

Statistics Department,
Fergusson College,
(Established 1885)

TIME SERIES:
OIL PRICES
IN INDIA

Project Under
CPE-UG

PROJECT BY:
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CERTIFICATE

This is to certify that the project report entitled
“TIME SERIES: OIL PRICES IN INDIA” is an original work done
by

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Shubhangi Zanan**

under my supervision and able guidance.

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We would also like to express a sense of gratitude to our teachers for their support and efforts that encouraged us to undertake this project work because of which we are able to get a practical experience of handling the data using the statistical tools which we have been studying from the last few years.

Also, a thanks to the non-teaching staff of the Statistics Department for their cooperation in making this project a success.

ABSTRACT AND KEYWORDS

ABSTRACT

This project is based on studying the changes in Oil Prices over the years. We study the trends in variations of Oil Prices over the years 1989 - 2010.

We use Time Series Analysis to process the data and observe the changes in the prices of Diesel, Petrol, Kerosene and LPG through the time period.

Using the patterns, hence, observed we have estimated and forecasted oil prices for the successive years from 2011 - 2020 for Diesel, Petrol, Kerosene and LPG each.

KEYWORDS

- Time Series
- Fitting of Trend Line- Linear, Logarithmic, Exponential, Polynomial Trend Lines.
- R- Squared Value
- Residual Sum of Squares, Residual Plots.
- Exponential Smoothing
- Double Exponential Smoothing

INTRODUCTION

CHANGES IN OIL PRICES AND HOW THEY EFFECT OUR ECONOMY

This project report is basically focused on the changes in oil prices and how they affect our economy. Oil is an important resource in every walk of our life, may it be our households or our vehicles, and they all are in some way or the other dependent on oil.

The project is focused on the oil prices in India. A major oil bed in India is situated in Digboi, Assam & Bombay High, Mumbai. Most of our oil is imported from the Arab countries and as the oil prices are affected there, the prices here in India also get affected.

Our industries especially the transportation industry is deeply influenced by the oil prices due to which transportation costs may it be the railway fare, the bus fare or even the local public transport fare is affected by the changes in oil prices. Other than this the chemical industries, import-export business is also affected.

So, in order to take a close look into the topic, the oil prices from the year 1989-2010 have been taken as the data. Analysis has been done using different statistical methods.

DATA SOURCE AND SOFTWARES USED

DATA SOURCE

The type of data used is Secondary Data.

The address of the website from which the data was obtained
is:

in.reuters.com

SOFTWARES USED

- ❖ Microsoft Excel
- ❖ Microsoft Word

TIME SERIES ANALYSIS: DEFINITIONS AND CONCEPTS

Time Series Analysis

A time series or a stochastic process (stochastic means random) is a sequence of data points, measured typically at successive points in time spaced at uniform time intervals. Time series are very frequently plotted via line charts. This involves analysis of data over time with the intention of finding some pattern in order to construct short term forecasting.

Forecasting is the use of a model to predict future values based on previously observed values. If history had been different, we would observe a different outcome, thus we can think of time series as the outcome of a random variable. The difference is that instead of dealing with individuals as units; the unit of interest here is time.

Allowing one to answer question such as what is the causal effect on a variable Y of a change in variable x

Types of Time Series

Continuous Time Series

A time series is said to be continuous when observations are made continuously in time. The term continuous is used for series of this type even when the measured variable can take only discrete set of values.

Discrete Time Series

A time series is said to be discrete when observations are taken at specific time, usually equally spaced. The term discrete is used for series of this type even when the measured variable is a continuous variable.

The Simple Moving Average

The simplest way to smooth a time series is to calculate a simple, or unweighted, moving average. The smoothed statistics is then just the mean of the last k observations:

$$s_t = \frac{1}{k} \sum_{n=0}^{k-1} x_{t-n} = \frac{x_t + x_{t-1} + x_{t-2} + \cdots + x_{t-k+1}}{k} = s_{t-1} + \frac{x_t - x_{t-k}}{k},$$

where the choice of an integer $k > 1$ is arbitrary.

A small value of k will have less of a Smoothing effect and be more responsive to recent changes in the data, while a larger k will have a greater Smoothing effect, and produce a more pronounced lag in the smoothed sequence. This means effectively extrapolating outside the existing data, and the validity of this section would therefore be questionable and not a direct representation of the data.

A major drawback with the SMA is that it lets through a significant amount of the signal shorter than the window length. Worse, it actually inverts it. This can lead to unexpected artefacts, such as peaks in the "smoothed" result appearing where there were troughs in the data. It also leads to the result being less "smooth" than expected since some of the higher frequencies are not properly removed.

The observed phenomenon may be an essentially random process, or it may be an orderly, but noisy, process. Whereas in the simple moving average the past observations are weighted equally, Exponential Smoothing assigns Exponentially decreasing weights over time.

Weighted Moving Average

A slightly more intricate method for Smoothing a raw time series $\{x_t\}$ is to calculate a weighted moving average by first choosing a set of weighting factors

$\{w_1, w_2, \dots, w_k\}$ Such that $\sum_{n=1}^k w_n = 1$

and then using these weights to calculate the smoothed statistics $\{s_t\}$:

$$s_t = \sum_{n=1}^k w_n x_{t+1-n} = w_1 x_t + w_2 x_{t-1} + \dots + w_k x_{t-k+1}.$$

In practice the weighting factors are often chosen to give more weight to the most recent terms in the time series and less weight to older data. Notice that this technique has the same disadvantage as the simple moving average technique (i.e., it cannot be used until at least k observations have been made), and that it entails a more complicated calculation at each step of the Smoothing procedure. In addition to this disadvantage, if the data from each stage of the averaging is not available for analysis, it may be difficult if not impossible to reconstruct a changing signal accurately (because older samples may be given less weight). If the number of stages missed is known however, the weighting of values in the average can be adjusted to give equal weight to all missed samples to avoid this issue.

What is Exponential Smoothing?

This is a very popular scheme to produce a smoothed Time Series. Whereas in Single Moving Averages the past observations are weighted equally, Exponential Smoothing assigns Exponentially decreasing weights as the observation get older.

In other words, recent observations are given relatively more weight in forecasting than the older observations.

In the case of moving averages, the weights assigned to the observations are the same and are equal to $1/N$. In Exponential Smoothing, however, there are one or more Smoothing parameters to be determined (or estimated) and these choices determine the weights assigned to the observations.

Simple Exponential Smoothing

The Equation used in Simple Exponential Smoothing is

Current Forecast = α *previous value + $(1-\alpha)$ *previous Forecast value

$$\text{i.e. } \hat{Y}_{t+1} = \alpha Y_t + (1-\alpha) \hat{Y}_t$$

Thus \hat{Y}_{t+1} is a Weighted arithmetic mean of Y_t and \hat{Y}_t . In this procedure, the recent past has assigned more weights whereas distant past has assigned discounted weights which decrease exponentially.

Where α is the *Smoothing factor*, and $0 < \alpha < 1$. In other words, the smoothed statistic s_t is a simple weighted average of the previous observation x_{t-1} and the previous smoothed statistic s_{t-1} . The term *Smoothing factor* applied to α here is something of a misnomer, as larger values of α actually reduce the level of Smoothing, and in the limiting case with $\alpha = 1$ the output series is just the same as the original series (with lag of one time unit). Simple Exponential Smoothing is easily applied, and it produces a smoothed statistic as soon as two observations are available.

Values of α close to one have less of a Smoothing effect and give greater weight to recent changes in the data, while values of α closer to zero have a greater Smoothing effect and are less responsive to recent changes. There is no formally correct procedure for choosing α . Sometimes the statistician's judgment is used to choose an appropriate factor. Alternatively, a statistical technique may be used to optimize the value of α . For example, the method of least squares might be used to determine the value of α for which the sum of the quantities $(s_{n-1} - x_{n-1})^2$ is minimized.

Why is Exponential Smoothing, Exponential?

By direct substitution of the defining equation for simple Exponential Smoothing back into itself we find that

$$\begin{aligned}s_t &= \alpha x_{t-1} + (1 - \alpha)s_{t-1} \\ &= \alpha x_{t-1} + \alpha(1 - \alpha)x_{t-2} + (1 - \alpha)^2 s_{t-2} \\ &= \alpha [x_{t-1} + (1 - \alpha)x_{t-2} + (1 - \alpha)^2 x_{t-3} + (1 - \alpha)^3 x_{t-4} + \cdots + (1 - \alpha)^{t-1} x_0] + (1 - \alpha)^t s_0.\end{aligned}$$

In other words, as time passes the smoothed statistic s_t becomes the weighted average of a greater and greater number of the past observations x_{t-n} , and the weights assigned to previous observations are in general proportional to the terms of the geometric progression $\{1, (1 - \alpha), (1 - \alpha)^2, (1 - \alpha)^3, \dots\}$. A geometric progression is the discrete version of an exponential function, so this is where the name for this Smoothing method originated.

Double Exponential Smoothing

Simple Exponential Smoothing does not do well when there is a trend in the data. In such situations, several methods were devised under the name "Double Exponential Smoothing" or "second-order Exponential Smoothing." The basic idea behind Double Exponential Smoothing is to introduce a term to take into account the possibility of a series exhibiting some form of trend. This slope component is itself updated via Exponential Smoothing.

One method sometimes referred to as "Holt-Winters Double Exponential Smoothing" works as follows:

Again, the raw data sequence of observations is represented by $\{x_t\}$, beginning at time $t = 0$. We use $\{s_t\}$ to represent the smoothed value for time t , and $\{b_t\}$ is our best estimate of the trend at time t . The output of the algorithm is now written as F_{t+m} , an estimate of the value of x at time $t+m$, $m > 0$ based on the raw data up to time t . Double Exponential Smoothing is given by the formulas

$$s_1 = x_1$$

$$b_1 = x_1 - x_0$$

And for $t > 1$ by

$$s_t = \alpha x_t + (1 - \alpha)(s_{t-1} + b_{t-1})$$

$$b_t = \beta(s_t - s_{t-1}) + (1 - \beta)b_{t-1}$$

where α is the data Smoothing factor, $0 < \alpha < 1$, and β is the trend Smoothing factor, $0 < \beta < 1$.

To forecast beyond x_t

$$F_{t+m} = s_t + mb_t$$

Where a_t is level or intercept & b_t is slope of the trend. Single Exponential Smoothing smoothes just the level. We need to smooth out slope (b_t) also since it undergoes frequent ups & downs, F_{t+1} = Current forecast, s_t = previous year's smoothed value of intercept & b_t = previous year's smoothed value of slope of the trend.

We have two smoothing constants (α_1, α_2). The constant α_1 is used for smoothing the level while α_2 is used for smoothing the slope.

Setting the initial value b_0 is a matter of preference. An option other than the one listed above is $(x_n - x_0)/n$ for some $n > 1$.

Note that F_0 is undefined (there is no estimation for time 0), and according to the definition $F_1 = s_0 + b_0$, which is well defined, thus further values can be evaluated.

A second method, referred to as either Brown's Linear Exponential Smoothing (LES) or Brown's Double Exponential Smoothing works as follows.

$$s'_0 = x_0$$

$$s''_0 = x_0$$

$$s'_t = \alpha x_t + (1 - \alpha)s'_{t-1}$$

$$s''_t = \alpha s'_t + (1 - \alpha)s''_{t-1}$$

$$F_{t+m} = a_t + mb_t,$$

where a_t , the estimated level at time t and b_t , the estimated trend at time t are:

$$a_t = 2s'_t - s''_t$$

$$b_t = \frac{\alpha}{1 - \alpha}(s'_t - s''_t).$$

EXAMPLES OF TIME SERIES

RICE PRODUCTION

DATA:

