.72381
x2	1.66	0.082107	20.217	2.711e-26

Number of observations: 55, Error degrees of freedom: 52

Root Mean Squared Error: 0.169

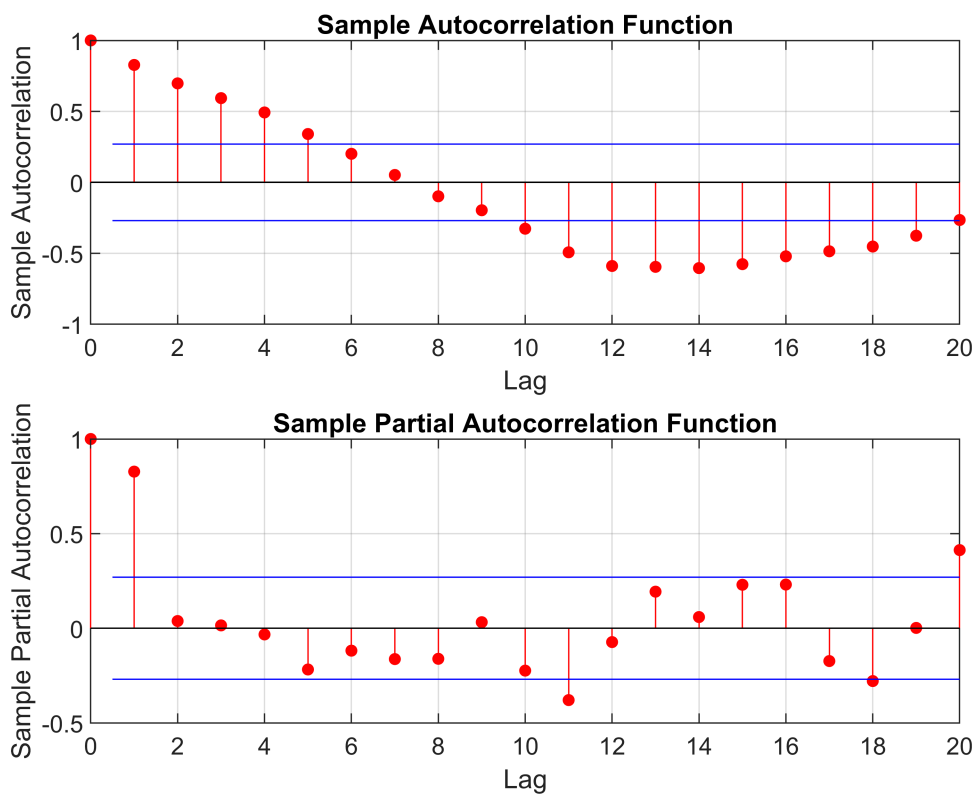
R-squared: 0.962, Adjusted R-Squared: 0.96

F-statistic vs. constant model: 650, p-value = 1.63e-37

```
fprintf('R-squared = %.4f, Adjusted R-squared = %.4f \n',ols_md12.Rsquared.Ordinary,ols_md12.Rs
```

R-squared = 0.9615, Adjusted R-squared = 0.9601

```
figure;
subplot(211); autocorr(res2);
subplot(212); parcorr(res2);
```



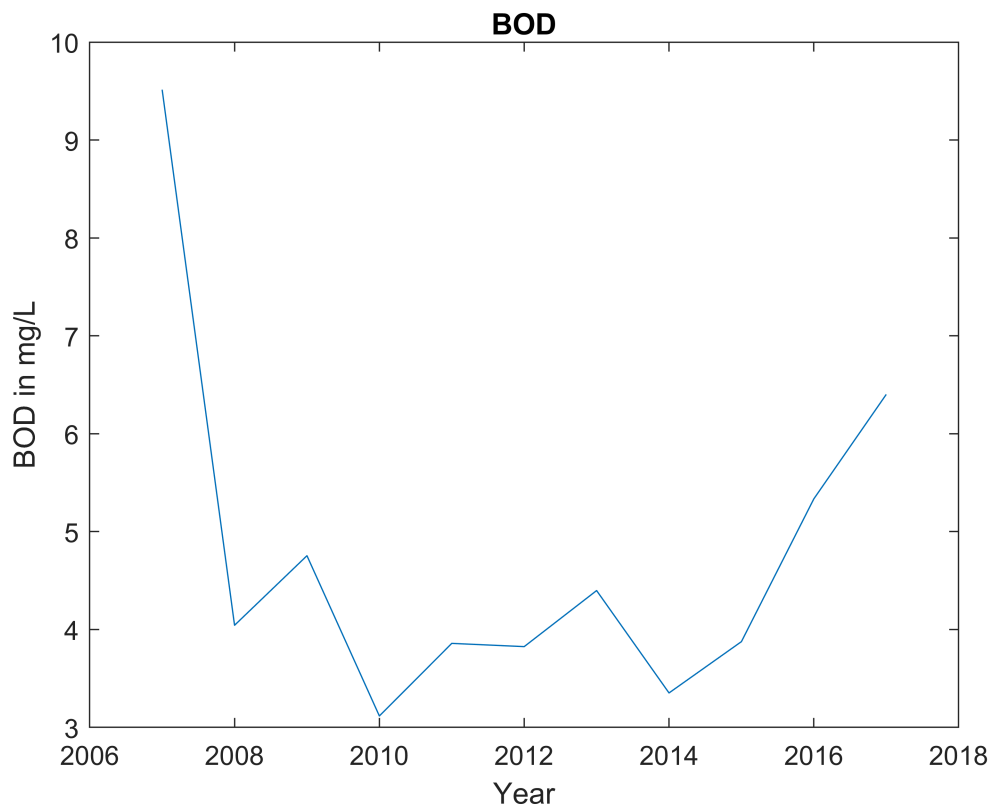
```
% Slowly dying ACF, and immediately dying PACF => MA(1)
[hres2,pres2] = lbqtest(res2);
% Residuals are correlated!
```

