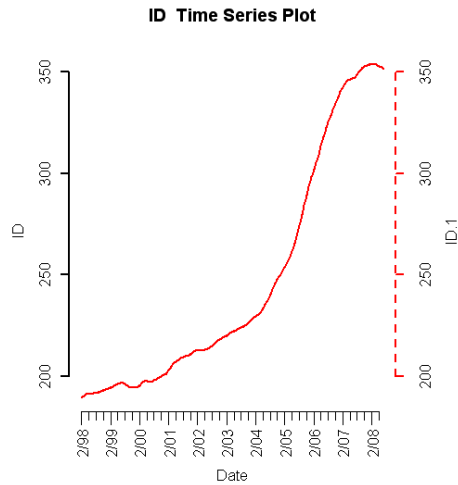


I. PHASE I AND PHASE II RESULTS

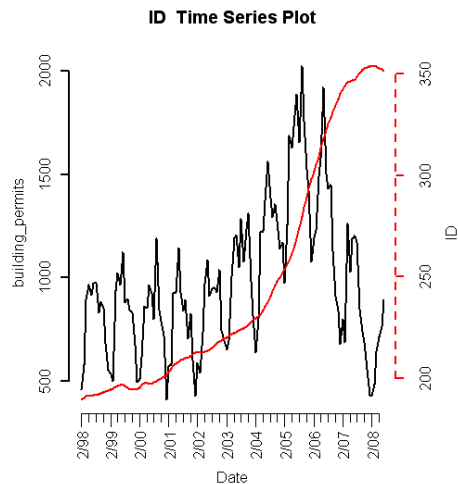
IDAHO (ID)

A. Plots of Predictors vs. HPI (non-simple return)

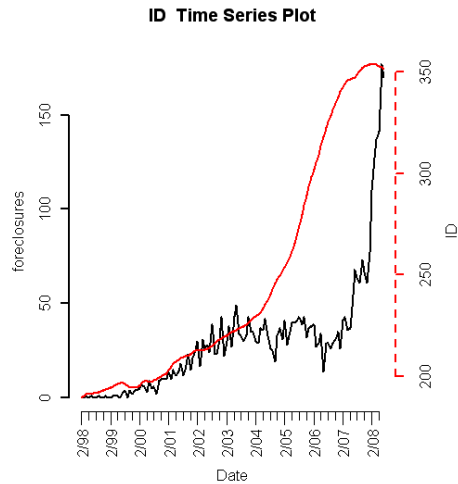
Original HPI Series



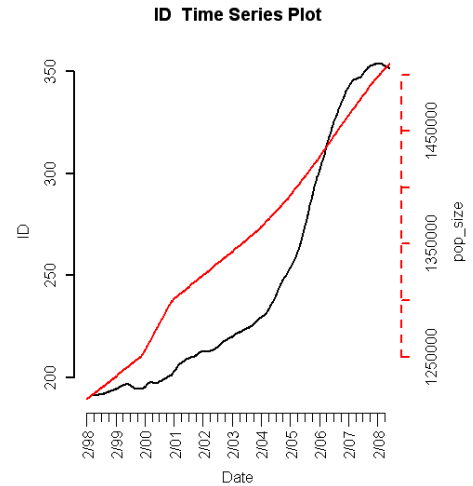
Building Permits



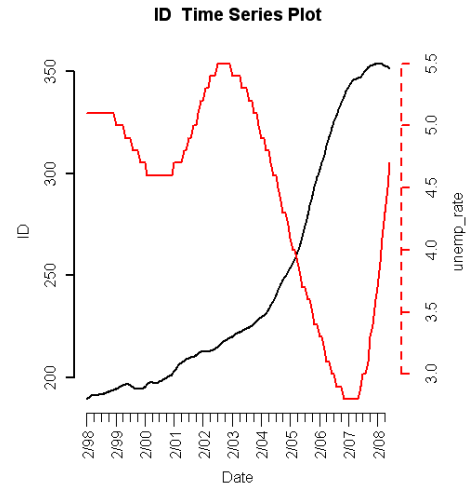
Foreclosures



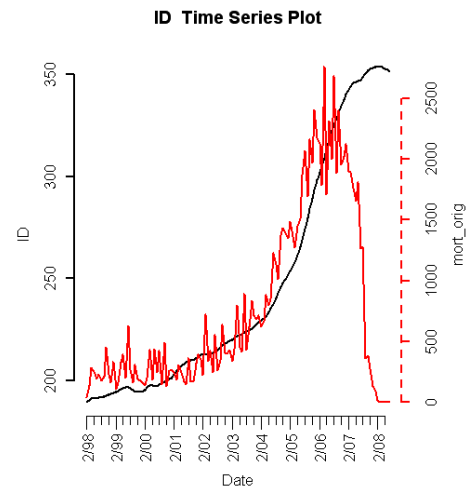
Population Size

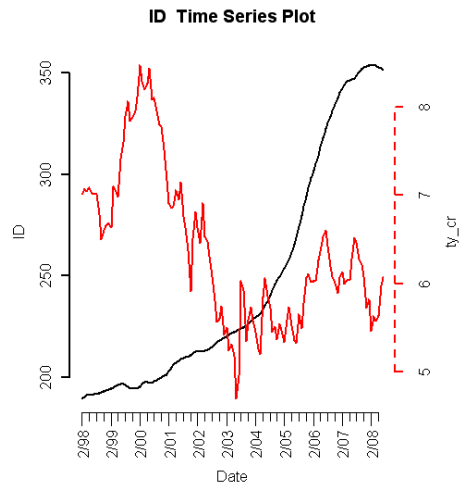
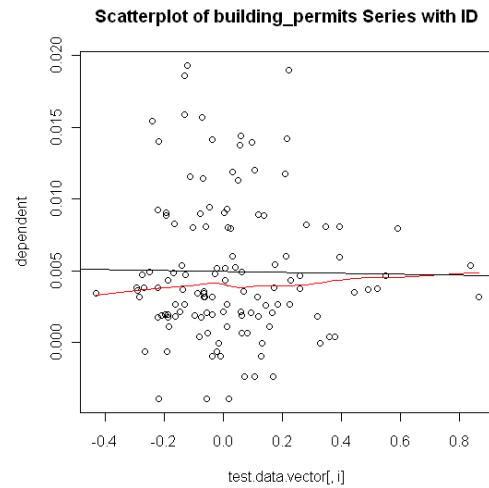
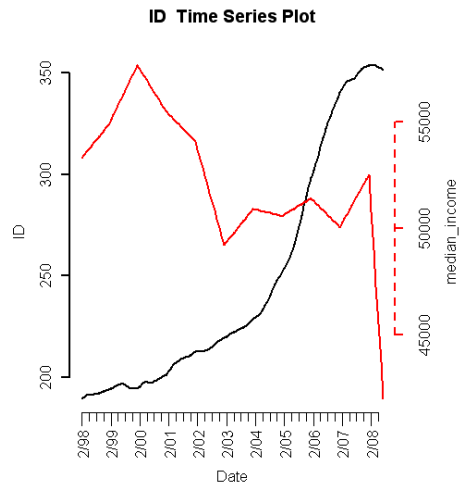
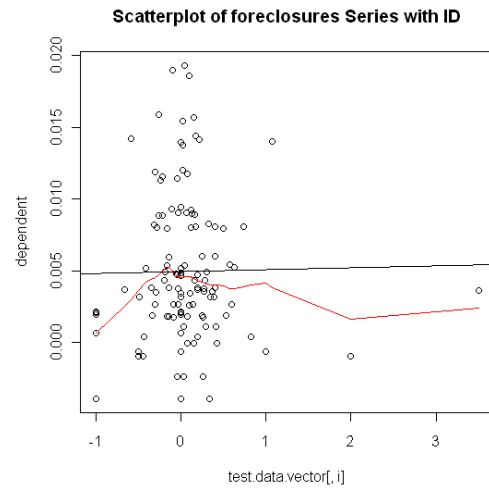
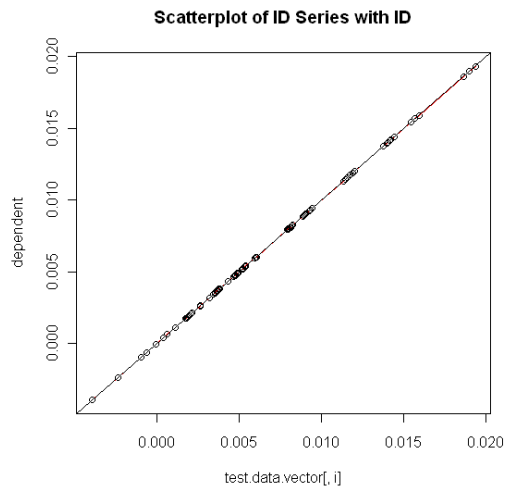
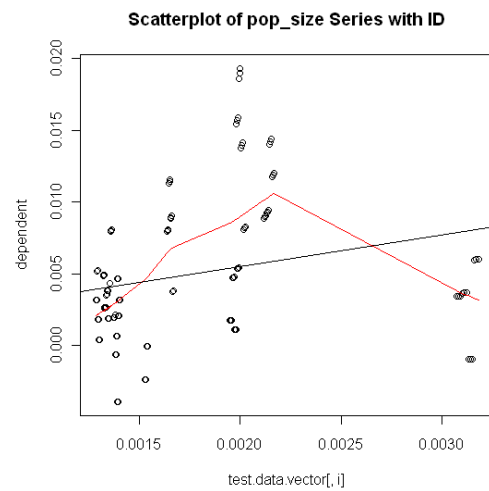