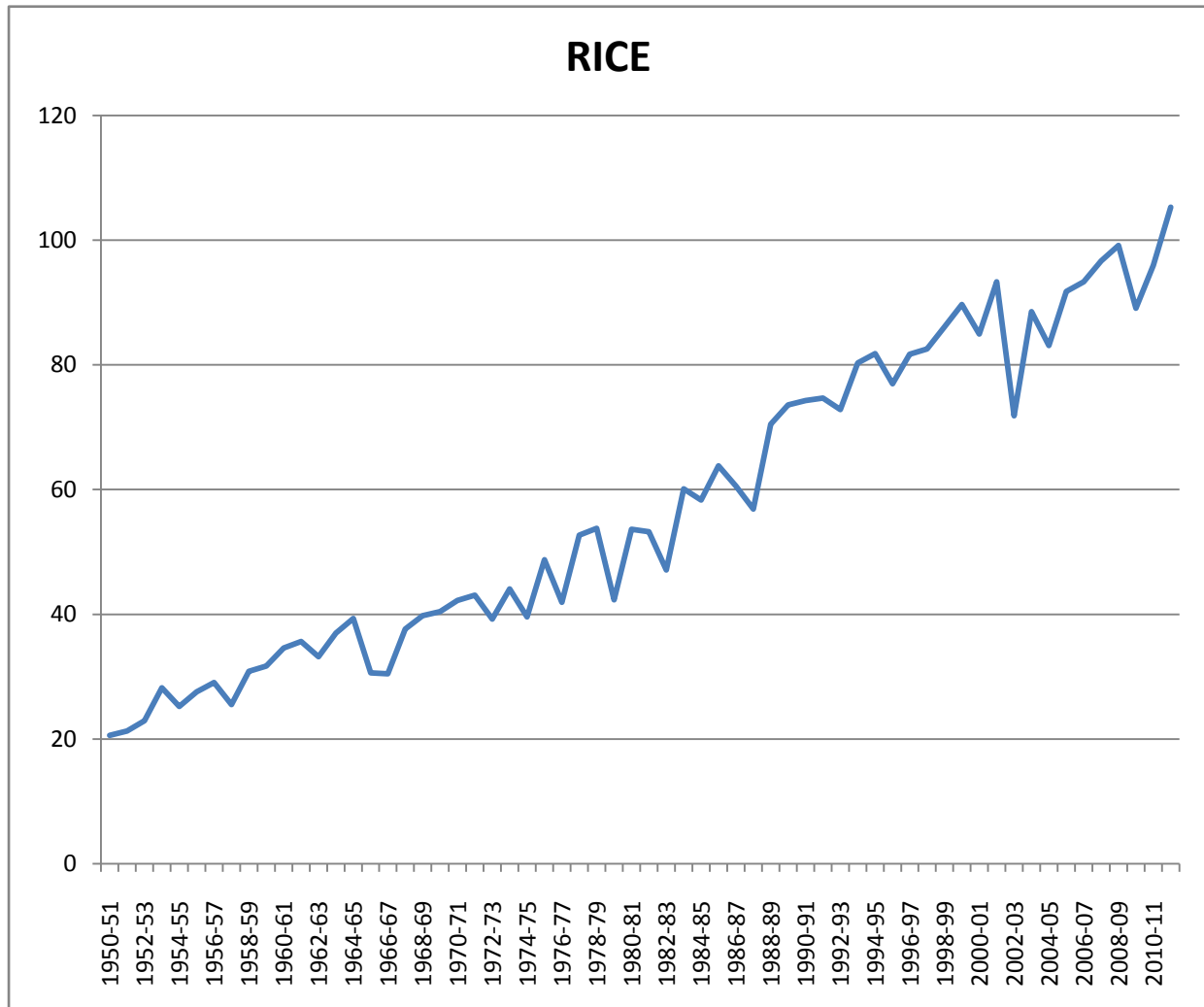
The following table gives the average annual Production of Rice in India in million tonnes:

YEAR	RICE PROD	YEAR	RICE PROD	YEAR	RICE PROD	YEAR	RICE PROD
1950-51	20.58	1965-66	30.59	1980-81	53.63	1995-96	76.98
1951-52	21.3	1966-67	30.44	1981-82	53.25	1996-97	81.73
1952-53	22.9	1967-68	37.61	1982-83	47.12	1997-98	82.54
1953-54	28.21	1968-69	39.76	1983-84	60.1	1998-99	86.08
1954-55	25.22	1969-70	40.43	1984-85	58.34	1999-00	89.68
1955-56	27.56	1970-71	42.22	1985-86	63.83	2000-01	84.98
1956-57	29.04	1971-72	43.07	1986-87	60.56	2001-02	93.34
1957-58	25.53	1972-73	39.24	1987-88	56.86	2002-03	71.82
1958-59	30.85	1973-74	44.05	1988-89	70.49	2003-04	88.53
1959-60	31.68	1974-75	39.58	1989-90	73.57	2004-05	83.13
1960-61	34.58	1975-76	48.74	1990-91	74.29	2005-06	91.79

1961-62	35.66	1976-77	41.92	1991-92	74.68	2006-07	93.35
1962-63	33.21	1977-78	52.67	1992-93	72.86	2007-08	96.69
1963-64	37	1978-79	53.77	1993-94	80.3	2008-09	99.18
1964-65	39.31	1979-80	42.33	1994-95	81.81	2009-10	89.09
						2010-11	95.98
						2011-12	105.3

TIME SERIES:

The Time Series, thus, obtained for Prices of Rice grain is as follows:



CONCLUSION:

The Trendline of the Production of Rice from the year 1951-2012 shows that the production increases throughout. Initially, we see that the production in the beginning was increasing slowly but after the **GREEN REVOLUTION** in **1975-76** there was a rapid increase in the production of rice, and by 2012 the production rose to five times that of the production in 1950's.

GOLD PRICES

DATA:

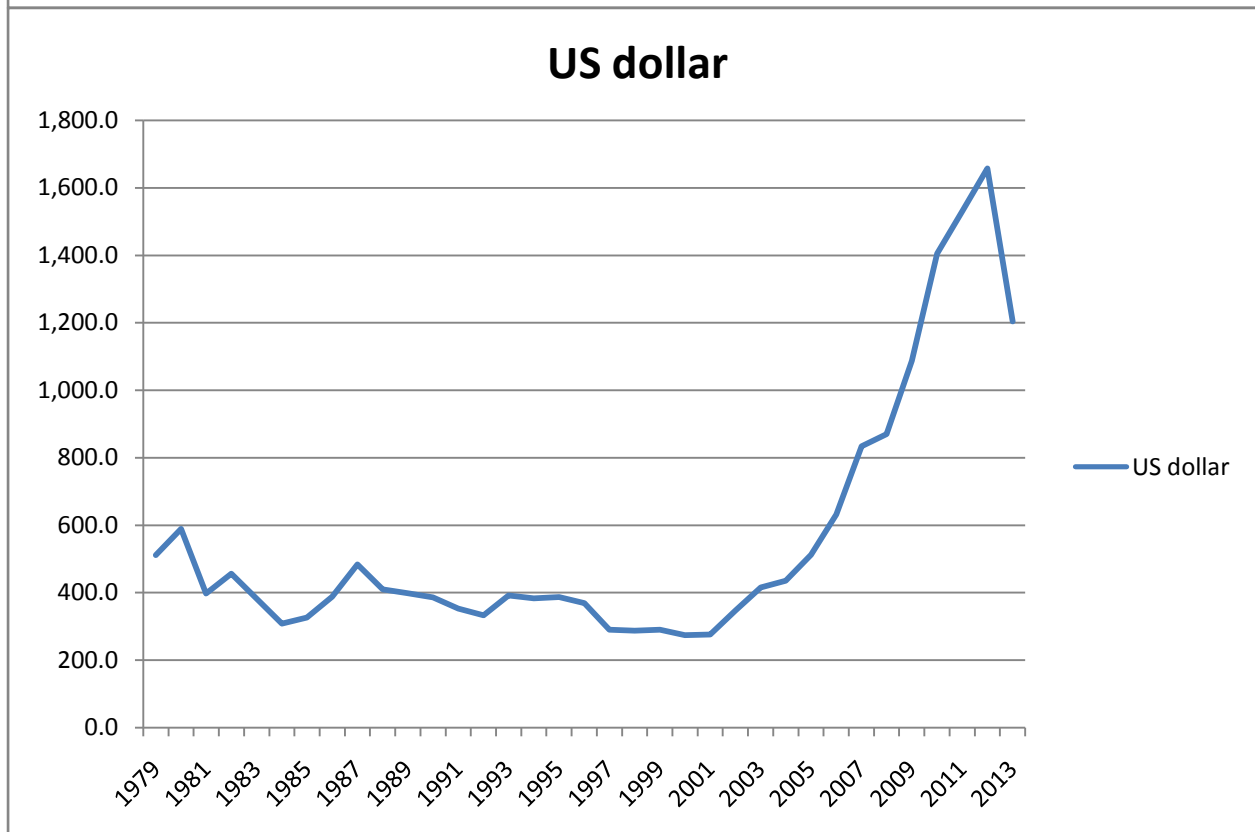
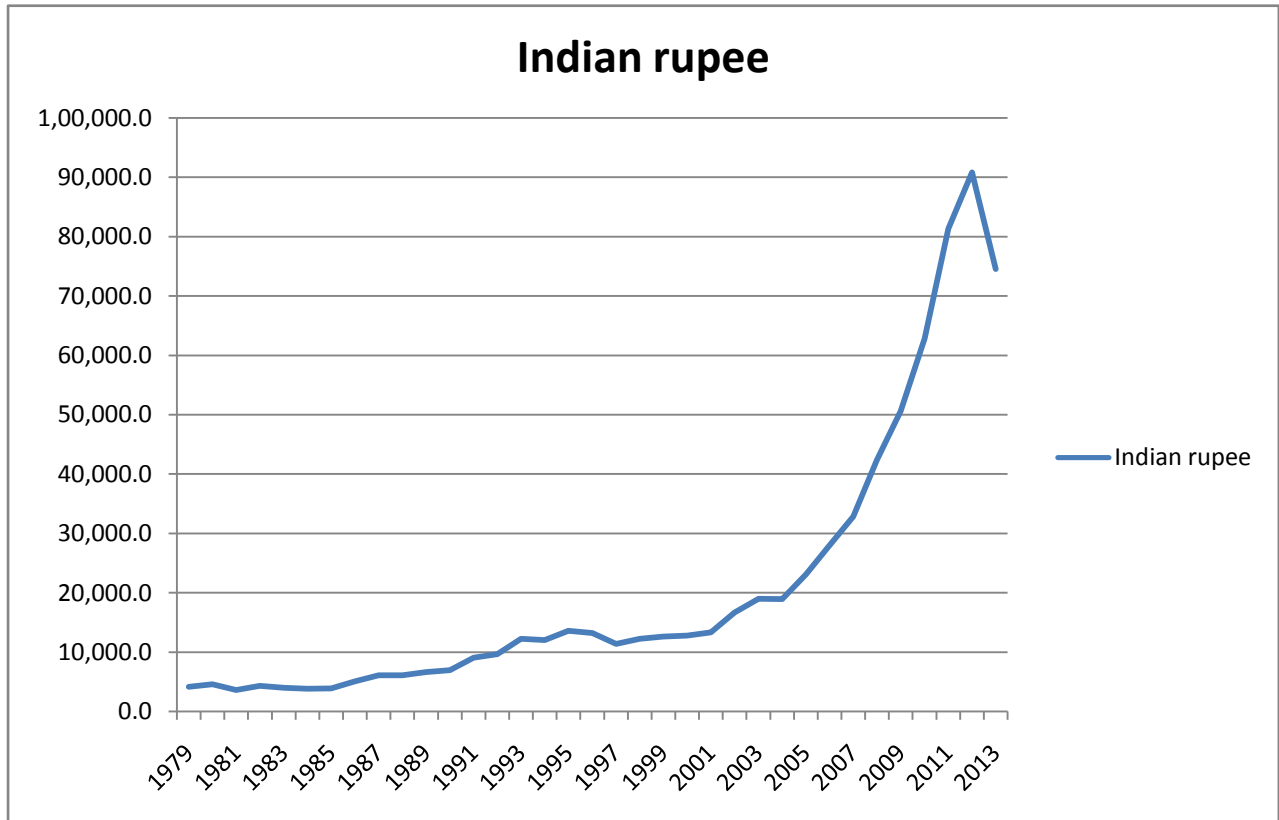
The following table gives the average annual Gold Prices in National Currency per unit Troy ounce.

(1 Troy ounce= 31.1 grams)

Year	Indian rupee	US dollar	Year	Indian rupee	US dollar
1979	4,189.60	512	1996	13,228.40	369.3
1980	4,620.30	589.8	1997	11,378.70	290.2
1981	3,625.10	397.5	1998	12,230.40	287.8
1982	4,320.00	456.9	1999	12,631.00	290.3
1983	3,995.00	382.4	2000	12,811.30	274.5
1984	3,825.80	308.3	2001	13,332.80	276.5
1985	3,920.60	326.8	2002	16,648.20	347.2
1986	5,061.00	388.8	2003	18,991.40	416.3
1987	6,116.80	484.1	2004	18,935.50	435.6
1988	6,123.10	410.3	2005	23,090.10	513
1989	6,624.80	398.6	2006	27,972.30	632
1990	6,948.90	386.2	2007	32,862.30	833.8
1991	9,080.60	353.2	2008	42,374.20	869.8
1992	9,639.30	332.9	2009	50,606.80	1,087.50
1993	12,252.60	391.8	2010	62,846.90	1,405.50
1994	12,022.10	383.3	2011	81,303.70	1,531.00
1995	13,608.80	387	2012	90,814.40	1,657.50
			2013	74,504.30	1,204.50

TIME SERIES:

The Time Series for Gold Prices in Indian Rupee and US Dollar are as obtained below:



CONCLUSION:

Here, we have plotted the trendline for the Gold Prices (in Indian rupee& US dollar/unit Troy ounce) for the years 1979-2013. In the Indian market, we see that the prices are following an increasing trend, but initially the rise in the prices was a lower and after 1999-00, there was a price hike in the gold prices. In the US market, till 2001 the trend was approximately constant but suddenly after 2002, there was a steep increase in the prices. This also gives a fair idea of the Deterioration of the Indian Rupee in comparison with the US Dollar.

