

# Leading Indicators of Heating Coal Pricing in Turkey: A Coal Pricing Model (2003-2009)

Mehmet Mithat Uner<sup>1</sup>, Nezir Kose<sup>2</sup>, Soner Gokten<sup>3</sup>

<sup>1</sup>Department of Business Administration, Gazi University, Ankara, Turkey; <sup>2</sup>Department of Econometrics, Gazi University, Ankara, Turkey; <sup>3</sup>(Corresponding Author) Department of Accounting and Finance, Gazi University, Ankara, Turkey.  
E-mail: soner.gokten@gazi.edu.tr

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## ABSTRACT

*In this study, a coal pricing model for Turkey is developed employing Granger causality and cointegration analysis by using monthly data between January 2003 and April 2009. Empirical results based on Granger causality tests indicate that foreign coal futures prices and domestic consumer price index for energy sector can be used as the leading indicators for domestic coal prices for Turkey. An error correction model for Turkish coal pricing is specified by taking into account the results of Granger causality. The forecast of the coal prices based on error correction model is giving very successful results. It is observed that the coal prices and forecasted coal prices values are almost moving together or very close to each other.*

**Keywords:** Pricing, Granger Causality, Heating Coal

## 1. Introduction

Leading indicators of Turkish coal prices can be considered as foreign energy prices, domestic inflation, cost factors, and economic growth of Turkey. The list of leading indicators [1] used in this study is given in **Table 1**.

Our analysis is based on monthly data over the period January 2003-April 2009. Turkish coal prices (TL per tone) were taken from the Turkish Statistics Institution. The code for this data is 0454001 according to Classification of Individual Consumption by Purpose (COICOP). We use US West Texas Intermediate (WTI) for foreign crude oil spot prices and Cushing, Oklahoma Crude Oil Future Contract 1 for foreign crude oil futures prices. Henry Hub Natural Gas Spot Prices and Natural Gas Futures Contract 1 are considered for spot and futures prices of foreign natural gas. Spot and futures prices for foreign coal are Coal Commodity Spot Prices Central Appalachia (CAPP) 12,500 Btu, 1.2 SO<sub>2</sub>, and CAPP Coal Futures, respectively.

The source for the spot prices is the Energy Information Administration (EIA), while futures prices were taken from the New York Mercantile Exchange (NY-MEX). Foreign energy prices are converted to Turkish Lira from US Dollar to consider the effect of exchange rate on Turkish coal prices.

**Table 1. The indicators for Turkish coal prices.**

<i>Foreign Energy Prices</i>	
•	Crude Oil Spot Prices
•	Crude Oil Futures Prices
•	Natural Gas Spot Prices
•	Natural Gas Futures Prices
•	Coal Spot Prices
•	Coal Futures Prices
<i>Domestic Inflation</i>	
•	Producer Price Index
•	Producer Price Index for Mining Sector
•	Consumer Price Index
•	Consumer Price Index for Energy Sector
<i>Cost Factors</i>	
•	Unit Labour Cost
•	Other Inputs Cost
<i>Real Sector</i>	
•	Industrial Production

Four indicators are used for domestic inflation;

namely, producer price index, producer price index for mining and stone quarrying industry, consumer price index, and consumer price index for electricity, gas and other fuels. Base year of the price indexes is 2003. Data were taken from the electronic data delivery system of Central Bank of the Republic of Turkey (CBRT). Index of Wages per Production Hour Worked in Manufacturing of Cook and Refined Petroleum, whose base year is 1997, is used as a proxy for the unit labour costs. Monthly data are obtained through an interpolation by linear method as the wage index is available only at a quarterly frequency in the CBRT Electronic Data Delivery System. The electricity price is used as a proxy variable to capture the effect of prices of other cost inputs on coal price levels. The data were taken from the Turkish Statistics Institution (TL per KWh, The COICOP code is 0451001).

Industrial Production Index, whose base year is 2005, is used as a proxy to measure real income at a monthly frequency. Data source is the Electronic Data Delivery System of the CBRT. To account for the seasonal effects, the data are seasonally adjusted by using the Tramo/Seats method. All data in this study are in logarithmic form.