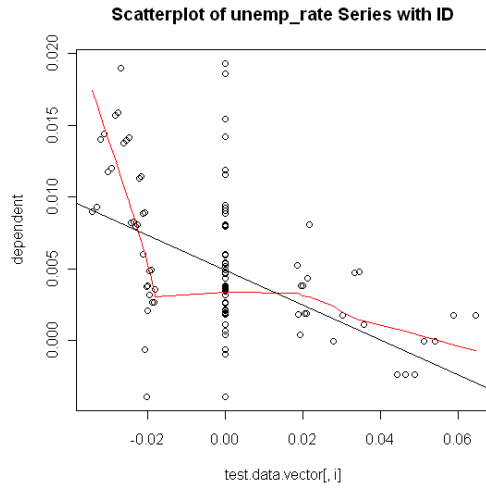
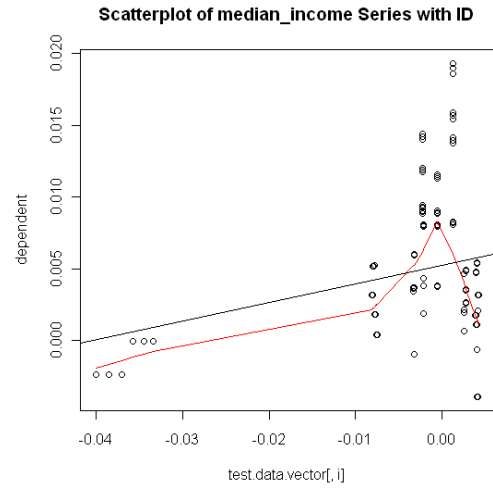
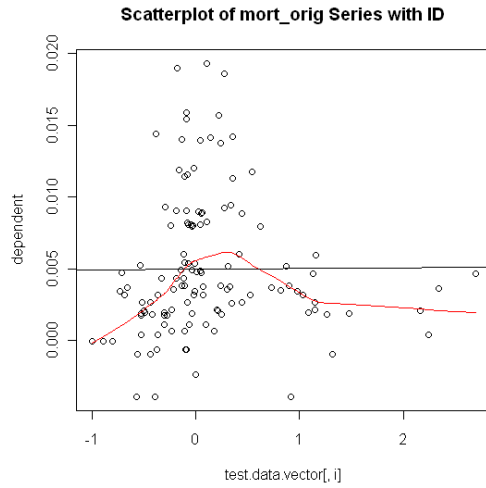
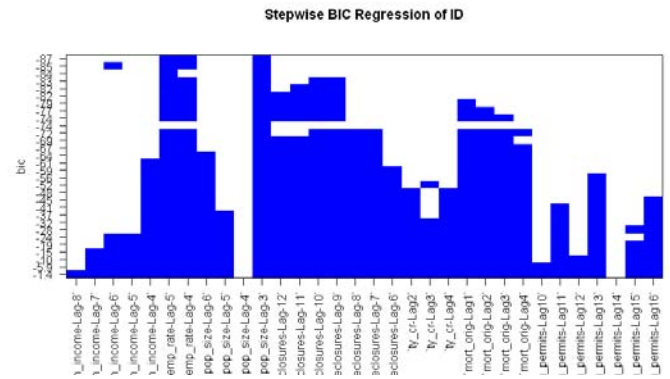
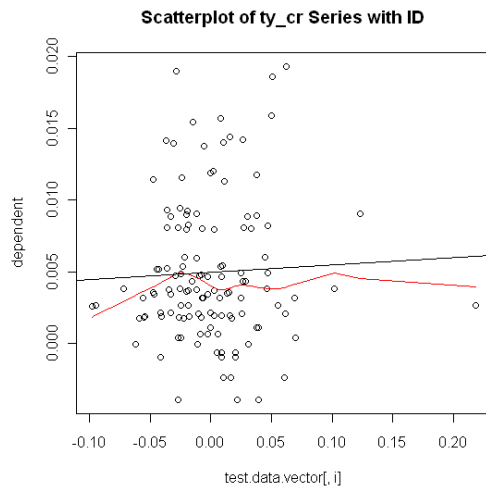
Unemployment Rate



Mortgage Originations



30-Year Commitment RateHPI vs. Building PermitsMedian IncomeHPI vs. ForeclosuresB. Scatterplots (simple return)HPI vs. HPIHPI vs. Population Size

HPI vs. Unemployment RateHPI vs. Median IncomeHPI vs. Mortgage OriginationsC. Exhaustive Stepwise RegressionHPI vs. 30-Year Commitment Rate

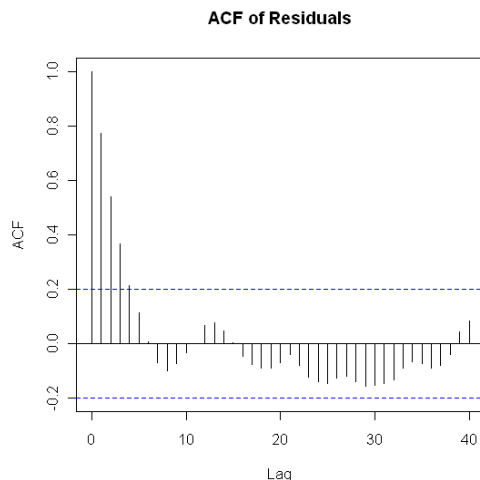
D. Final RLM Model

Coefficients:

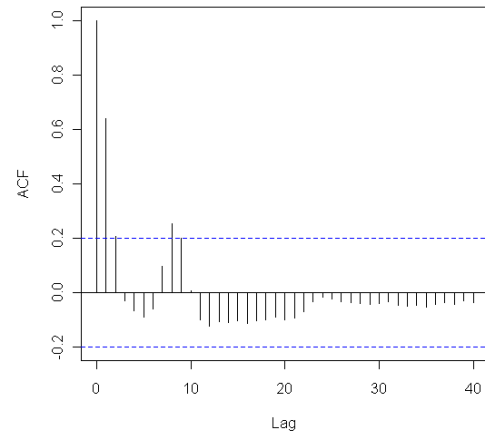
	Value	Std. Error	t-value
median_income-Lag-8	0.031	0.128	0.240
median_income-Lag-7	-0.034	0.165	-0.209
median_income-Lag-6	0.165	0.320	0.516
median_income-Lag-5	-0.108	0.417	-0.258
median_income-Lag-4	-0.326	0.299	-1.093
unemp_rate-Lag-5	-0.051	0.033	-1.534
unemp_rate-Lag-4	-0.050	0.035	-1.422
pop_size-Lag-6	3.880	2.263	1.715
pop_size-Lag-5	-1.655	3.452	-0.479
pop_size-Lag-4	-0.712	3.614	-0.197
pop_size-Lag-3	3.450	2.278	1.514
foreclosures-Lag-12	-0.002	0.002	-1.173
foreclosures-Lag-11	-0.002	0.001	-1.928
foreclosures-Lag-10	-0.005	0.001	-3.567 *
foreclosures-Lag-9	-0.005	0.001	-3.803 *
foreclosures-Lag-8	-0.003	0.001	-2.362 *
foreclosures-Lag-7	-0.003	0.001	-2.004 *
foreclosures-Lag-6	-0.002	0.001	-1.343
ty_cr-Lag2	-0.007	0.012	-0.549
ty_cr-Lag3	-0.001	0.011	-0.119
ty_cr-Lag4	-0.008	0.012	-0.609
mort_orig-Lag1	-0.003	0.001	-2.650 *
mort_orig-Lag2	-0.004	0.001	-3.009 *
mort_orig-Lag3	-0.004	0.001	-3.128 *
mort_orig-Lag4	-0.003	0.001	-2.467 *
building_permits-Lag10	0.000	0.002	-0.037
building_permits-Lag11	-0.002	0.003	-0.885
building_permits-Lag12	0.000	0.003	-0.064
building_permits-Lag13	-0.001	0.002	-0.344
building_permits-Lag14	0.002	0.002	0.937
building_permits-Lag15	0.000	0.002	0.017
building_permits-Lag16	0.001	0.002	0.546

1. Phase I

ACF of Residuals

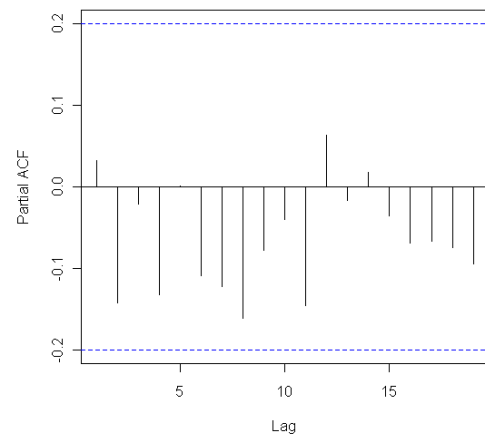


ACF of Sq-Residuals



PACF and ADF Test

PACF of Differenced Residuals



```
> adf.test(y$residuals, k = 3)
Augmented Dickey-Fuller Test
data: y$residuals
Dickey-Fuller = -3.6903, Lag order = 3, p-value = 0.02902
```

2. Phase II

```
ARIMA(1,0,0) with zero mean
Call: arima(x = resid.data, order = c(1, 0, 0), include.mean = FALSE)
```

Coefficients:

```
ar1
0.7702
s.e. 0.0629
sigma^2 estimated as 6.112e-06: log likelihood = 444.17
AIC = -884.33 AICc = -884.2 BIC = -879.18
```

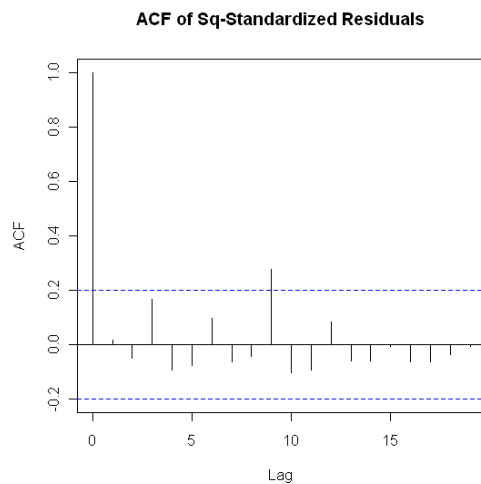
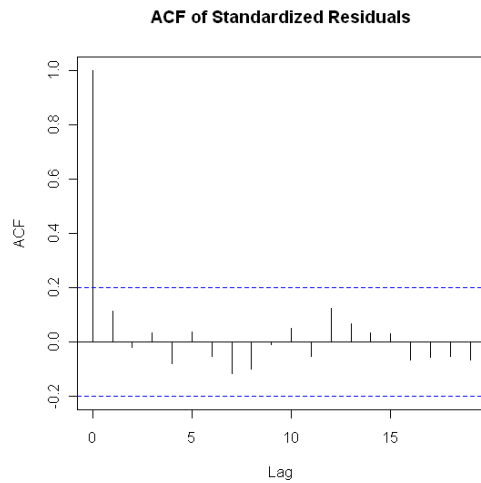
Box-Ljung Test

```
> Box.Ljung.test(arima.fit$resid, lag = 12, adj.DF = 12-params)
Q(12) P-value df
7.384673 0.7671 11

> Box.Ljung.test(arima.fit$resid^2, lag = 12, adj.DF = 12)
Q(12) P-value df
25.26165 0.0136 12
```

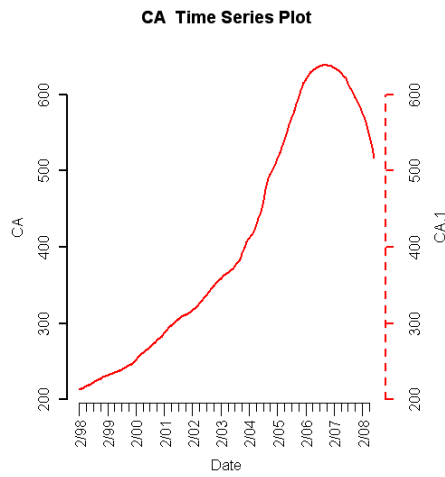
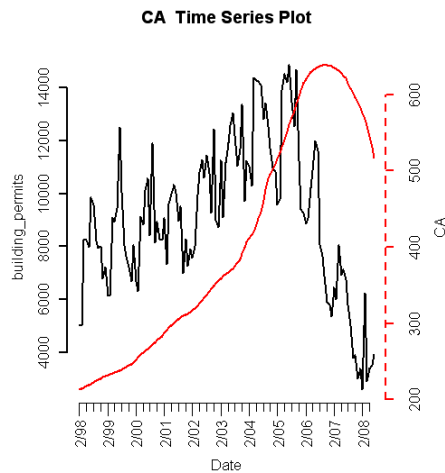
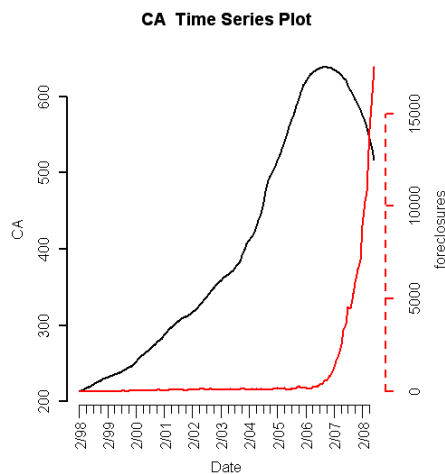
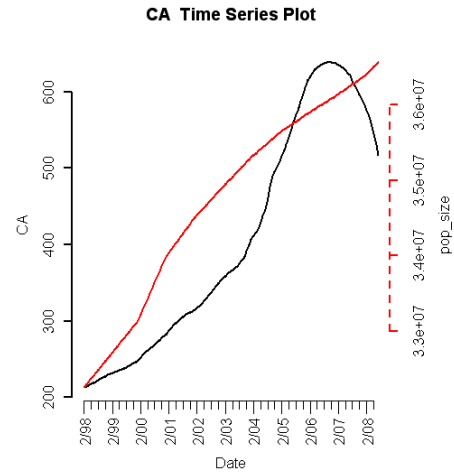
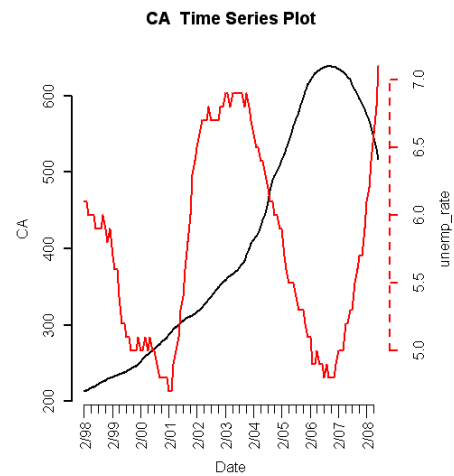
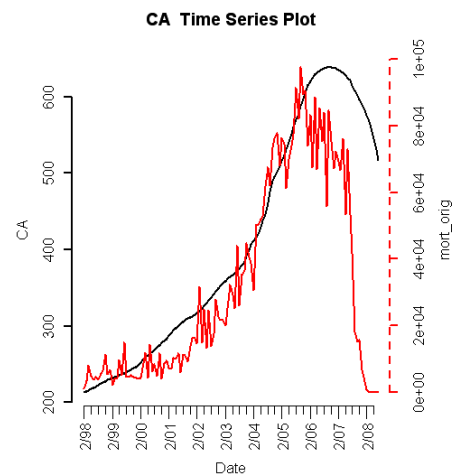
	Estimate	Std. Error	t value	Pr(> t)
ar1	0.7550	0.0716	10.55	<2e-16 ***
omega	1.644E-06	0.0000	1.03	0.304
alpha1	0.0649	0.0652	1.00	0.320
beta1	0.6636	0.2942	2.26	0.024 *