METHODOLOGY

METHODOLOGY

Our project group consisted of 4 people and our study was about the changes in oil prices and their effect on the economy. We contacted both the Economics and Statistics departments of our college whose guidelines throughout the project proved to be of great help.

In this project, we have made use of Secondary Data. For data collection we have gone through different sites and finally collected the data which consisted of Oil Prices from the year 1989 to 2010 in Rupee per Litre. The data consisted of four categories namely Kerosene, Petrol, Diesel and LPG (liquefied Petroleum gas).

The data was the revision of the prices of the above four categories, so we averaged out the data on monthly basis and did the further analysis. The main focus in this project has been given on time series analysis and its applications.

Firstly, we have plotted the time series for the data then we have fitted different models for the different categories, the models being Linear, Exponential, Logarithmic, Polynomial (with different degrees). Thereby we found out the best fit on the basis of the graphs (Here we found out that though the value of R^2 value was the largest still the fit was not the best fit). Thereafter estimates were calculated from the best fitted curves using the given equation and the Residual Plots plotted using the estimates. Finally we did Exponential Smoothing of the data to reduce the error factor.

MOTIVATION

MOTIVATION

Today a common man has a lot of issues regarding his day-to-day life. Life is becoming so fast that to compete with it one has to be fast, vehicles at every house appears as a basic need today. There is at least one vehicle at everyone's place these days. If there is no gas connection you'll be frustrated cooking on coal or on a heater.

Even though the standard of living is improving, the prices of goods are pushing it down; especially the oil prices and you can see that they are soaring these days.

When our parents were at high school oil prices at that time was merely a big issue. However days are not the same if we look back the prices of oil was the same as that of the prices of chocolates today. So, you see how days change, values change, affecting the common man. How fascinating would it be if we knew about what will be the prices of oil 10 years from now? There will be a boost in the share market, but it is not possible to tell what will the exact prices of oil be after 10 years but we could try to predict the Prices by carrying a bit of statistical analysis, Forecasting and getting an Estimated value. These things in some way or the other motivated us for doing this project.

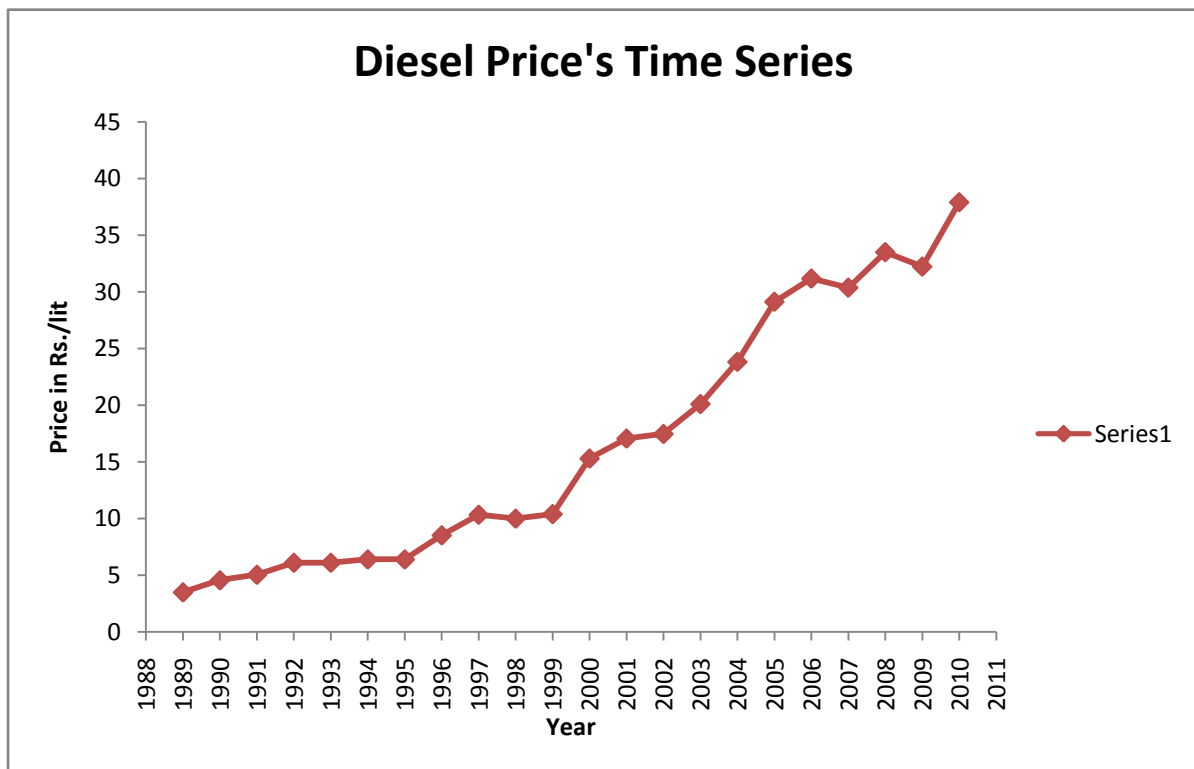
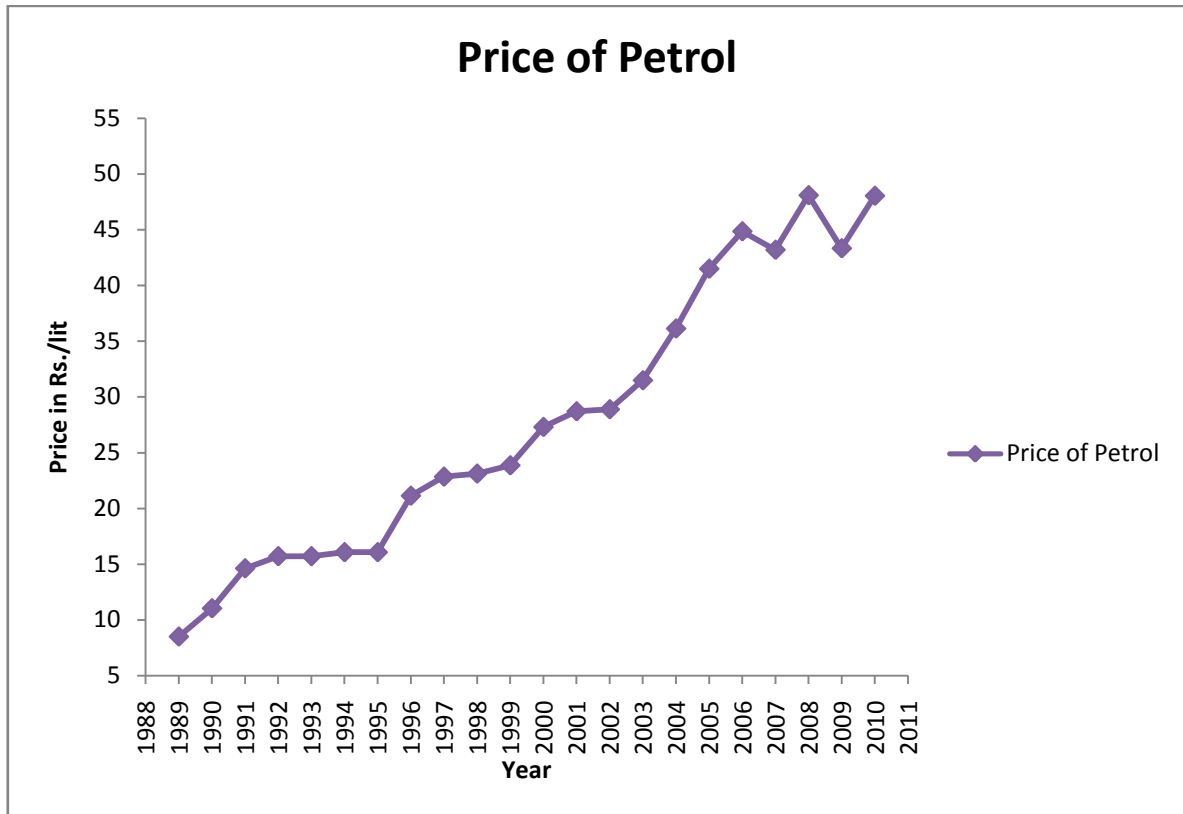
DATA

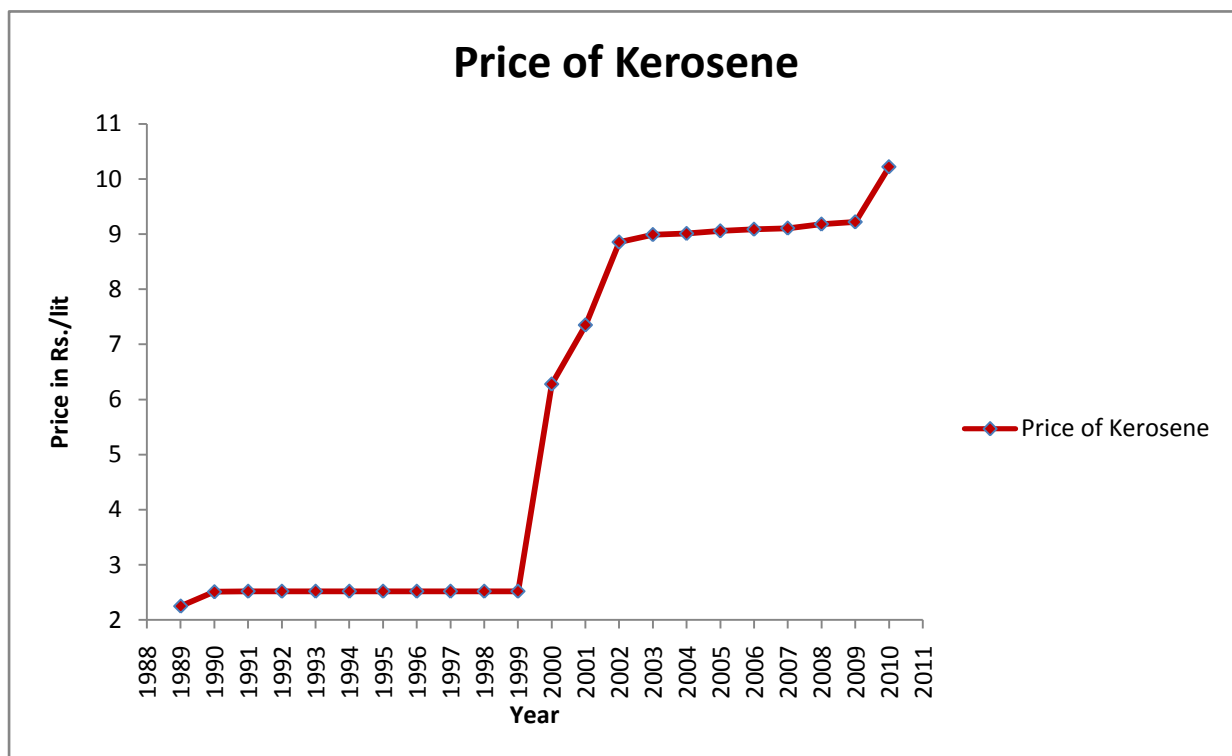
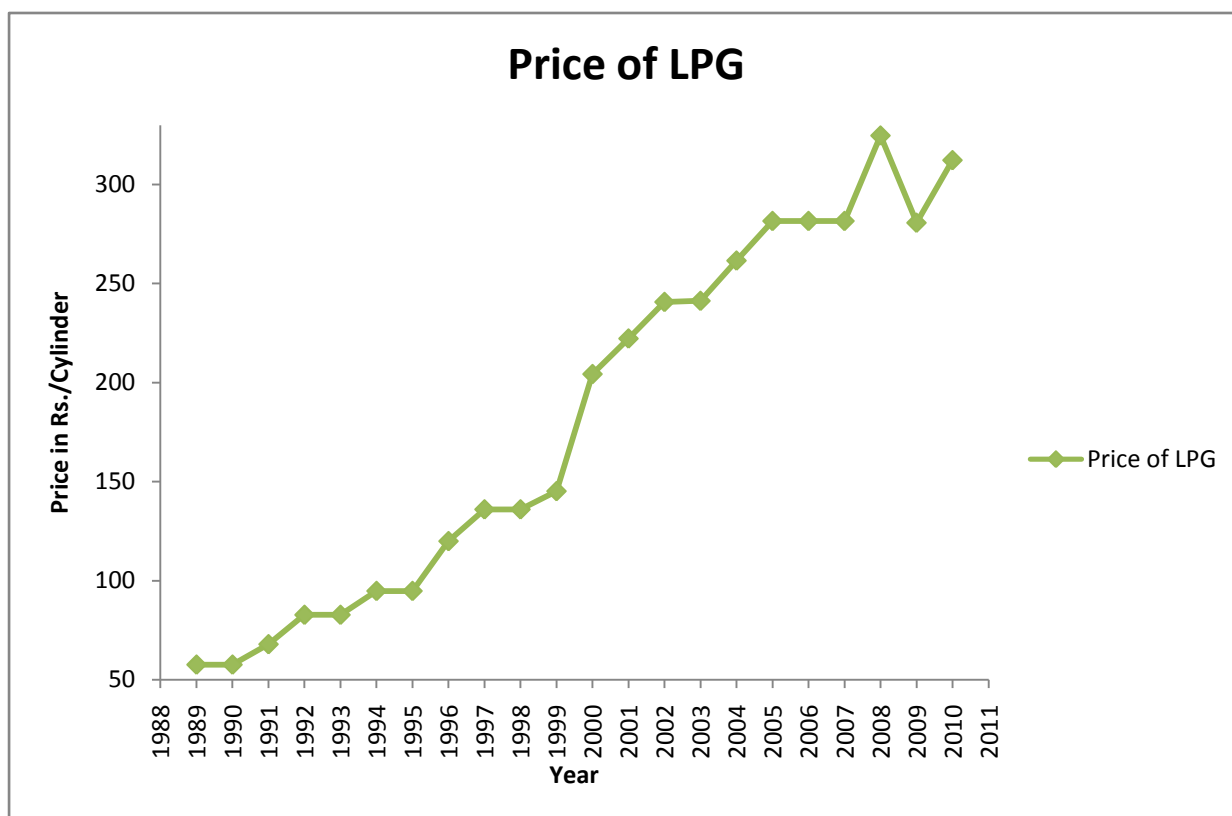
DATA OF OIL PRICES