## 2. Methods

### 2.1. Granger Causality Test for the Leading Indicators

Granger [2], [3], [4], proposed a time-series data based approach in order to determine causality relationships among variables. According to Granger [2], the definition of causality is based entirely on the predictability of some series, say. If some other series contains information in past terms that helps in the prediction of and if this information is contained in no other series used in the predictor, then is said to Granger cause. Granger causality has been used in the context of rational expectations, definition of strong exogeneity, and econometric modelling strategy. A better term for Granger causality is precedence [5]. Therefore, this test can be used for determining leading indicators of any variable. We may also use the results of Granger causality for evaluating forecasting performance since it is concerned with one-ahead forecast accuracy.

Park and Philips [6], Sims, Stock and Watson [7] and Toda and Philips [8] have shown that the standard asymptotic theory is not applicable to hypothesis testing in level VAR model if the variables are integrated or cointegrated. Therefore, the usual Wald test statistics for Granger non-causality based on level VAR not only has nonstandard asymptotic distribution but depends on nuisance parameters in general if variables are nonstationary.

In this study, we examine Granger causal relationships between Turkish coal prices and other variables using Toda-Yamamoto [9] approach to determine the leading indicators for Turkish coal price level. Toda-Yamamoto procedure considers a lag augmented or modified Wald (M-Wald) test which has conventional asymptotic chi-square ( $\chi^2$ ) distribution when a VAR (p + dmax) is estimated where dmax is the maximal order of integration suspected to occur in the system. In other words, this lag augmentation procedure provides standard asymptotic although the time series have integration/cointegration properties, and therefore, can be applied without a priori information about the presence (absence) and location of unit roots. The results of Granger causality test based on M-Wald statistic are given in **Table 2**.

Foreign coal future prices and consumer price index for energy sector are Granger causes of Turkish coal prices at 5% level. There is also Granger causality from domestic electricity prices and industrial production to Turkish coal prices at 10% level. These results imply that foreign coal futures prices and consumer price index for energy sector can be used as the leading indicators for domestic coal prices for Turkey. Therefore, these variables will be used the next steps of our analysis.

### 2.2. Unit Root Test with one Structural Break

**Figure 1** shows a time plot of the data set over the sam-

**Table 2. The Results of Pairwise Granger Causality Tests**

Null hypothesis: X does not Granger cause of Turkish coal prices	Lag	M-Wald Statistic	p-value
Foreign oil spot prices	3	2.97	0.3961
Foreign oil future prices	2	1.28	0.5273
Foreign natural gas spot prices	3	2.30	0.5122
Foreign natural gas future prices	3	0.68	0.8774
Foreign coal spot prices	11	15.94	0.1432
Foreign coal future prices	6	1454	0.0241
Consumer price index	2	3.59	0.2741
Consumer price index for electricity, gas and other fuels	12	34.55	0.0006
Producer price index	2	0.48	0.7872
Producer price index for mining and stone quarrying	2	1.53	0.4655
Unit labour cost	6	4.37	0.6256
Electricity Prices	10	17.58	0.0624
Industrial Production	8	14.11	0.0790

While maximum lag is 12, optimal lag length is determined by using two types of information criteria (Schwarz and Akaike). If the two selection criteria determine different lag order, Modified-Wald test, developed by Toda and Yamamoto [9] is performed to eliminate lags from a general to a more specific model.

Maximum order of integration in the system is equal to 1.

ple period that displays to have an upward trend in levels with a non-deterministic structure. Domestic coal prices for Turkey and consumer price index for energy sector both exhibit similar shapes while foreign coal futures prices represent different shape from other series. Foreign coal futures prices have been very volatile, changing their trajectories and behaviour with respect to the economic situation. Moreover, all variables include structural breaks in 2008. The visual analysis therefore tentatively suggests that all the variables are not stationary. The next step is to verify this conclusion using unit root procedure.