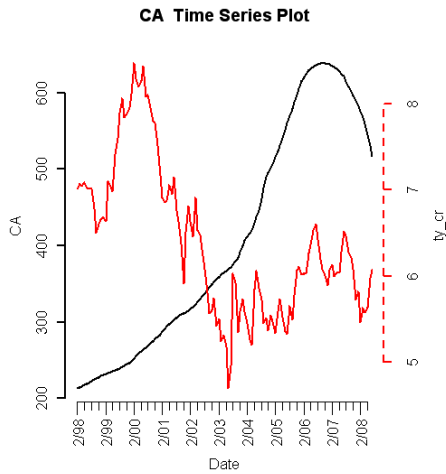
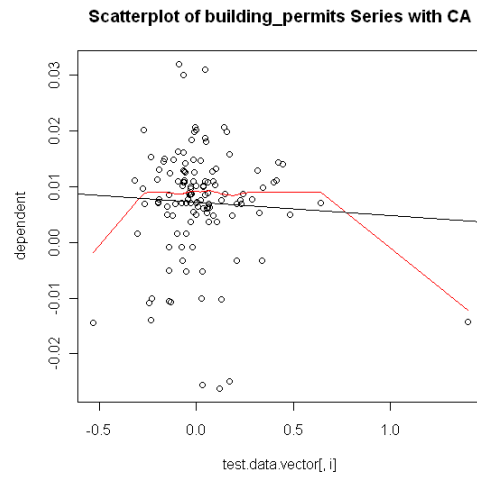
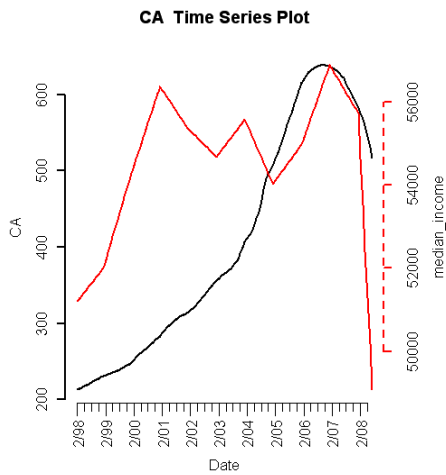
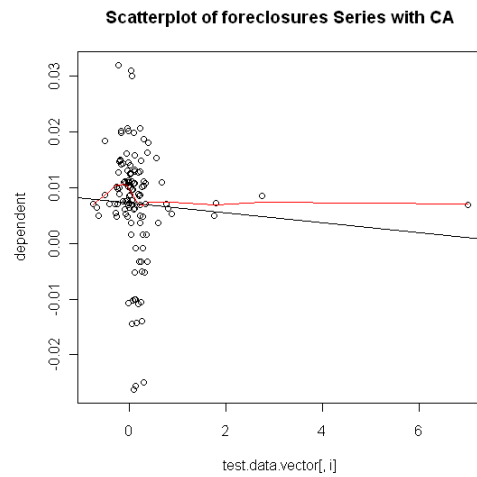
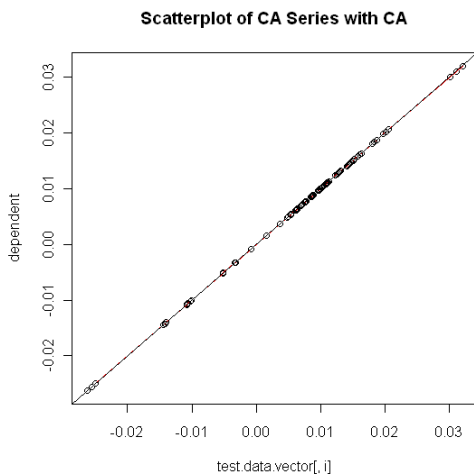
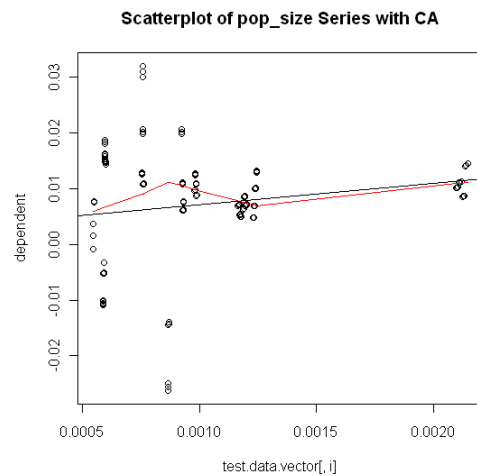
ACF of Residuals



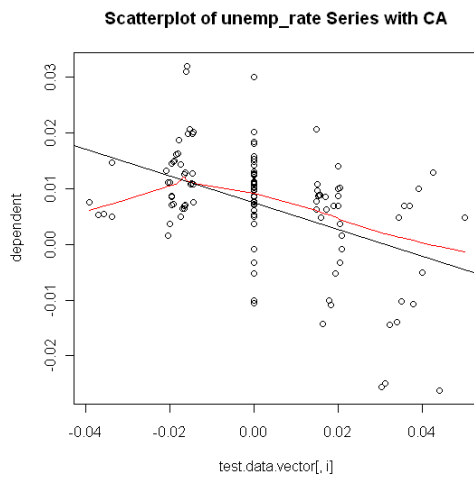
Box-Ljung Test

```
> Box.Ljung.test(sresi, lag = 12, adj.DF = 12-meanParams)
Q(12) P-value df
7.3162 0.7729 11
> Box.Ljung.test(sresi^2, lag = 12, adj.DF = 12-volParams )
Q(12) P-value df
17.41181 0.0657 10
```

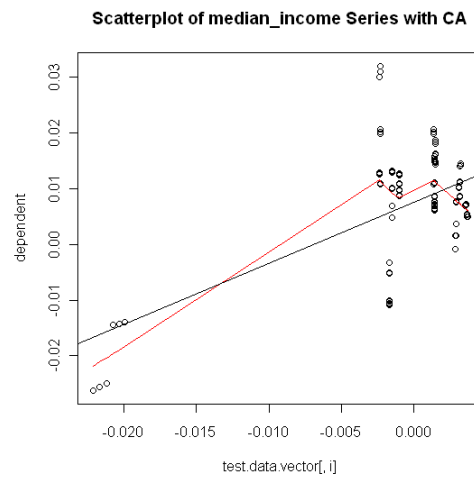
CALIFORNIA (CA)*A. Plots of Predictors vs. HPI (non-simple return)*Original HPI SeriesBuilding PermitsForeclosuresPopulation SizeUnemployment RateMortgage Originations

30-Year Commitment RateHPI vs. Building PermitsMedian IncomeHPI vs. ForeclosuresB. Scatterplots (simple return)HPI vs. HPIHPI vs. Population Size