Year	Price of Diesel (Rs./Litre)	Price of Petrol (Rs./Litre)	Price of Kerosene (Rs./Litre)	Price of LPG (Rs./Litre)
1989	3.5	8.5	2.25	57.6
1990	4.565	11.035	2.51	57.6
1991	5.05	14.62	2.52	67.9
1992	6.11	15.71	2.52	82.75
1993	6.11	15.71	2.52	82.75
1994	6.4	16.0666667	2.52	94.8
1995	6.4	16.0666667	2.52	94.8
1996	8.53	21.13	2.52	119.95
1997	10.34	22.84	2.52	136
1998	10	23.115	2.52	136
1999	10.4	23.856	2.52	145.2
2000	15.295	27.2983333	6.27833333	204.308333
2001	17.06	28.7	7.35	222.25
2002	17.463846	28.8838462	8.85461538	240.723077
2003	20.102667	31.4826667	8.99	241.28
2004	23.821667	36.1283333	9.01	261.6
2005	29.116667	41.49	9.05666667	281.6
2006	31.165	44.845	9.0875	281.6
2007	30.365	43.185	9.1075	281.6
2008	33.49	48.0833333	9.18	324.733333
2009	32.216667	43.3233333	9.22	280.7
2010	37.89	48.03	10.22	312.3

DATA ANALYSIS

TIME SERIES OF OIL PRICES





CONCLUSION:
All The above graphs show an increasing trend.

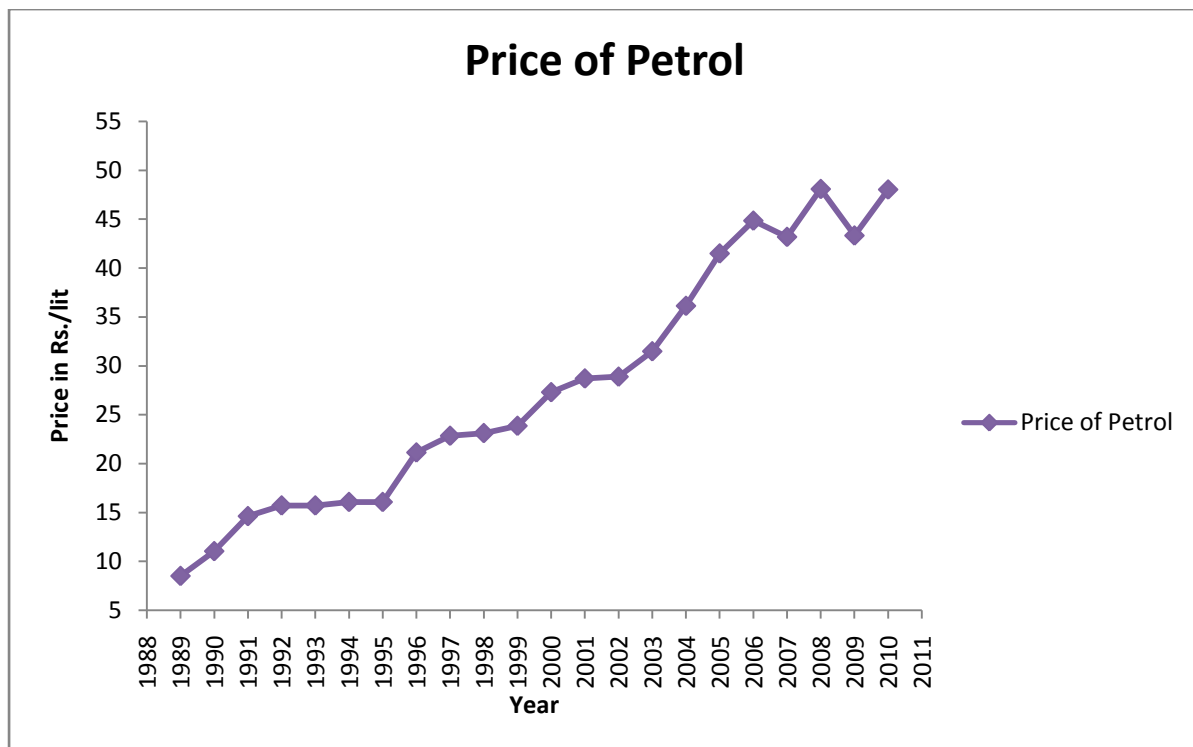
For each fuel data we could fit more than one Trend Lines, only one is selected for the analysis and estimation purpose. The selection of the graph depends on its R-squared value and whether it gives Realistic values or not.

For each fuel, Trend Line fits and their analyses are as follows-

PETROL:

DATA:

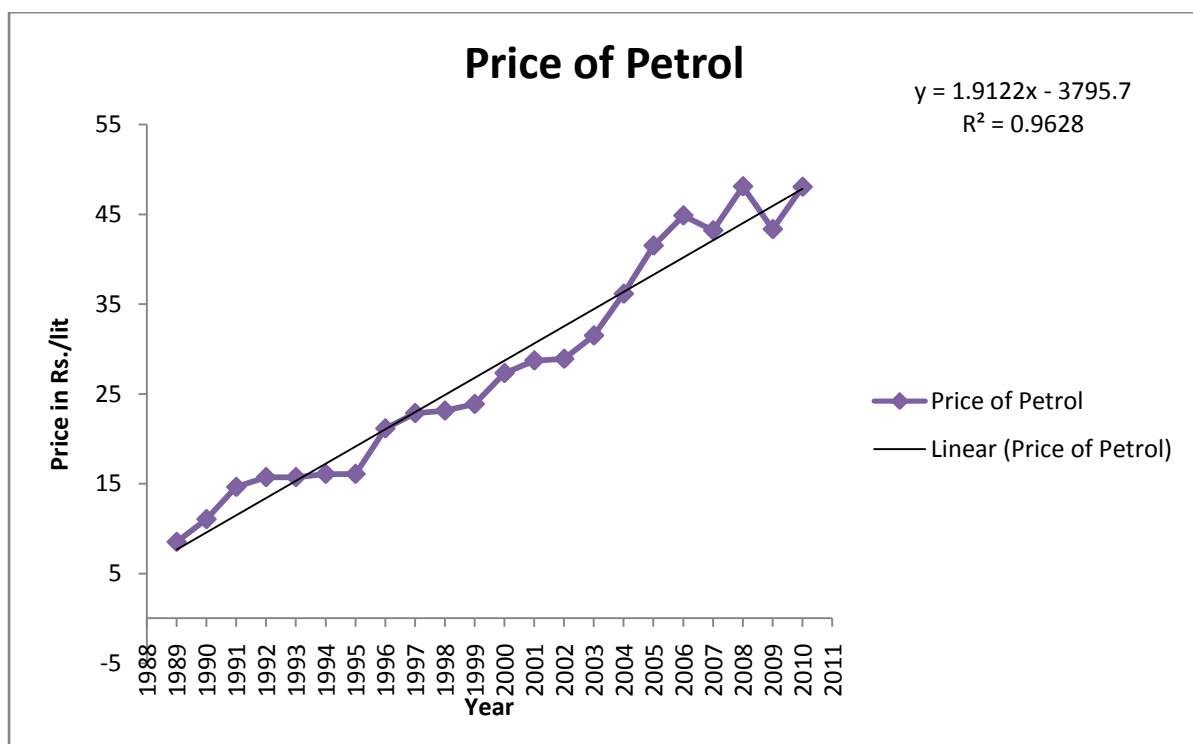
Year	Price of Petrol (Rs./Litre)
1989	8.5
1990	11.035
1991	14.62
1992	15.71
1993	15.71
1994	16.06666667
1995	16.06666667
1996	21.13
1997	22.84
1998	23.115
1999	23.856
2000	27.29833333
2001	28.7
2002	28.88384615
2003	31.48266667
2004	36.12833333
2005	41.49
2006	44.845
2007	43.185
2008	48.08333333
2009	43.32333333
2010	48.03



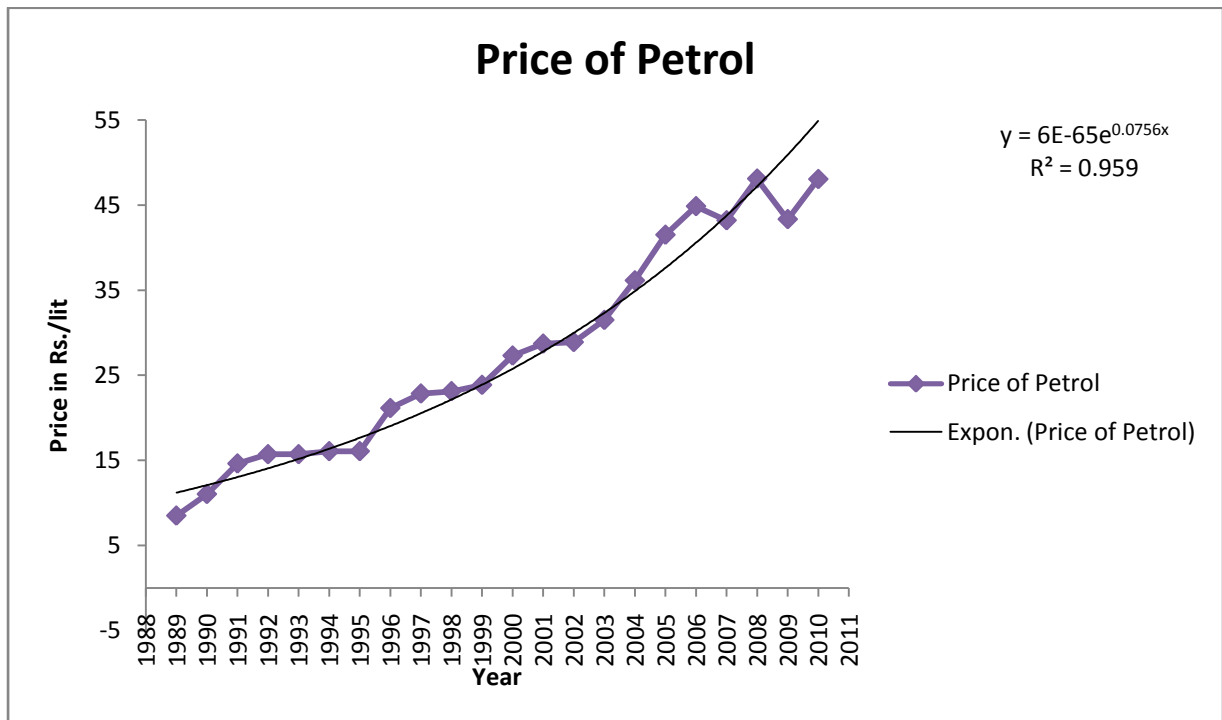
FITTING OF TREND LINES:

We fit the following Trend Lines to the the above data:

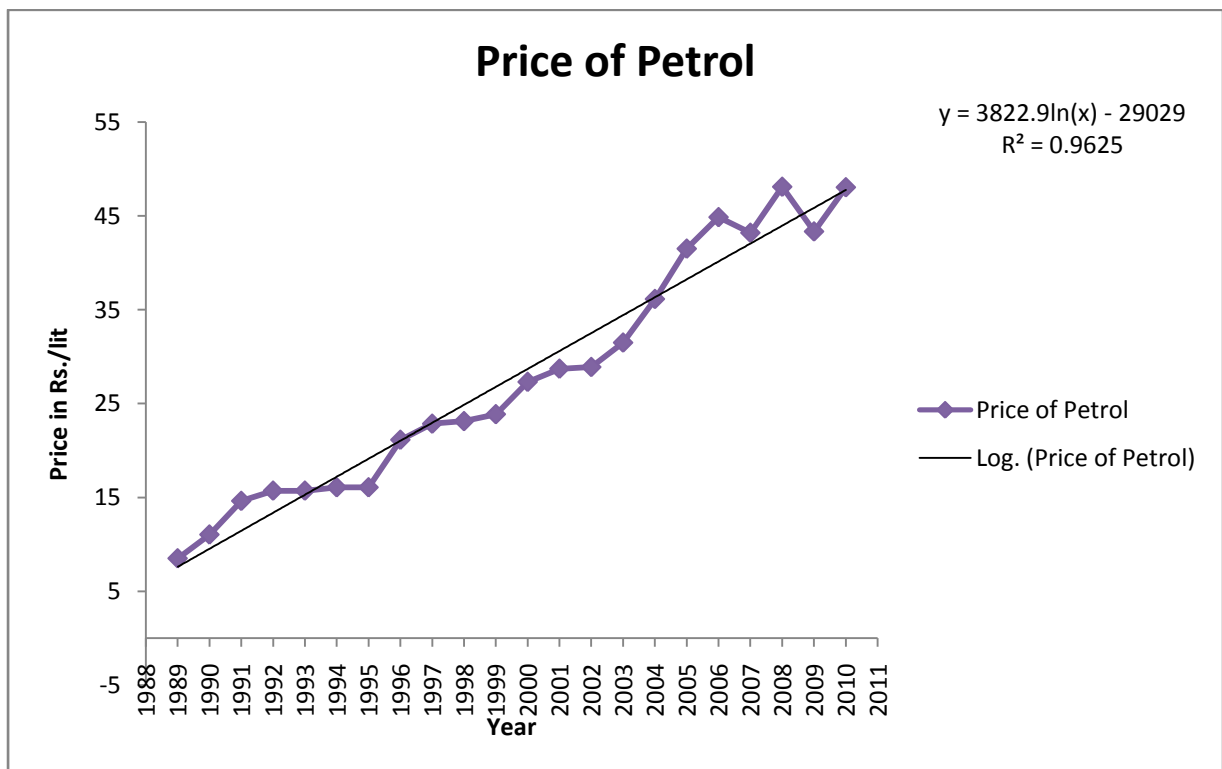
1. Linear-



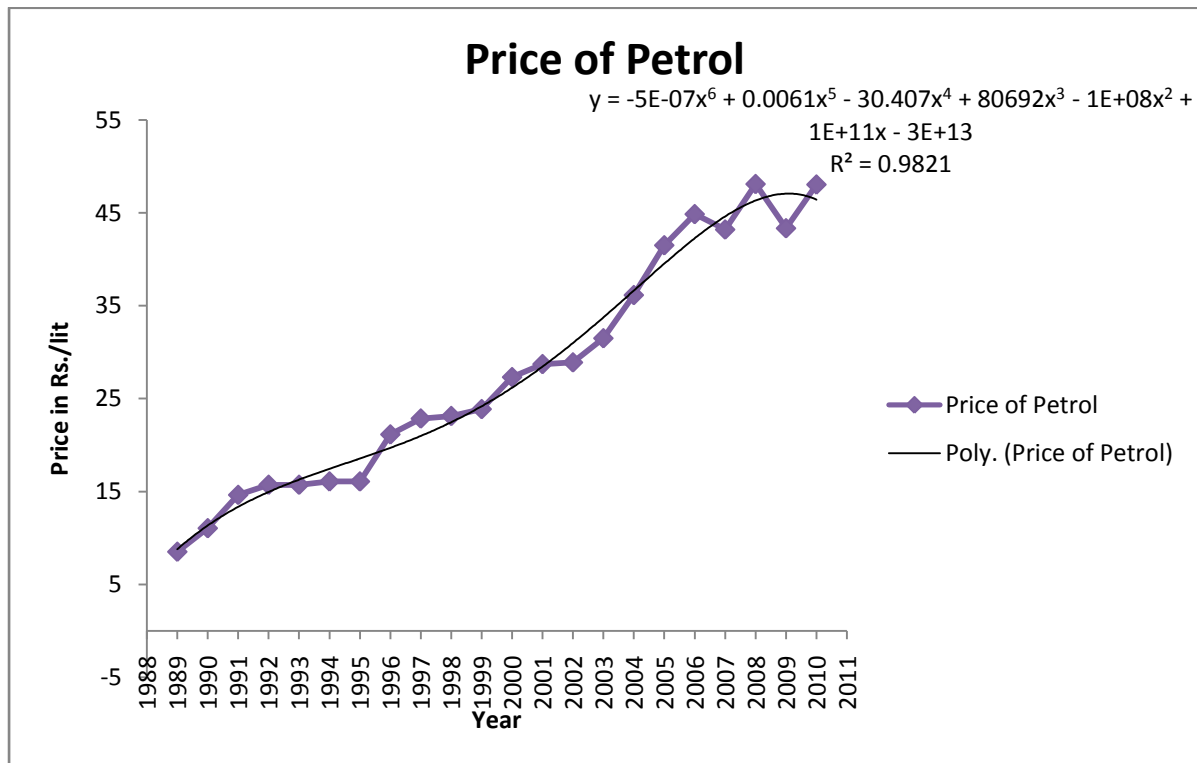
1. Exponential -



3. Logarithmic -



4. Polynomial of degree six-



EQUATIONS:

Hence we obtain the equations of the fit from the graphs as follows:

Trend Line	Equation
Exponential	$y = 6E-65e^{0.075x}$
Linear	$y = 1.912x - 3795$
Logarithmic	$y = 3822.\ln(x) - 29029$
Polynomial	$y = -5E-07x^6 + 0.006x^5 - 30.40x^4 + 80692x^3 - 1E+08x^2 + 1E+11x - 3E+13$

ESTIMATES:

Using these equations the Petrol prices were estimated as follows:

		Estimates			
Year	Price of Petrol	Polynomial	Linear	Logarithmic	Exponential
1989	8.5	-108002	7.6658	7.406028	12.08697
1990	11.035	-108001	9.578	9.327566	13.03618
1991	14.62	-107999	11.4902	11.24814	14.05992
1992	15.71	-107998	13.4024	13.16775	15.16406
1993	15.71	-107996	15.3146	15.08639	16.35491
1994	16.06666667	-107995	17.2268	17.00408	17.63928
1995	16.06666667	-107993	19.139	18.9208	19.02451
1996	21.13	-107992	21.0512	20.83656	20.51853
1997	22.84	-107990	22.9634	22.75136	22.12987
1998	23.115	-107988	24.8756	24.6652	23.86776
1999	23.856	-107986	26.7878	26.57808	25.74212
2000	27.29833333	-107984	28.7	28.49001	27.76367
2001	28.7	-107982	30.6122	30.40098	29.94398
2002	28.88384615	-107980	32.5244	32.311	32.29552
2003	31.48266667	-107978	34.4366	34.22007	34.83172
2004	36.12833333	-107976	36.3488	36.12818	37.56709
2005	41.49	-107973	38.261	38.03534	40.51727
2006	44.845	-107971	40.1732	39.94154	43.69914
2007	43.185	-107968	42.0854	41.8468	47.13088
2008	48.08333333	-107966	43.9976	43.75111	50.83212
2009	43.32333333	-107963	45.9098	45.65447	54.82402
2010	48.03	-107961	47.822	47.55689	59.12941

Here we observe that prices given by the Polynomial fit are not realistic hence we do not consider it.

SUM OF SQUARE OF RESIDUALS (SSR):

For the other fits we calculate the residuals and their SSR.

The one with least SSR is considered as the best fit.

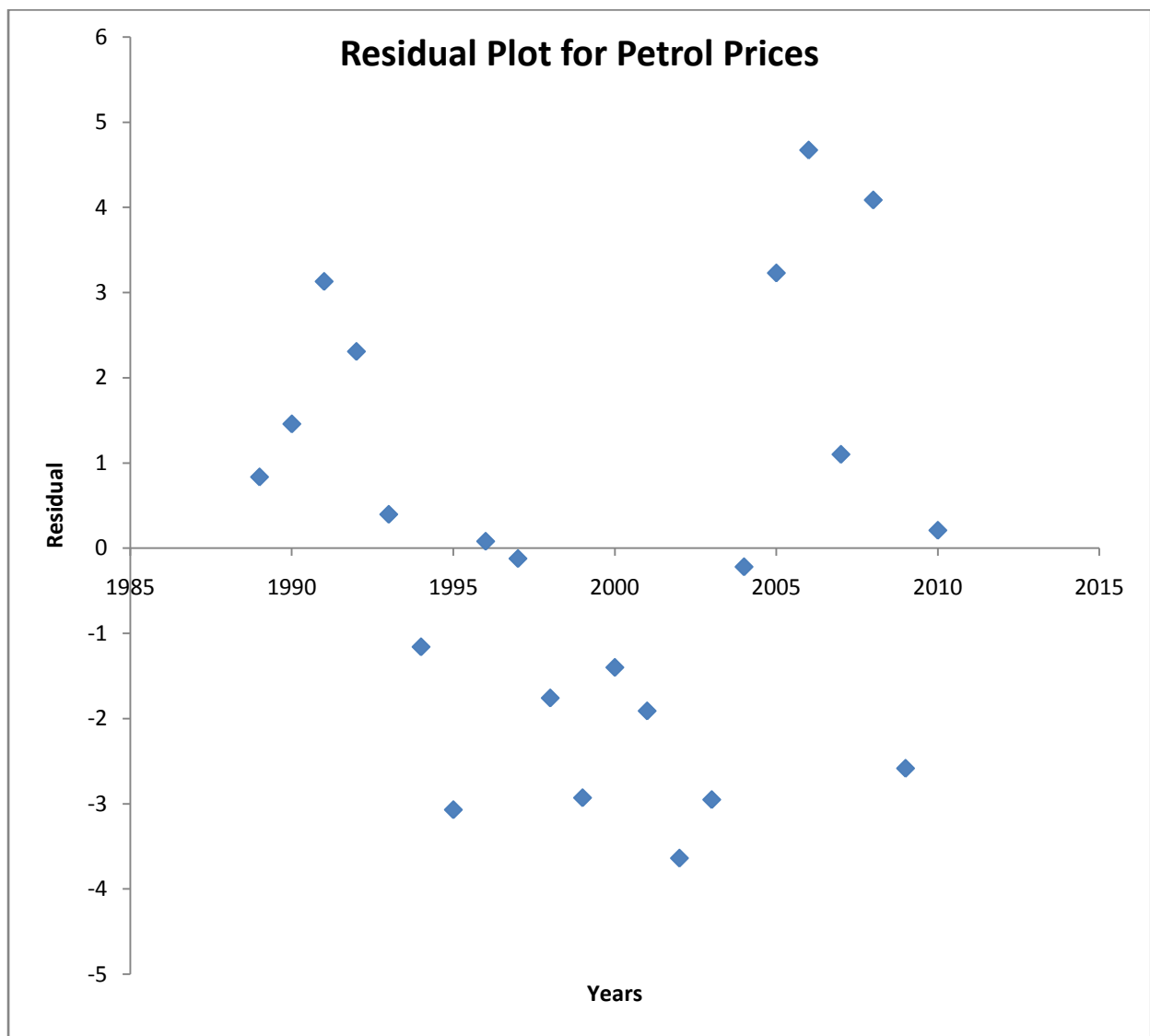
Year	Price of Petrol	Linear		Logarithmic		Exponential	
		Residual	SSR	Residual	SSR	Residual	SSR
1989	8.5	0.8342	0.69589	1.093972	1.196774	11.25277	12.86639
1990	11.035	1.457	2.122849	1.707434	2.915329	11.57918	4.004713
1991	14.62	3.1298	9.795648	3.371861	11.36945	10.93012	0.313686
1992	15.71	2.3076	5.325018	2.542252	6.463047	12.85646	0.298046
1993	15.71	0.3954	0.156341	0.623607	0.388886	15.95951	0.415914
1994	16.06666667	-	-	-	-	-	-
		1.16013	1.345909	0.93741	0.878734	18.79942	2.473122
1995	16.06666667	-	-	-	-	-	-
		3.07233	9.439232	2.85413	8.146055	22.09685	8.748865
1996	21.13	0.0788	0.006209	0.293443	0.086109	20.43973	0.373895
1997	22.84	-	-	-	-	-	-
		0.1234	0.015228	0.088642	0.007857	22.25327	0.504281
1998	23.115	-	-	-	-	-	-
		1.7606	3.099712	1.5502	2.40312	25.62836	0.566641
1999	23.856	-	-	-	-	-	-
		2.9318	8.595451	2.72208	7.409744	28.67392	3.557433
20	27.2983	-	-	-	-	-	-
		-	1.9646	-	1.420	29.16	0.216

00	3333	1.401 67	69	1.191 68	099	534	54
20 01	28.7	- 1.912 2	3.6565 09	- 1.700 98	2.893 35	31.85 618	1.547 493
20 02	28.8838 4615	- 3.640 55	13.253 63	- 3.427 16	11.74 54	35.93 607	11.63 949
20 03	31.4826 6667	- 2.953 93	8.7257 22	- 2.737 4	7.493 356	37.78 565	11.21 614
20 04	36.1283 3333	- 0.220 47	0.0486 06	0.000 156	2.45E- 08	37.78 756	2.070 021
20 05	41.49	3.229	10.426 44	3.454 664	11.93 47	37.28 827	0.946 196
20 06	44.845	4.671 8	21.825 72	4.903 456	24.04 388	39.02 734	1.312 997
20 07	43.185	1.099 6	1.2091 2	1.338 198	1.790 774	46.03 128	15.56 996
20 08	48.0833 3333	4.085 733	16.693 22	4.332 223	18.76 815	46.74 638	7.555 817
20 09	43.3233 3333	- 2.586 47	6.6898 1	- 2.331 14	5.434 205	57.41 049	132.2 658
20 10	48.03	0.208	0.0432 64	0.473 115	0.223 838	58.92 141	123.1 969
Total-			125.1 342		127.0 129		341.6 603

The SSR of Linear is the minimum. Hence Linear fit is the best fit for the Petrol prices.