**FGLS again**

Not continuing this. However the procedure: fit an MA model to res2, convert to AR using arma2ar. Then find significant terms in AR, denote number of terms as p. Pass that p to FGLS [coeff2,se2,EstCov2] = fgls(X,y,'arlags',1,'display','final');

## BOD x Gini

```
bod = xlsread("./Data/BOD.xlsx","Averaged");
figure;
plot(bod(1,:),bod(2,:));
title("BOD"); ylabel("BOD in mg/L"); xlabel("Year");
```



```
% Since sample size is small (~10) not subjecting it to adftest
bod_data = log(bod(2,1:end-4))'; % because other data till 2014
% Resize matrices
idx = find(time==2007);
dlogy_new = dlogy(idx:end);
dlogX_new = [bod_data,dlogX(idx:end,2)];

% run ols
ols_md1_bod = fitlm(dlogX,dlogy);
res_bod = ols_md1_bod.Residuals.Raw;
ols_md1_bod
```

```
ols_md1_bod =
Linear regression model:
y ~ 1 + x1 + x2
```

```
Estimated Coefficients:
                Estimate          SE          tStat          pValue
```

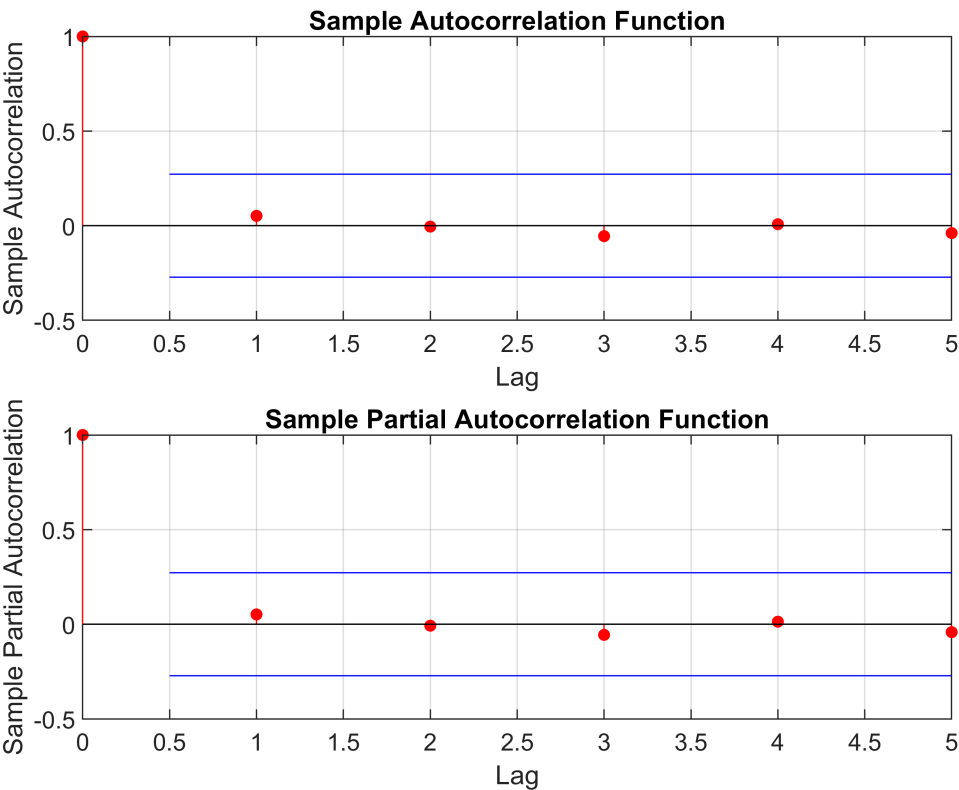
(Intercept)	0.065078	0.018354	3.5456	0.00084955
x1	-0.047764	0.64863	-0.073639	0.94159
x2	0.02079	0.025749	0.8074	0.42318