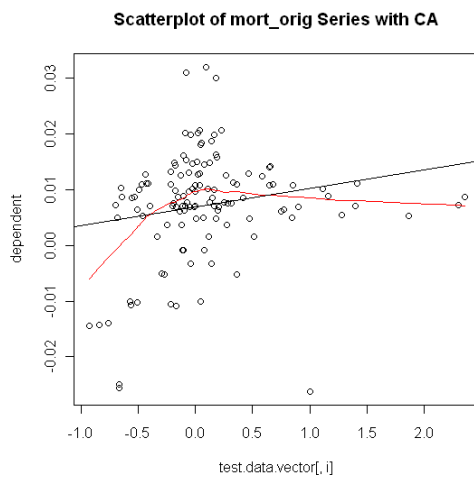
HPI vs. Unemployment Rate



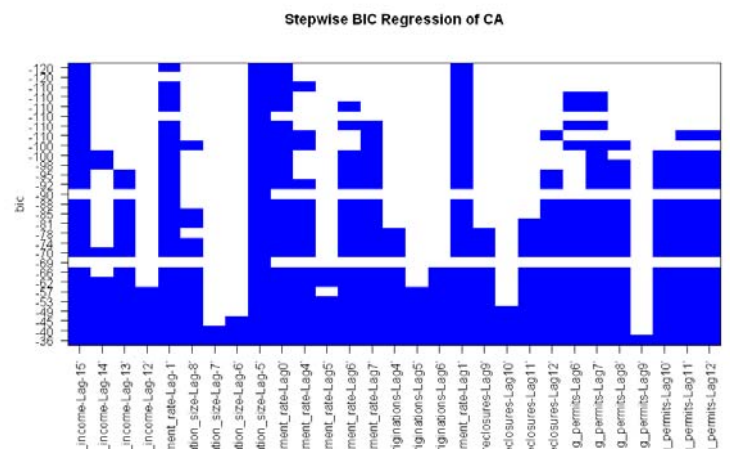
HPI vs. Median Income



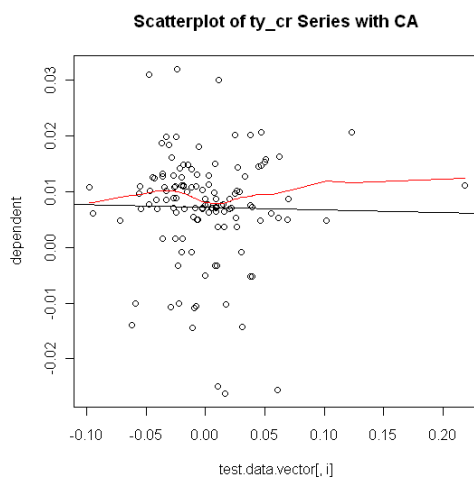
HPI vs. Mortgage Originations



C. Exhaustive Stepwise Regression

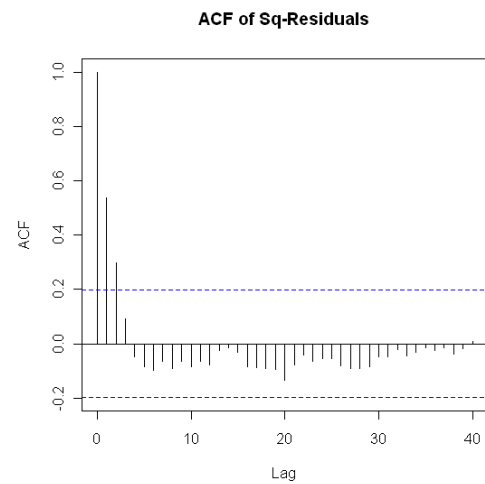
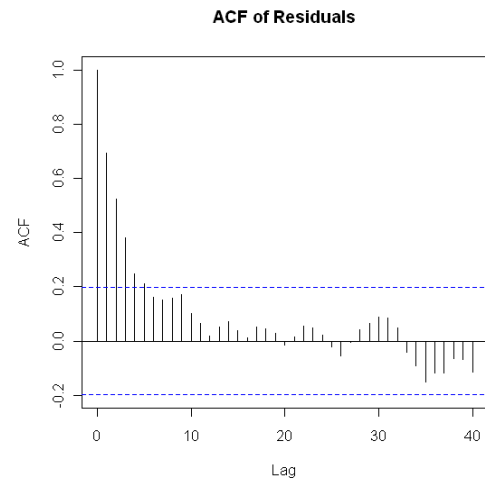
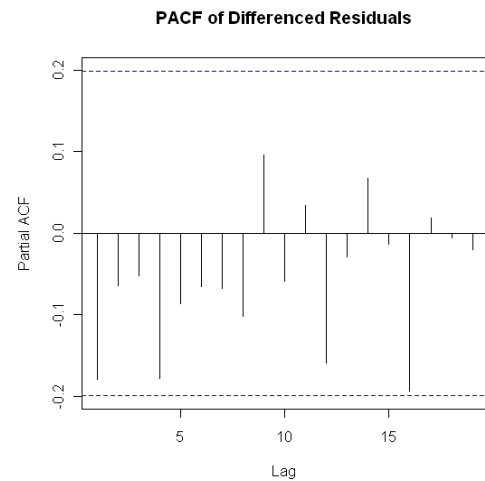


HPI vs. 30-Year Commitment Rate



E. Final RLM Model

Coefficients	Value	Std. Error	t-value
median_income-Lag-15	0.4799	0.3456	1.3888
median_income-Lag-14	-0.0537	0.4965	-0.1083
median_income-Lag-13	0.105	0.4816	0.2181
median_income-Lag-12	-0.1179	0.3923	-0.3005
ty_cr-Lag4	0.0271	0.0174	1.5552
ty_cr-Lag5	0.003	0.0167	0.1812
ty_cr-Lag6	-0.0142	0.0168	-0.8429
ty_cr-Lag7	-0.0177	0.0164	-1.0781
mort_orig-Lag4	-0.0037	0.0018	-2.0486
mort_orig-Lag5	-0.007	0.0022	-3.2181
mort_orig-Lag6	-0.004	0.0018	-2.2392
unemp_rate-Lag1	-0.1261	0.0514	-2.4554
unemp_rate-Lag2	-0.0759	0.0502	-1.5111
unemp_rate-Lag3	-0.0311	0.0504	-0.6156
unemp_rate-Lag4	-0.0354	0.0482	-0.7356
unemp_rate-Lag5	-0.0304	0.0478	-0.6358
unemp_rate-Lag6	0.0301	0.0493	0.61
pop_size-Lag2	6.0132	5.0598	1.1884
pop_size-Lag3	-4.2653	7.2838	-0.5856
pop_size-Lag4	1.6781	7.358	0.2281
pop_size-Lag5	3.9044	7.558	0.5166
pop_size-Lag6	1.5299	7.5535	0.2025
pop_size-Lag7	-6.2109	7.5069	-0.8274
pop_size-Lag8	9.5897	5.2106	1.8404
foreclosures-Lag9	-0.0018	0.001	-1.8421
foreclosures-Lag10	-0.0022	0.0009	-2.3617
foreclosures-Lag11	-0.0025	0.0009	-2.7381
foreclosures-Lag12	-0.0024	0.0009	-2.5308
building_permits-Lag6	-0.0018	0.005	-0.3537
building_permits-Lag7	0.0009	0.0053	0.1741
building_permits-Lag8	0.005	0.0052	0.952
building_permits-Lag9	0.007	0.0051	1.3639
building_permits-Lag10	0.0005	0.0051	0.0993
building_permits-Lag11	-0.0038	0.0049	-0.7675
building_permits-Lag12	-0.0059	0.0051	-1.1608

*1. Phase I*ACF of ResidualsPACF and ADF Test

```
> adf.test(y$residuals, k = 2)
Augmented Dickey-Fuller Test
data: y$residuals
Dickey-Fuller = -3.5496, Lag order = 2, p-value = 0.04141
```

Box-Ljung Test

```
> Box.test(y$residuals, lag = 12, type = c("Ljung-Box"))
Box-Ljung test
X-squared = 115.341, df = 12, p-value < 2.2e-16
```