RESIDUAL PLOT:

After fitting the appropriate Trend Line to the oil prices, the prices were estimated for the collected data using the equations. Then the residuals were obtained as the difference of the observed prices and the estimated prices. Then a graph of the year and the corresponding residual were plotted to obtain a Residual Plot to study the goodness of the fit.



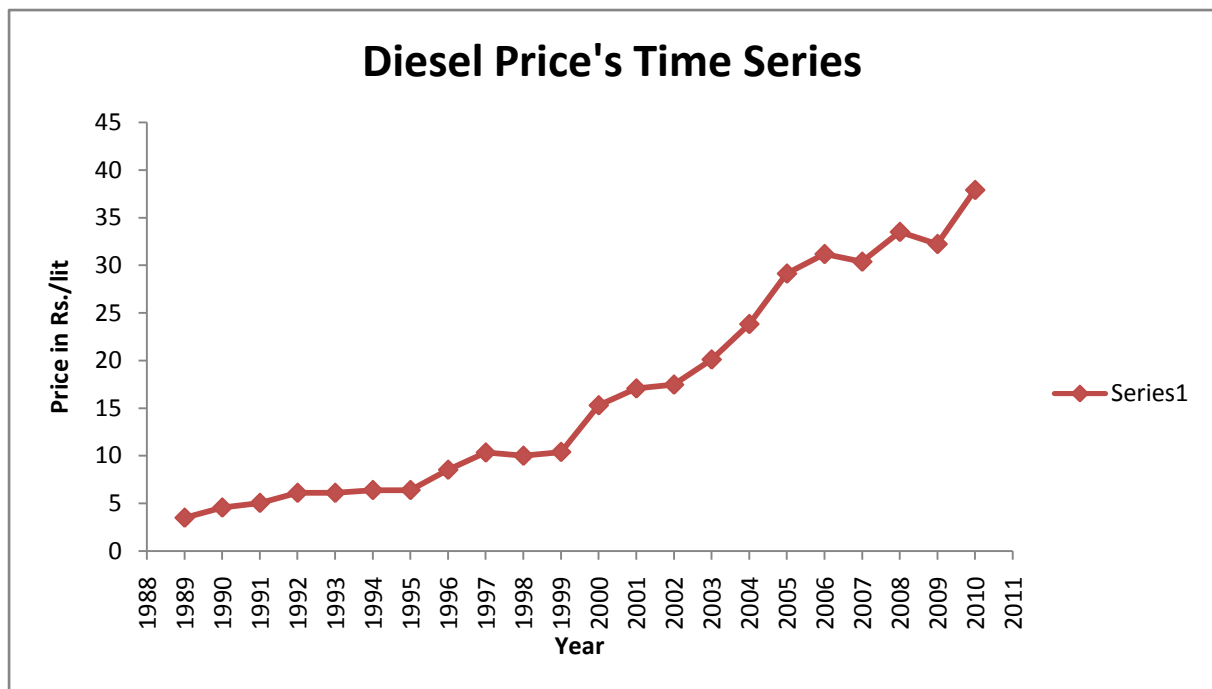
CONCLUSION:

From the graphs we can conclude that the Linear fit for Petrol is good as the Residual Plot does not show any pattern

DIESEL:

DATA:

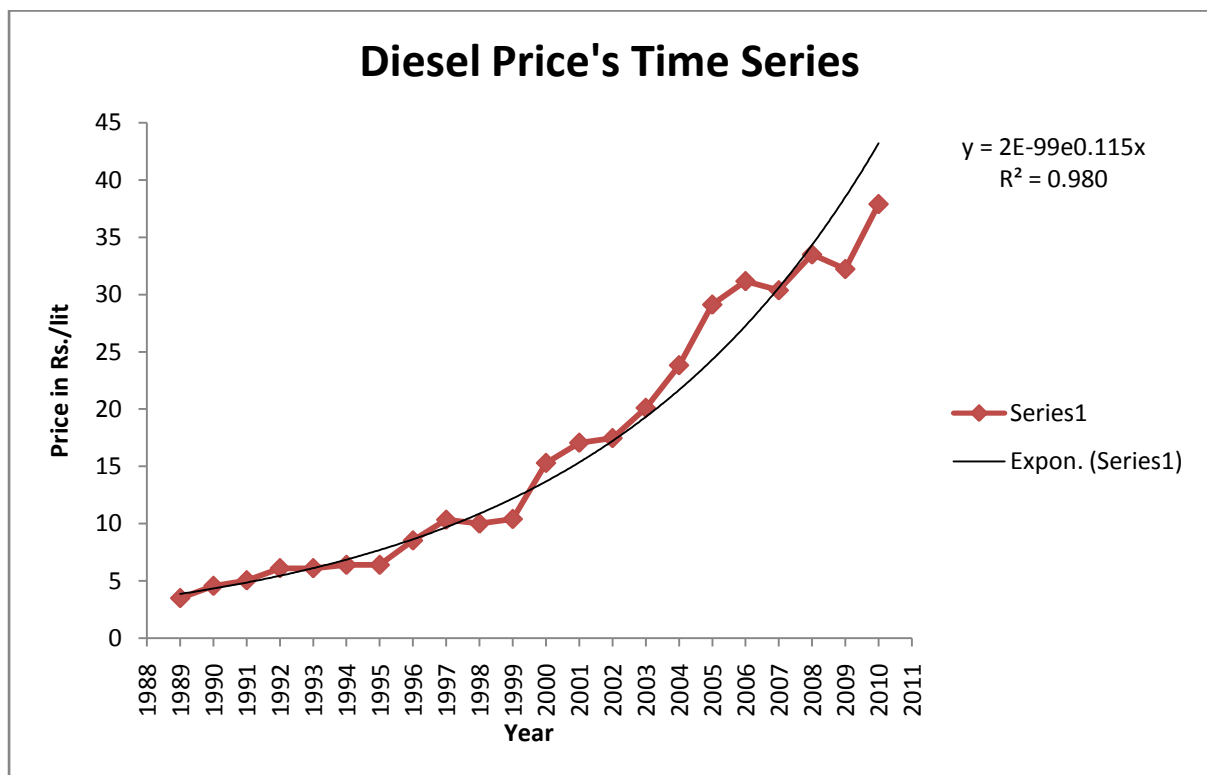
Year	Price of Diesel (Rs./Litre)
1989	3.5
1990	4.565
1991	5.05
1992	6.11
1993	6.11
1994	6.4
1995	6.4
1996	8.53
1997	10.34
1998	10
1999	10.4
2000	15.295
2001	17.06
2002	17.463846
2003	20.102667
2004	23.821667
2005	29.116667
2006	31.165
2007	30.365
2008	33.49
2009	32.216667
2010	37.89



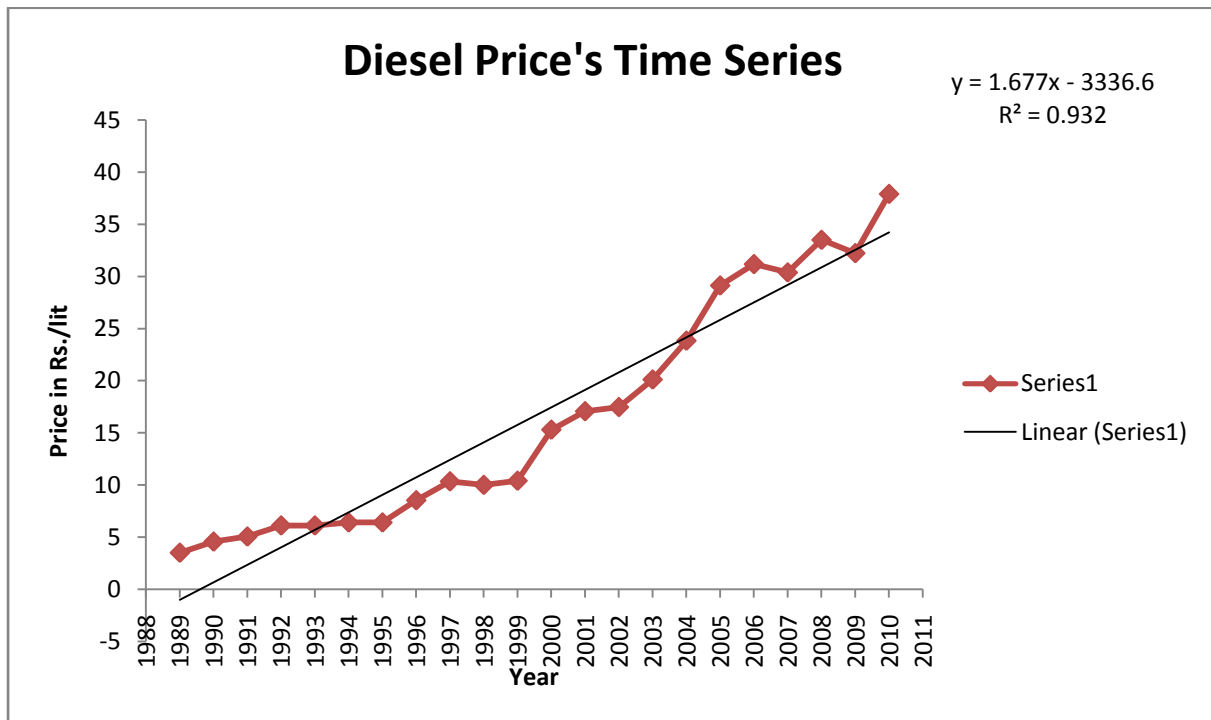
TREND LINE:

We fit the following Trend Lines to the above data:

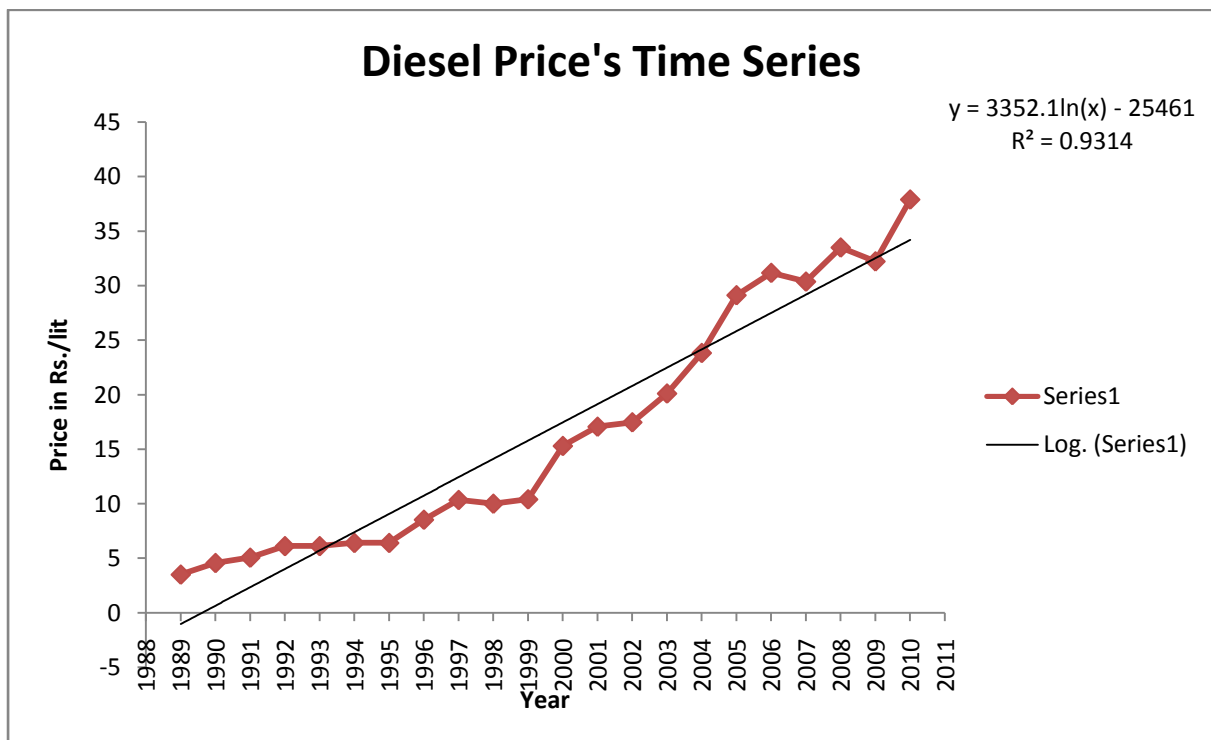
1. Exponential-



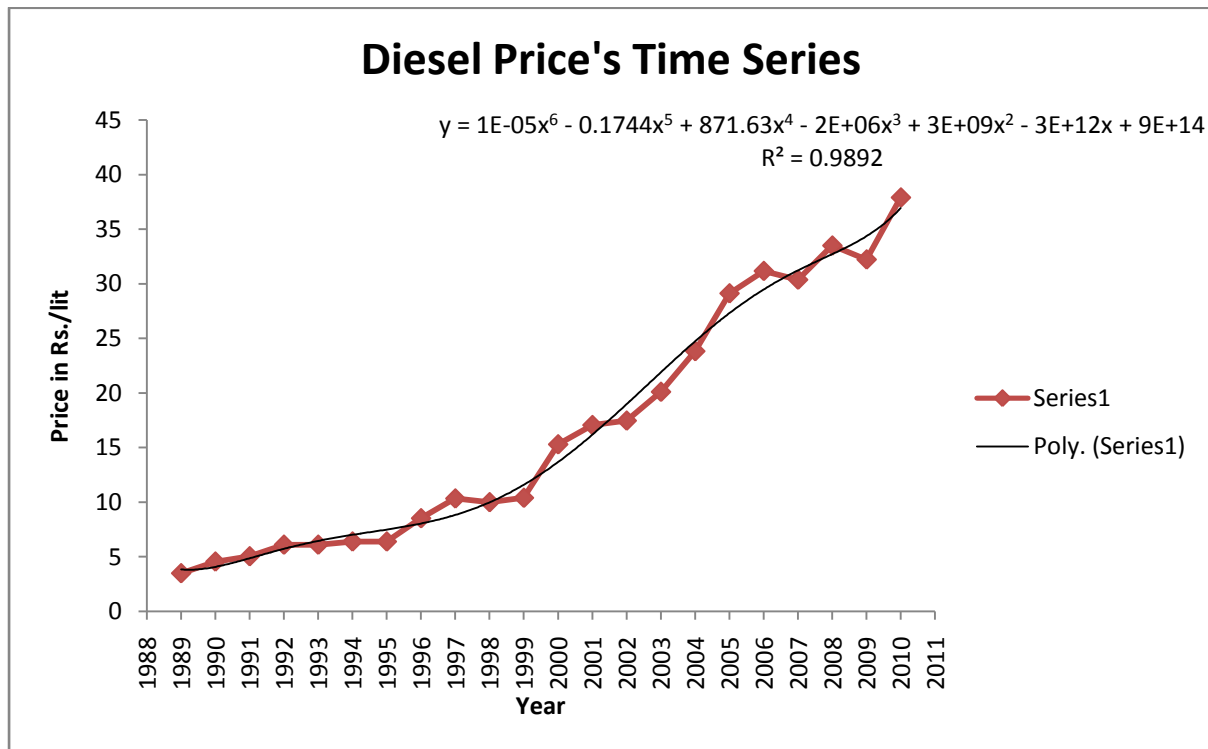
2. Linear-



3. Logarithmic-



4. Polynomial-



EQUATIONS:

Hence we obtain the equations of the fit from the graphs as follows:

<i>Trend Line</i>	<i>Equation</i>
Exponential	$y = 2E-99e^{0.115x}$
Linear	$y = 1.677x - 3336$
Logarithmic	$y = 3352.1\ln(x) - 25461$
Polynomial	$y = 1E-05x^6 - 0.174x^5 + 871.6x^4 - 2E+06x^3 + 3E+09x^2 - 3E+12x + 9E+14$

ESTIMATES:

Using these equations the Diesel prices were estimated as follows:

Year	Price of Diesel	Estimation			
		Linear	Logarithmic	Exponential	Polynomial
1989	3.5	- 0.447	- 1.26184128	4.35891396	- 9.21545E+13
1990	4.565	1.23	0.42300419	4.89014979	- 9.12718E+13
1991	5.05	2.907	2.10700322	5.48612916	- 9.03882E+13
1992	6.11	4.584	3.79015666	6.15474258	- 8.95039E+13
1993	6.11	6.261	5.47246535	6.90484221	- 8.86187E+13
1994	6.4	7.938	7.15393015	7.74635907	- 8.77328E+13
1995	6.4	9.615	8.83455189	8.69043448	- 8.6846E+13
1996	8.53	11.292	10.5143314	9.7495676	- 8.59584E+13
1997	10.34	12.969	12.1932696	10.9377809	-8.507E+13
1998	10	14.646	13.8713673	12.2708059	- 8.41808E+

					13
199 9	10.4	16.32 3	15.548625 2	13.766291 2	- 8.32908E+ 13
200 0	15.295	18	17.225044 4	15.444036 4	-8.24E+13
200 1	17.06	19.67 7	18.900625 5	17.326254 2	- 8.15084E+ 13
200 2	17.463846 15	21.35 4	20.575369 5	19.437864 4	- 8.0616E+1 3
200 3	20.102666 67	23.03 1	22.249277 2	21.806823 7	- 7.97228E+ 13
200 4	23.821666 67	24.70 8	23.922349 3	24.464496 3	- 7.88289E+ 13
200 5	29.116666 67	26.38 5	25.594586 8	27.446068 5	- 7.79341E+ 13
200 6	31.165	28.06 2	27.265990 5	30.791015 3	- 7.70385E+ 13
200 7	30.365	29.73 9	28.936561 2	34.543622 1	- 7.61422E+ 13
200 8	33.49	31.41 6	30.606299 7	38.753572 1	- 7.52451E+ 13
200 9	32.216666 67	33.09 3	32.275206 9	43.476603 2	- 7.43472E+ 13
201 0	37.89	34.77	33.943283 5	48.775246 2	- 7.34485E+ 13

Here we observe that prices given by the Polynomial line are not realistic hence we do not consider it.

SUM OF SQUARES OF RESIDUALS (SSR):

For the other fits we calculate the residuals and their SSR. The one with least SSR is considered as the best fit.

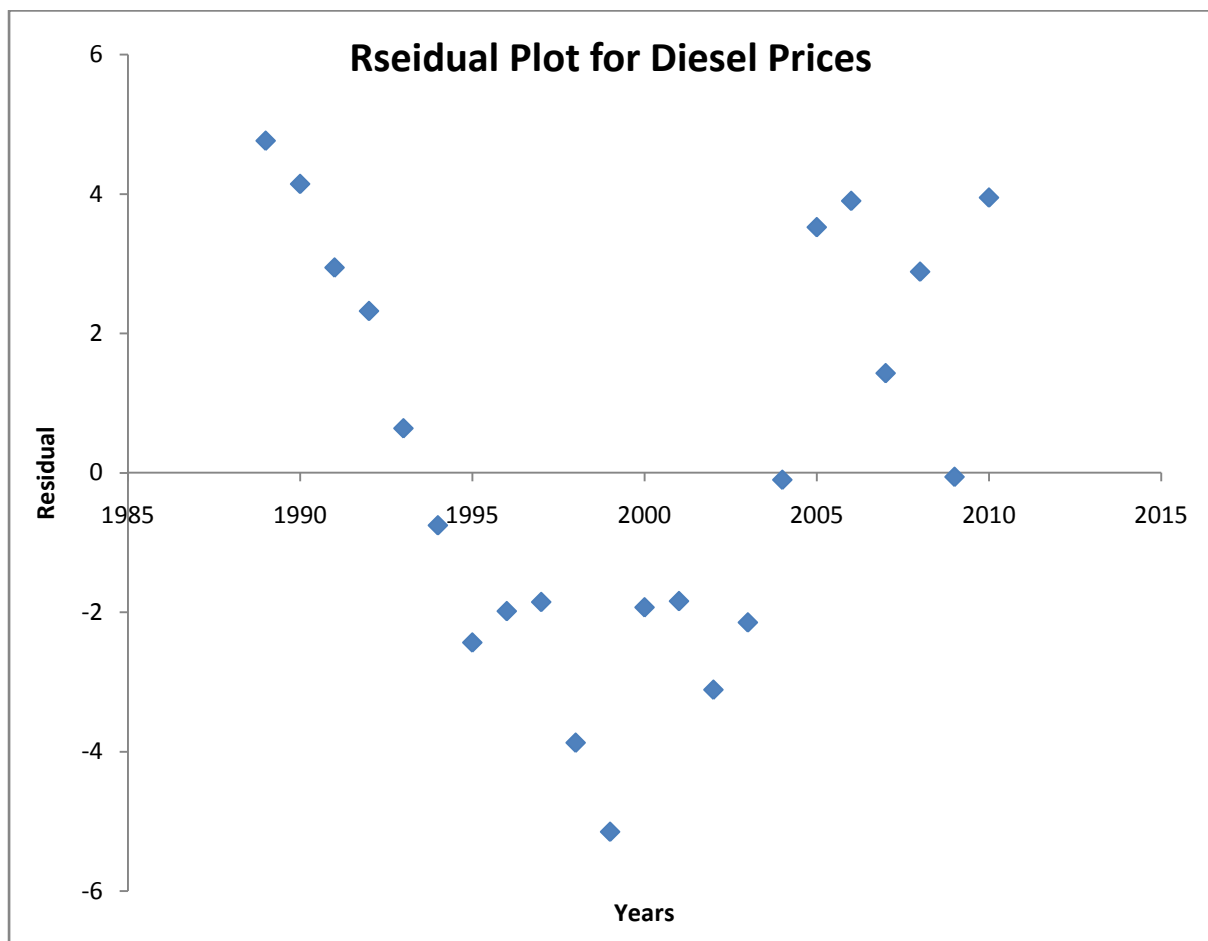
Year	Price of Diesel	Linear		Logarithmic		Exponential	
		Residual	SSR	Residual	SSR	Residual	SSR
1989	3.5	3.947	15.57881	4.761841	22.67513	-0.85891	0.737733
1990	4.565	3.335	11.12222	4.141996	17.15613	-0.32515	0.105722
1991	5.05	2.143	4.592449	2.942997	8.66123	-0.43613	0.190209
1992	6.11	1.526	2.328676	2.319843	5.381673	-0.04474	0.002002
1993	6.11	-0.151	0.022801	0.637535	0.40645	-0.79484	0.631774
1994	6.4	-1.538	2.365444	-0.75393	0.568411	-1.34636	1.812683
1995	6.4	-3.215	10.33623	-2.43455	5.927043	-2.29043	5.24609
1996	8.53	-2.762	7.628644	-1.98433	3.937571	-1.21957	1.487345
1997	10.34	-2.629	6.911641	-1.85327	3.434608	-0.59778	0.357342
1998	10	-4.646	21.58532	-3.87137	14.98748	-2.27081	5.156559
1999	10.4	-5.923	35.08193	-5.14863	26.50834	-3.36629	11.33192

20 00	15.295	-2.705	7.317 025	- 1.930 04	3.7250 71	- 0.149 04	0.022 212
20 01	17.06	-2.617	6.848 689	- 1.840 63	3.3879 02	- 0.266 25	0.070 891
20 02	17.4638 4615	- 3.890 15	15.13 33	- 3.111 52	9.6815 78	- 1.974 02	3.896 748
20 03	20.1026 6667	- 2.928 33	8.575 136	- 2.146 61	4.6079 37	- 1.704 16	2.904 151
20 04	23.8216 6667	- 0.886 33	0.785 587	- 0.100 68	0.0101 37	- 0.642 83	0.413 23
20 05	29.1166 6667	2.731 667	7.462 003	3.522 08	12.405 05	1.670 598	2.790 898
20 06	31.165	3.103	9.628 609	3.899 01	15.202 28	0.373 985	0.139 865
20 07	30.365	0.626	0.391 876	1.428 439	2.0404 38	- 4.178 62	17.46 088
20 08	33.49	2.074	4.301 476	2.883 7	8.3157 28	- 5.263 57	27.70 519
20 09	32.2166 6667	- 0.876 33	0.767 96	- 0.058 54	0.0034 27	- 11.25 99	126.7 862
20 10	37.89	3.12	9.734 4	3.946 716	15.576 57	- 10.88 52	118.4 886
	Total-		188.5 002		184.6 002		327.7 382

The SSR of Logarithmic is the minimum. Hence Logarithmic fit is the best fit for the Diesel prices.

RESIDUAL PLOT:

After fitting the appropriate Trend Line to the oil prices, the prices were estimated for the collected data using the equations. Then the residuals were obtained as the difference of the observed prices and the estimated prices. Then a graph of the year and the corresponding residual were plotted to obtain a Residual Plot to study the goodness of the fit.