To test for a unit root in time series data, the Augmented Dickey-Fuller [10] procedure is commonly used in empirical studies. Perron [11] was the first to point out that power to reject the unit root null declines if the data contains a structural break that is ignored. Perron [11] incorporated an exogenous structural break into an ADF test. More recently, Zivot Andrews [12] and Perron [13] proposed unit root tests that allow for a structural break to be determined endogenously from the data. In order to check whether a unit root is present in the data or not, we used Perron [13] test because of structural breaks in the series. Perron [11] defined three types of models for one-time break in the trend function (Model A, Model B,

Model C). Model A allows for a one-time change in the intercept of the trend function. It is known as the “Crash Model”. Model B allows only a change in the slope of the trend function at the time of the break. Model C includes a one time change in both level and trend. As suggested in **Figure 1**, we use Model A for all series due to there is only a change in the intercept of the trend function. The results for the Perron [13] unit root test are reported in **Table 3**. The unit root null hypothesis cannot be rejected for all variables at 5% significance level. These results indicate that the order of integration for all series is equal to one.

**2.3. Cointegration Analysis**

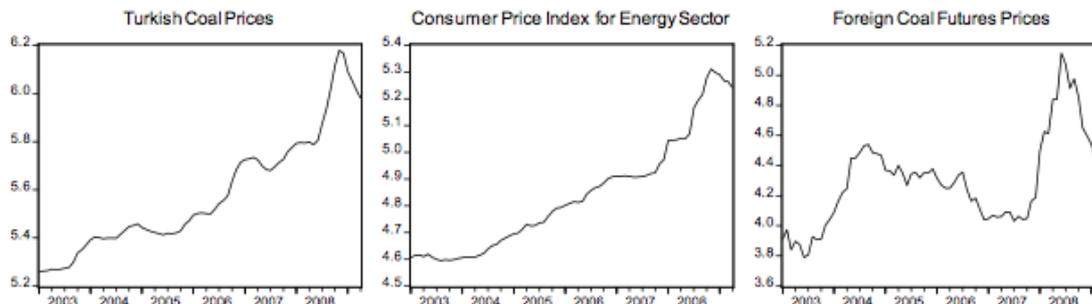
The long-run relationship is investigated using the cointegration analysis of Johansen and Juselius [14]. It is well known that the results of cointegration tests using this technique depend on the deterministic components included in the VAR and on the chosen lag length. The appropriate lag length is selected by using two types of information criteria (Schwarz and Akaike). The VAR order in logarithm level is estimated as 7 by Schwarz information criteria while it is estimated as 12 by Akaike information criteria. Due to the two selection criteria were determined different lag order as 7 and 12, respectively, Modified-Wald test [9], [15] was performed to eliminate lags, and the appropriate lag length is estimated as 7. We carried out both the trace and maximum eigenvalue type cointegration tests of Johansen and Juselius [14]. Owing to the trace statistic and the maximum eigenvalue statistic may yield conflicting results. Number of cointegrating equations by the deterministic components in model is summarized in **Table 4**. Both the trace and maximum eigenvalue (except test type 1) test statistics indicate that there are two cointegrating equations for all deterministic trend assumption.

The long-run analysis results are based on deterministic trend assumptions that both the time series and the cointegrating equation have linear trends which correspond to assumption 3 since all series have an upward

**Table 3. The Results of Unit Root Test**

Series	Estimated Break Point: $T_B$	Lag	t-Statistic	Methods	Critical Value*
Turkish Coal Prices	2004:M03	4	-4.5746	Min $t_\alpha$	-5.09
	2004:M03	4	-4.5746	Max $t_{\alpha,\hat{\rho}}$	-5.04
Consumer Price Index for Energy Sector	2008:M09	1	-4.8556	Min $t_\alpha$	-5.09
	2008:M09	1	-4.8556	Max $t_{\alpha,\hat{\rho}}$	-5.04
Foreign Coal Futures Prices	2006:M06	5	-4.7004	Min $t_\alpha$	-5.09
	2006:M06	5	-4.7004	Max $t_{\alpha,\hat{\rho}}$	-5.04

\*At 5% for T = 80. The appropriate lag length is determined through general to specific testing which is suggested by Perron [11].



**Figure 1. Graphic representation of the variables in logarithmic levels.**

trend in levels. When the error correction term is normalized with respect to coal prices for Turkey, the results of VEC model are given **Table 5**. The VEC model is very robust as all the diagnostic tests are insignificant, indicating that the residuals are normally distributed, homoskedastic and not serially correlated.