2. Phase II

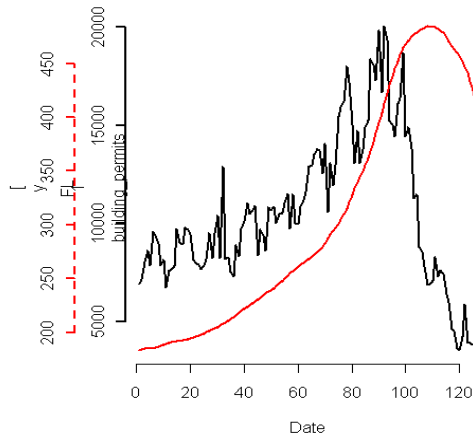
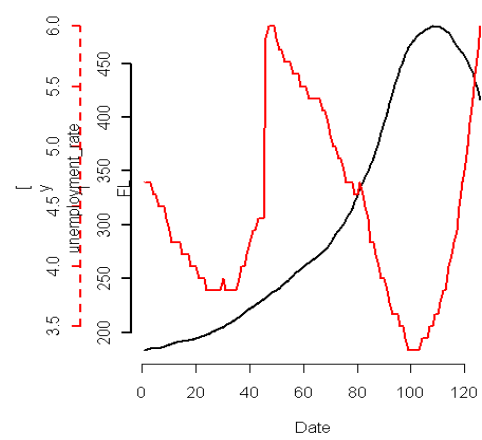
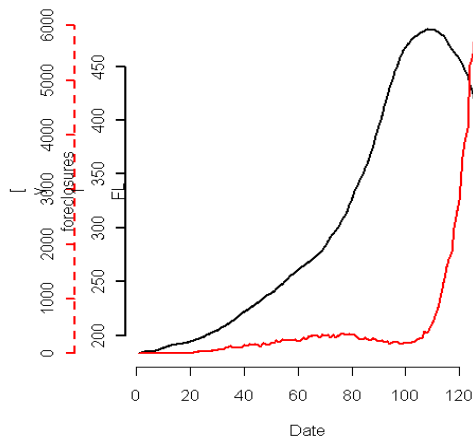
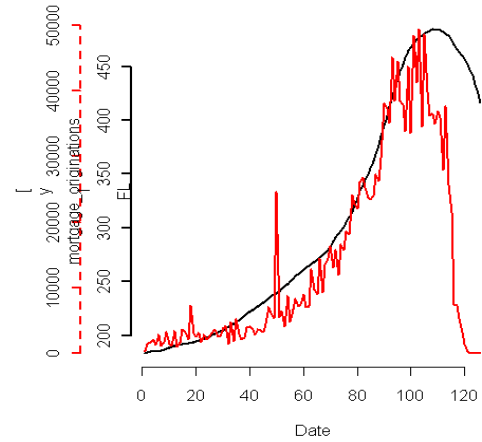
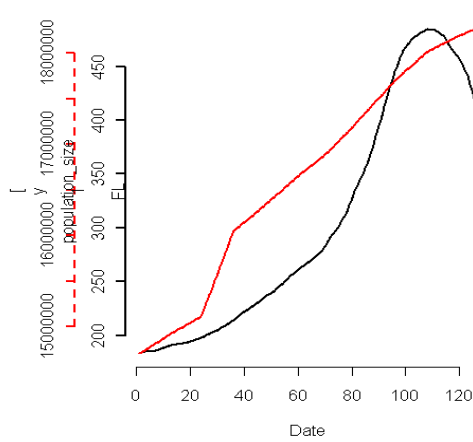
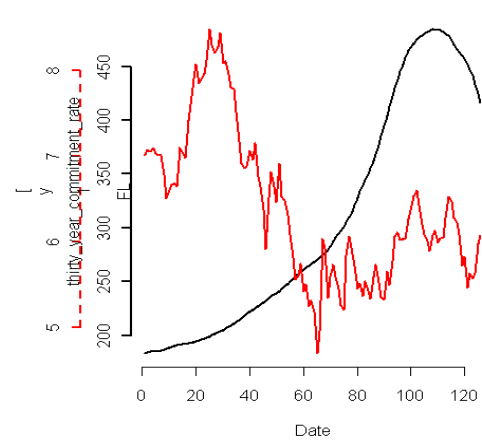
```
ARIMA(1,0,0) with zero mean
Call: auto.arima(x = resid.data, stationary = TRUE)

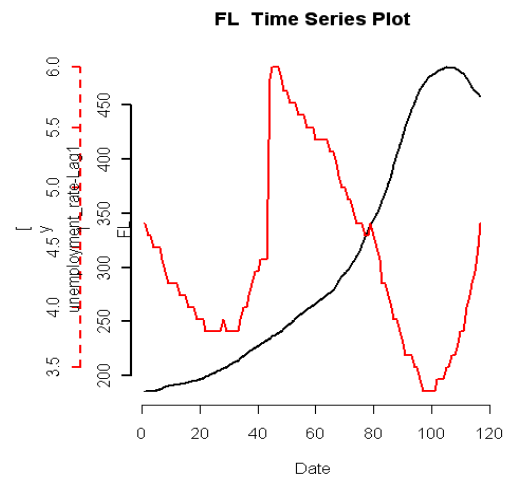
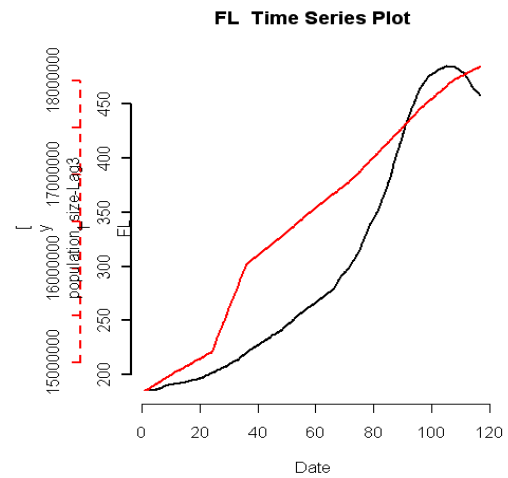
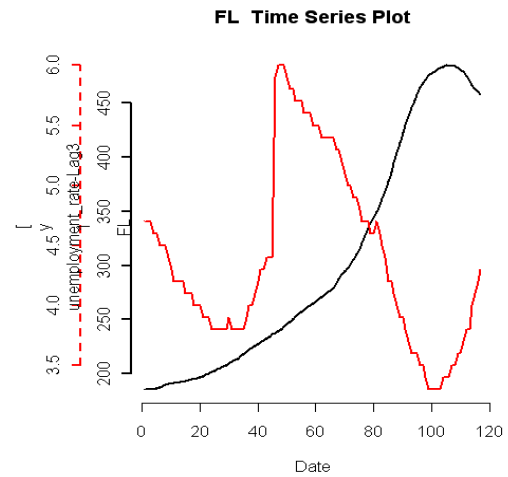
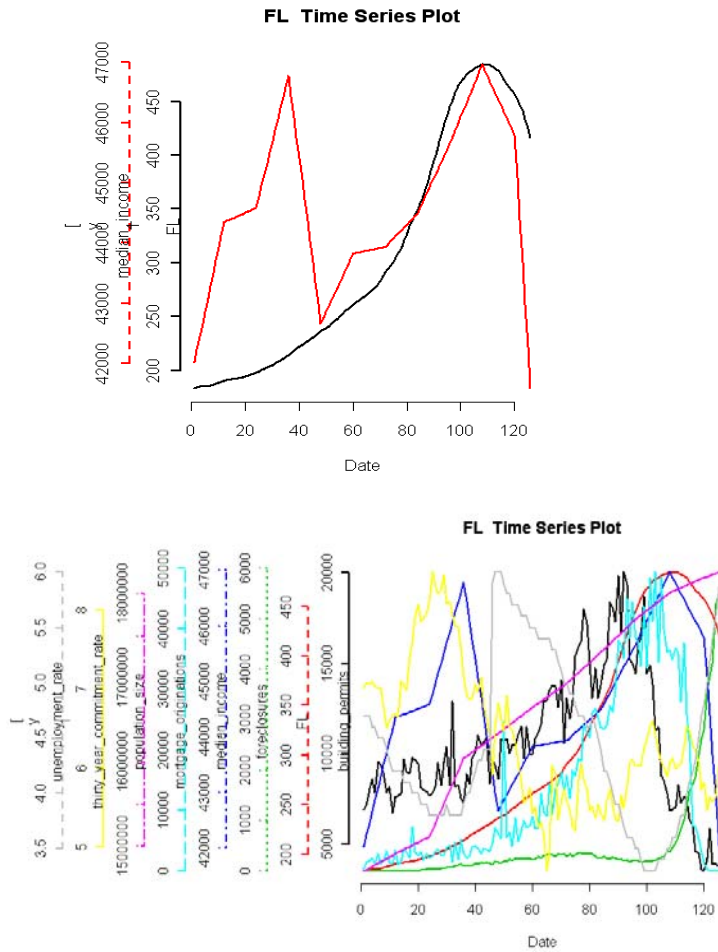
Coefficients:
    ar1
  0.7238
s.e. 0.0710
sigma^2 estimated as 1.539e-05: log likelihood = 403.57
AIC = -803.14  AICc = -803.02  BIC = -797.97
```