Number of observations: 54, Error degrees of freedom: 51  
 Root Mean Squared Error: 0.0867  
 R-squared: 0.0137, Adjusted R-Squared: -0.0249  
 F-statistic vs. constant model: 0.355, p-value = 0.703

```
fprintf('R-squared = %.4f, Adjusted R-squared = %.4f \n',ols_mdl_bod.Rsquared.Ordinary,ols_mdl_bod.Rsquared.Adjusted)
```

```
R-squared = 0.0137, Adjusted R-squared = -0.0249
```

```
figure;  
subplot(211); autocorr(res_bod,"NumLags",5);  
subplot(212); parcorr(res_bod,"NumLags",5);
```



No autocorrelation effects!

```
[hresBOD,presBOD] = lbqtest(res_bod);  
disp('Whiteness Test for Residuals results');
```

```
Whiteness Test for Residuals results
```

```
disp(hresBOD);disp(presBOD);
```

```
0
```



0.8872

Hence, Residuals are white

```
[h_adres_BOD,p_adres_BOD] = adtest(res_bod);  
disp('Gaussianity Test for Residuals results');
```

Gaussianity Test for Residuals results

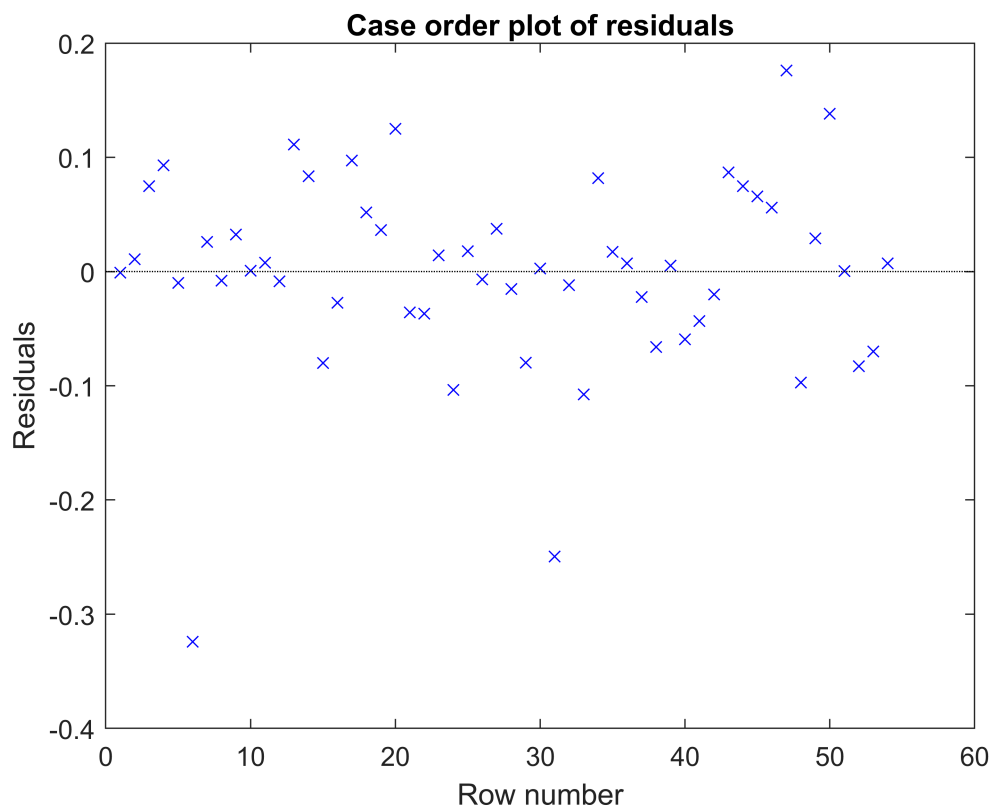
```
disp(h_adres_BOD);disp(p_adres_BOD);
```

1

0.0084

Gaussianity rejected! Residuals are not Gaussian

```
figure;  
plotResiduals(ols_md1_bod, 'caseorder');
```



Exhibits some heteroskedasticity

```
% run fgls  
[coeff_BOD,se_BOD,EstCoeffCov_BOD] = fgls(dlogX_new,dlogy_new,'innovMdl','HC0','display','final');
```

OLS Estimates:

	Coeff	SE
Const	0.6433	0.4334
x1	-0.2964	0.2146

x2		-0.4297	0.4274
----	--	---------	--------

FGLS Estimates:

		Coeff	SE
-----			
Const		0.5335	0.4380
x1		-0.2589	0.2118
x2		-0.3190	0.3938