The coefficient of error correction term in coal prices equation is estimated as  $-0.1350$  and it is statistically significant. It shows that the adjustment speed toward long-run equilibrium will be  $0.1350$  and the full adjustment of deviation takes about 7 months. The coefficients of consumer price index for energy sector and foreign coal future prices in cointegrating equation are estimated as  $1.54$  and  $0.44$  respectively and they are statistically significant at the 5% level.

The positive sign of coefficients is consistent with economic theory. The results indicate that an increase in consumer price index for energy sector and foreign coal future price of 1 percent will increase coal price for Turkey by approximately 1.54 percent and 0.44 percent re-

**Table 4. Number of cointegrating relations by the deterministic components in model (5% level).**

Data Trend	None	None	Linear	Linear	Quadratic
Test Type	No Intercept	Intercept	Intercept	Intercept	Intercept
	No Trend	No Trend	No Trend	Trend	Trend
Trace	2	2	2	2	2
Max-Eigen	1	2	2	2	2

**Table 5. VECM results (cointegrating equation).**

Turkish Coal Prices	1.0000
Consumer Price Index for Energy Sector	-1.5351 (19.60)
Foreign Coal Futures Prices	-0.4369 (4.38)
Intercept term	3.7425
Error Correction	-0.1350 (-6.97)

Residual diagnostic test	Test	Test Statistic	p-value
Normality test	Lutkepohl (Jarque-Bera)	3.36	0.1862
Serial correlation	Breusch-Godfrey (LM)	5.92 for lag 1	0.7483
		7.39 for lag 2	0.5968
		9.82 for lag 3	0.3657
Heteroskedasticity	White	5.53 for lag 4	0.7861
		203.60	0.8759

Values in parentheses are t-statistics.

spectively.

### 3. Results

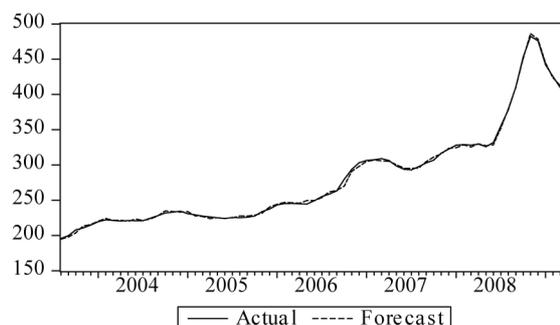
Actual values and forecasts for Turkish coal prices obtained from VEC model are graphed in **Figure 2**. The forecasting level of domestic coal prices for Turkey is remarkably close to actual level as is seen by examination of **Figure 2**.

To assess the forecast performance of VEC model, we compare their Root Mean Square Error (RMSE) and Theil Inequality Coefficient with a naive model, where the naive model is a first order autoregressive model for logarithmic domestic coal prices. RMSE and Theil Inequality Coefficient are computed for the forecasted value of domestic coal prices which enables us to make comparisons across different models.

**Table 6** shows that the VEC model has a much lower RMSE and Theil Inequality Coefficient than the naive model. These results imply that we have a significant power in predicting Turkish coal prices using consumer price index for energy sector and foreign coal futures prices as the leading indicators in Turkey.

### 4. Conclusions

This study tries to determine the important factors that affect the heating coal market of Turkey in the frame of pricing decisions. A coal pricing model for Turkey was developed by using Granger causality and cointegration analysis. Empirical results suggest that settlement prices of coal futures and domestic consumer price index for energy sector can be used as the leading indicators in order to determine and forecast the domestic heating coal



**Figure 2. Actual and forecast values of Turkish coal prices.**

**Table 6. Forecast error statistics.**

	RMSE	Theil Inequality Coefficient
VEC Model	2.4506	0.0042
Naive Model	10.0176	0.0176

prices for Turkey. An error correction model for Turkish coal pricing is specified by taking into account the results of Granger causality. The forecast of the coal prices based on error correction model is giving very successful results. It is observed that the coal prices and forecasted coal prices values are almost moving together or very close to each other.

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