Box-Ljung Test

```
> Box.Ljung.test(arima.fit$resid, lag = 12, adj.DF = 12-params)
Box-Ljung test
X-squared = 6.2959, df = 11, p-value = 0.8529

> Box.Ljung.test(arima.fit$resid^2, lag = 12, adj.DF = 12)
Box-Ljung test
X-squared = 17.9114, df = 12, p-value = 0.1184
```

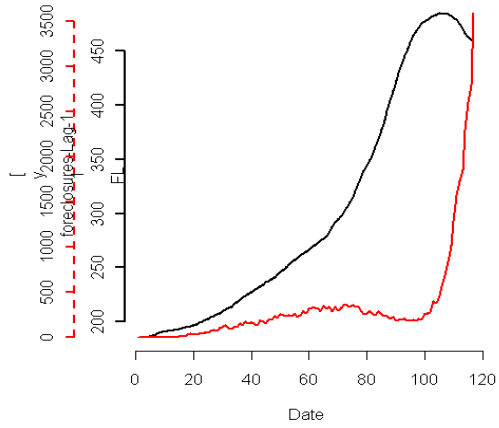
FLORIDA (CA)**FL Time Series Plot****FL Time Series Plot****FL Time Series Plot****FL Time Series Plot****FL Time Series Plot****FL Time Series Plot**

Plots of lagged predictor variables and FL HPI:

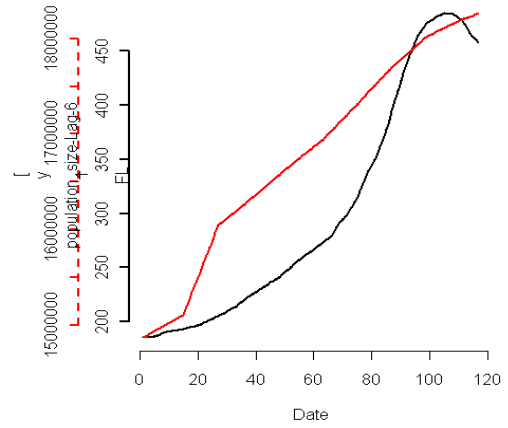
Based on these time series of the data, we picked the following lags to test:

Predictor Variable	Lag(s)
Median income	0
30yr commitment rate	0
Mortgage originations	0
Unemployment rate	-4, -3, 1, 3
Population size	-6, 3
Foreclosures	-3, -1
Building permits	0

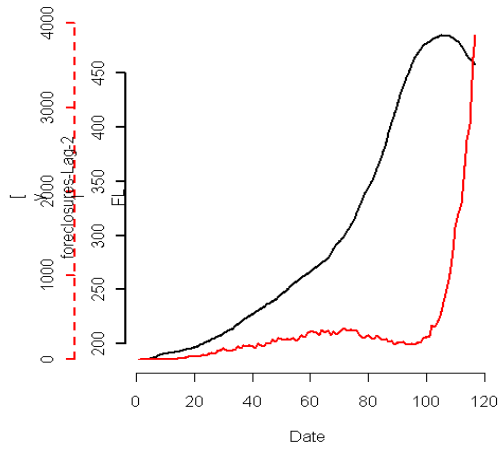
FL Time Series Plot



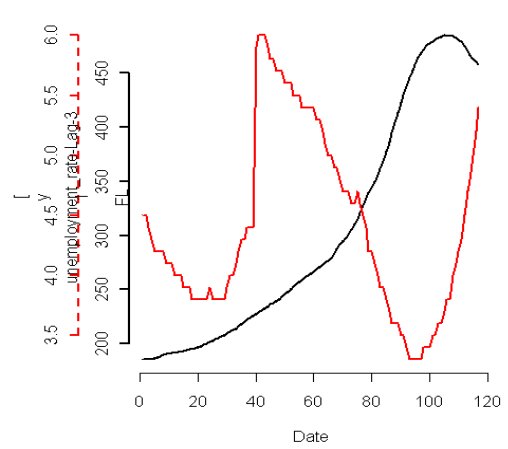
FL Time Series Plot



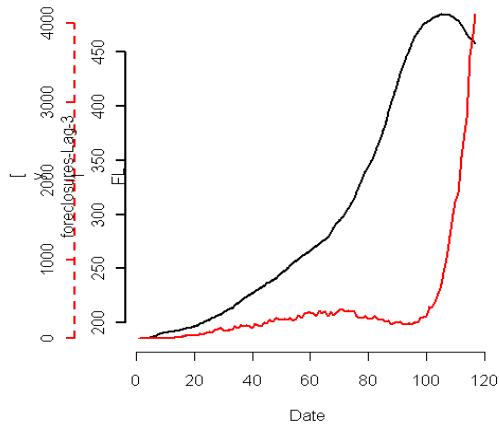
FL Time Series Plot



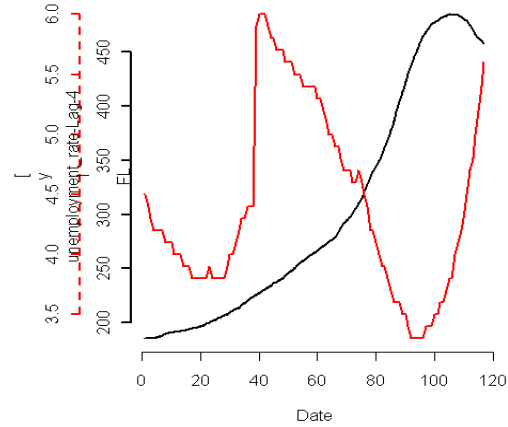
FL Time Series Plot



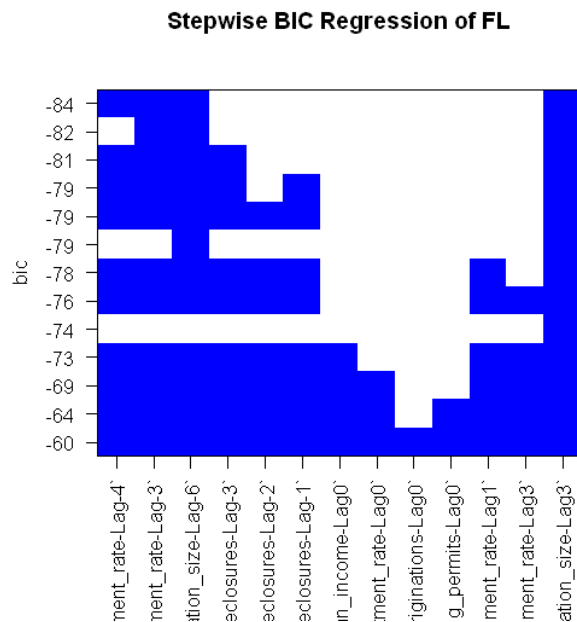
FL Time Series Plot



FL Time Series Plot



Stepwise BIC values for Regression of Lagged Predictor Variables with Florida



Fitted Robust Linear Model with all lagged variables

```
Call: rlm(x = shifted.multivar.test.data.vector[, 1:
(n - 1)], y = shifted.multivar.test.data.vector[,
n])
```

Residuals:

	Min	1Q	Median	3Q	Max
	-0.0159690	-0.0020288	0.0008338	0.0045595	0.0295564

Coefficients:

	Value	Std. Error	t value
unemployment_rate-Lag-4	-0.0712	0.0162	-4.3983
unemployment_rate-Lag-3	-0.0836	0.0165	-5.0723
population_size-Lag-6	1.3061	0.4229	3.0884
foreclosures-Lag-3	-0.0026	0.0016	-1.6460
foreclosures-Lag-2	-0.0026	0.0015	-1.6699
foreclosures-Lag-1	-0.0026	0.0015	-1.7549
median_income-Lag0	-0.2167	0.2059	-1.0522
thirty_year_commitment_rate-Lag0	0.0009	0.0024	0.3597
mortgage_originations-Lag0	-0.0002	0.0010	-0.2033
building_permits-Lag0	0.0017	0.0045	0.3742
unemployment_rate-Lag1	-0.0285	0.0176	-1.6213
unemployment_rate-Lag3	-0.0237	0.0160	-1.4781
population_size-Lag3	2.9363	0.4187	7.0134

Residual standard error: 0.005344 on 103 degrees of freedom

Fitted Robust Linear Model with unlagged variables

```
Call: rlm(x = test.data.vector[, 1:(length(test.data.vector)
- 1)],
```

```
y = test.data.vector[, length(test.data.vector)])
```

Residuals:

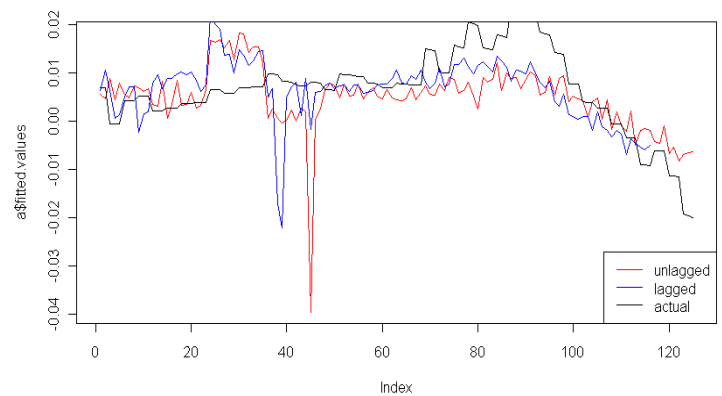
	Min	1Q	Median	3Q	Max
	-0.013701	-0.002252	0.001575	0.006133	0.047483

Coefficients:

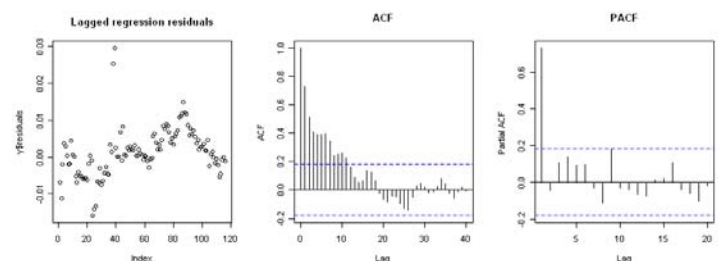
	Value	Std. Error	t value
median_income	0.2237	0.1666	1.3423
thirty_year_commitment_rate	0.0002	0.0030	0.0686
mortgage_originations	0.0002	0.0011	0.2014
unemployment_rate	-0.1259	0.0219	-5.7532
population_size	2.9903	0.3543	8.4397
foreclosures	-0.0009	0.0017	-0.5132
building_permits	0.0040	0.0051	0.7731

Residual standard error: 0.007476 on 118 degrees of freedom

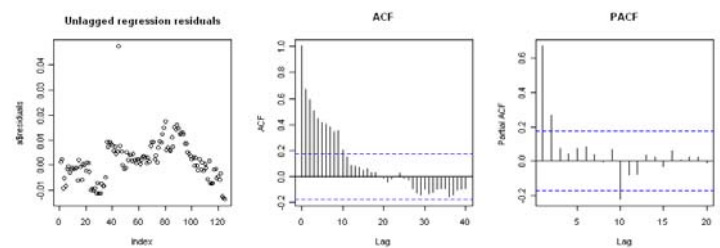
Model vs. Fitted FL



Lagged Regression Residuals Analysis:



Unlagged Regression Residuals Analysis:



The PACF shows a lag of 2 to use in the ADF test for stationarity.

Augmented Dickey-Fuller Test

```
data: y$residuals
Dickey-Fuller = -3.5887, Lag order = 2,
p-value = 0.03735
alternative hypothesis: stationary
```

Residuals are stationary so phase 1 complete. Continue to Phase 2.

First, check Box-Ljung.

Box-Ljung test

```
data: y$residuals
X-squared = 220.422, df = 12, p-value < 2.2e-16
```

Box-Ljung fails. Implies residuals are autocorrelated. So we will now model the residuals as time series.

EACF shows that we should try to fit ARMA(1,0).

Fit ARMA:

Series: y\$residuals

ARIMA(1,0,0) with non-zero mean

Call: arima(x = y\$residuals, order = c(1, 0, 0))

Coefficients:

	arl	intercept
	0.7356	0.0011
s.e.	0.0624	0.0015

sigma^2 estimated as 1.921e-05: log likelihood = 464.91

AIC = -923.82 AICc = -923.61 BIC = -915.56

Check adequacy of ARMA

```
> Box.Ljung.test(arima.fit$residuals, lag = 12,
adj.DF = 12-params)
```

Box-Ljung test

data: data

X-squared = 12.6459, df = 11, p-value = 0.3171

The mean equation passes.

```
> Box.Ljung.test(arima.fit$residuals^2, lag = 12,
adj.DF = 12)
```

Box-Ljung test

data: data

X-squared = 24.6408, df = 12, p-value = 0.01662

Residuals^2 fails. GARCH modeling needed.

FIT GARCH: Garch estimates

mu	arl	omega
alpha1	beta1	
4.111261e-04	3.241833e-01	3.089253e-06
1.000000e+00	2.568742e-01	

Check adequacy of GARCH model (using standardized residuals)

```
> sresi=garch.fit@residuals/garch.fit@sigma.t
> Box.Ljung.test(sresi, lag = 12, adj.DF = 12-
meanParams)
```

Box-Ljung test

data: data

X-squared = 10.337, df = 11, p-value = 0.5004

```
> Box.Ljung.test(sresi^2, lag = 12, adj.DF = 12-
volParams )
```

Box-Ljung test

data: data

X-squared = 5.7986, df = 10, p-value = 0.8319

Both Box-Ljung tests for the standardized residuals and squared standardized residuals of the ARMA(1,0) + GARCH(1,1) model pass.

The final fitted models are as follows.

Phase 1 Regression: Drift Modeling

$y_t = \sum \beta_i x_{t,i} + \text{res}_t$,
where res_t are the residuals

Variable	β_i
unemployment_rate-Lag-4	-0.0712
unemployment_rate-Lag-3	-0.0836
population_size-Lag-6	1.3061
foreclosures-Lag-3	-0.0026
foreclosures-Lag-2	-0.0026
foreclosures-Lag-1	-0.0026
median_income-Lag0	-0.2167
thirty_year_commitment_rate-Lag0	0.0009
mortgage_originations-Lag0	-0.0002
building_permits-Lag0	0.0017
unemployment_rate-Lag1	-0.0285
unemployment_rate-Lag3	-0.0237
population_size-Lag3	2.9363

Phase 2 Time Series: Volatility Modeling

$$(1 - \phi_1 B) \text{res}_t = \phi_0 + a_t$$

$$\sigma_t^2 = \alpha a_{t-1}^2 + \beta \sigma_{t-1}^2$$

$$a_t = \varepsilon_t \sigma_t$$

$$\varepsilon_t \sim N(0, 1)$$

Coefficient	Estimate
ϕ_0	0.0011
ϕ_1	0.7356
α	1.00
β	0.26