CONCLUSION:

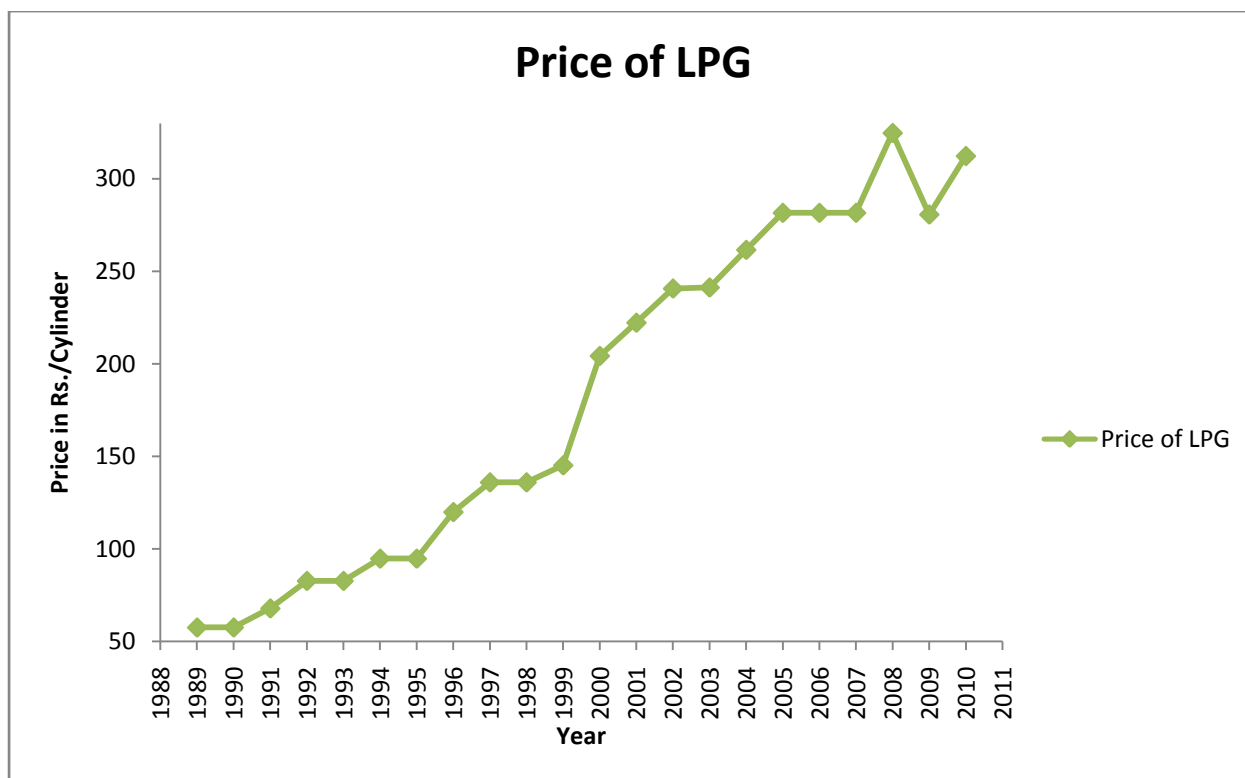
The Residual Plot of Diesel shows a certain pattern, which means that the fit is not very good. But due to the limitations of Microsoft Excel other fits, which could be better, cannot be used for the estimation. Hence we proceed with Logarithmic Fit as the best fit for Diesel Prices.

LPG:

DATA:

(1 Cylinder= 14.2 Litre)

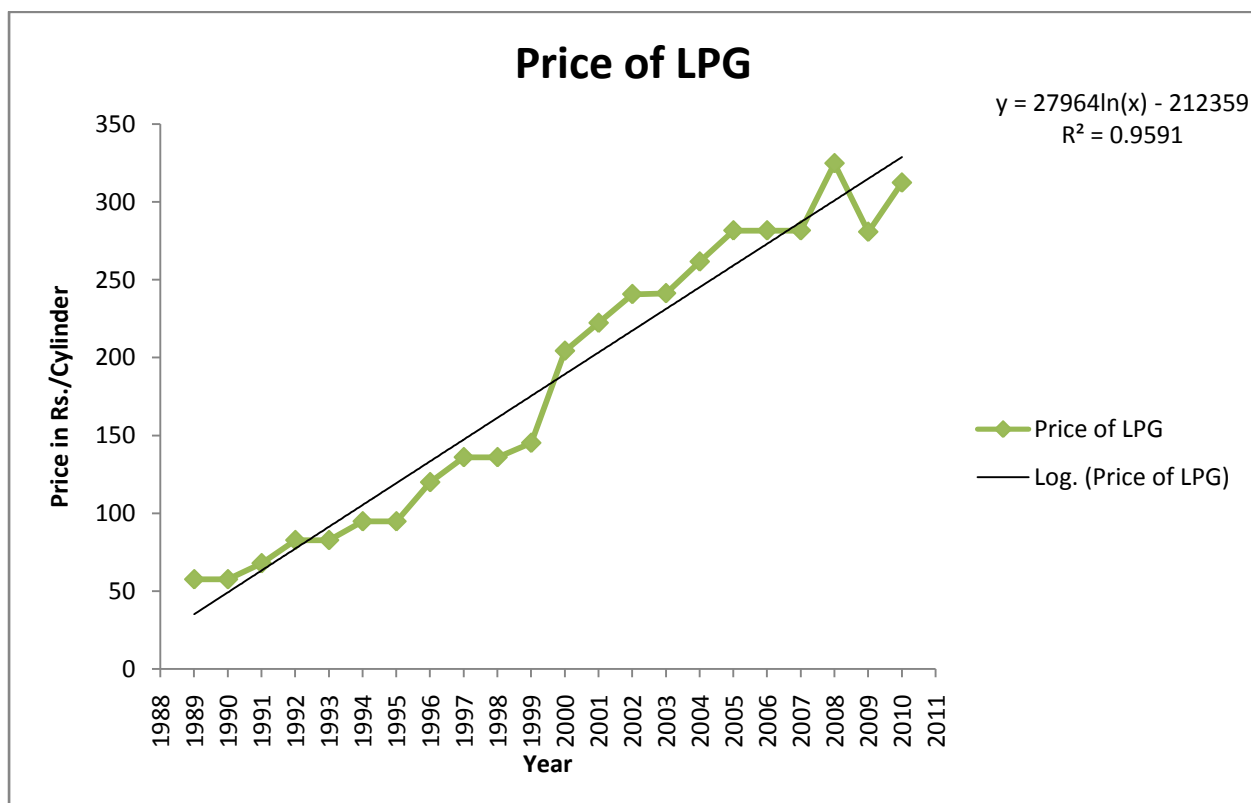
Year	Price of LPG (Rs./Cylinder)
1989	57.6
1990	57.6
1991	67.9
1992	82.75
1993	82.75
1994	94.8
1995	94.8
1996	119.95
1997	136
1998	136
1999	145.2
2000	204.3083
2001	222.25
2002	240.7231
2003	241.28
2004	261.6
2005	281.6
2006	281.6
2007	281.6
2008	324.7333
2009	280.7
2010	312.3



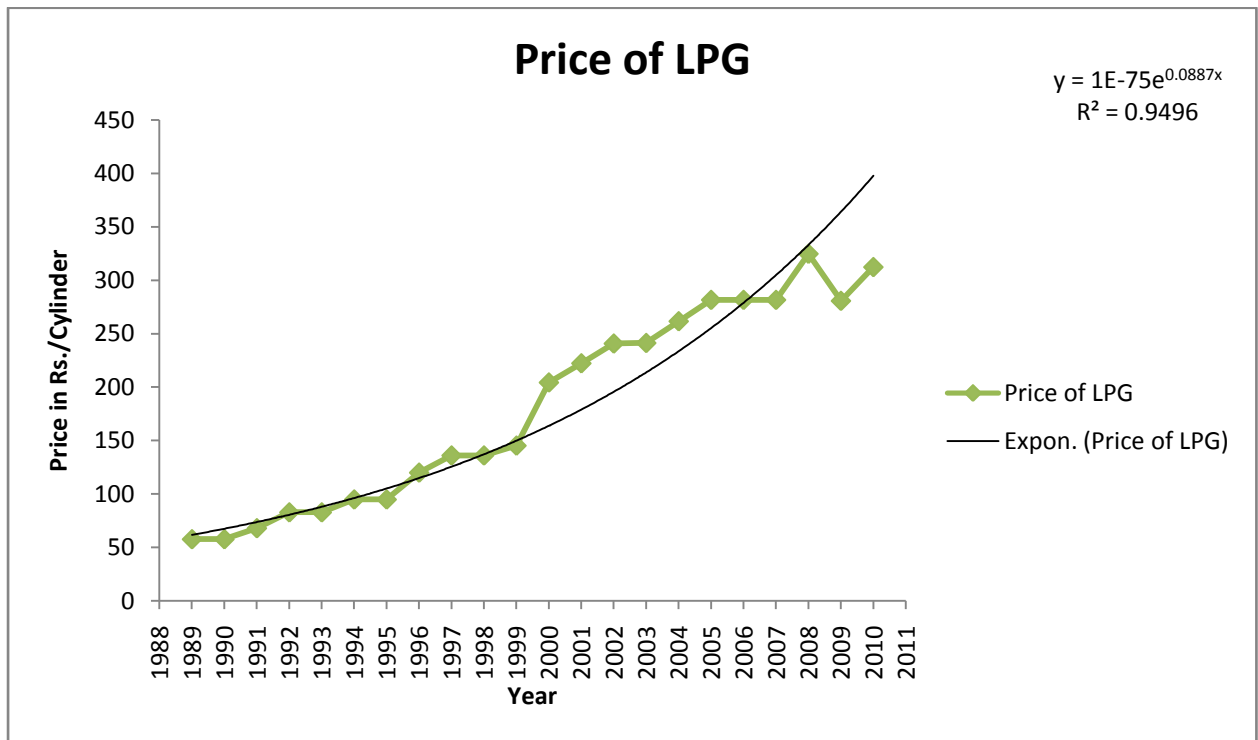
TREND LINE:

We fit the following Trend Lines to the above data:

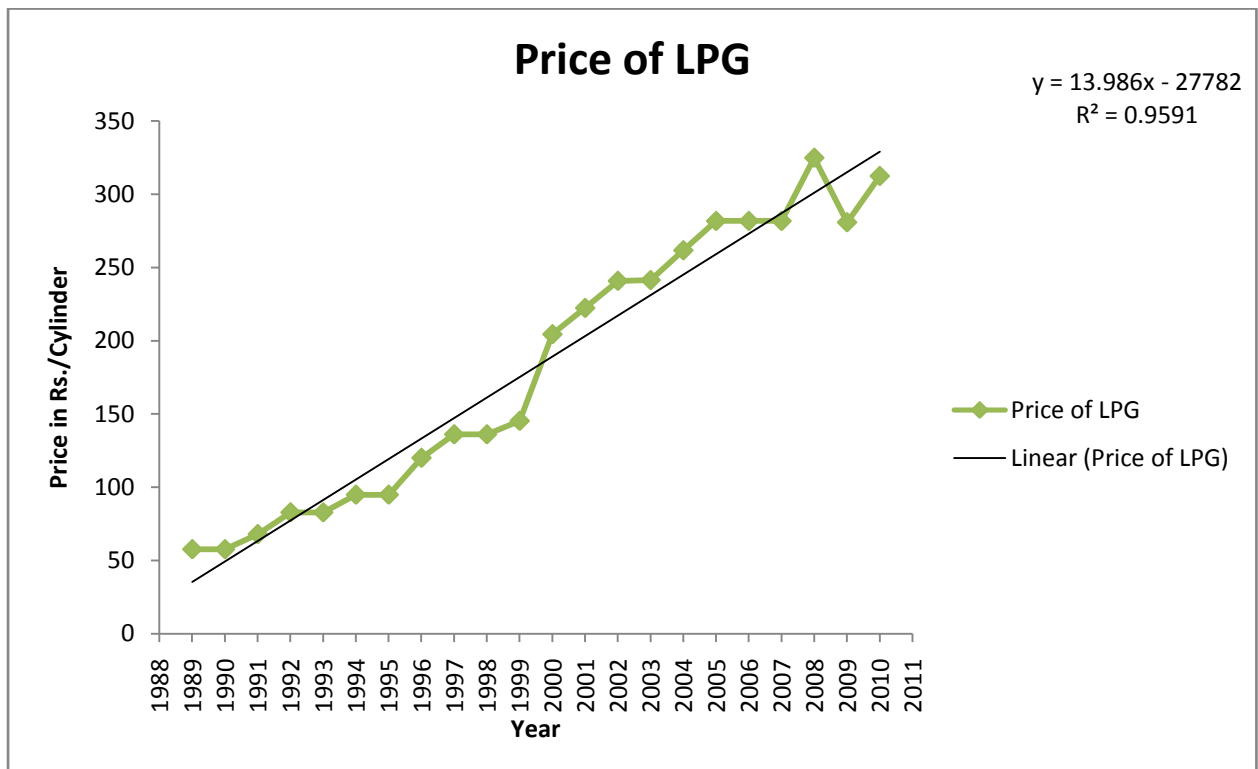
1. Logarithmic-



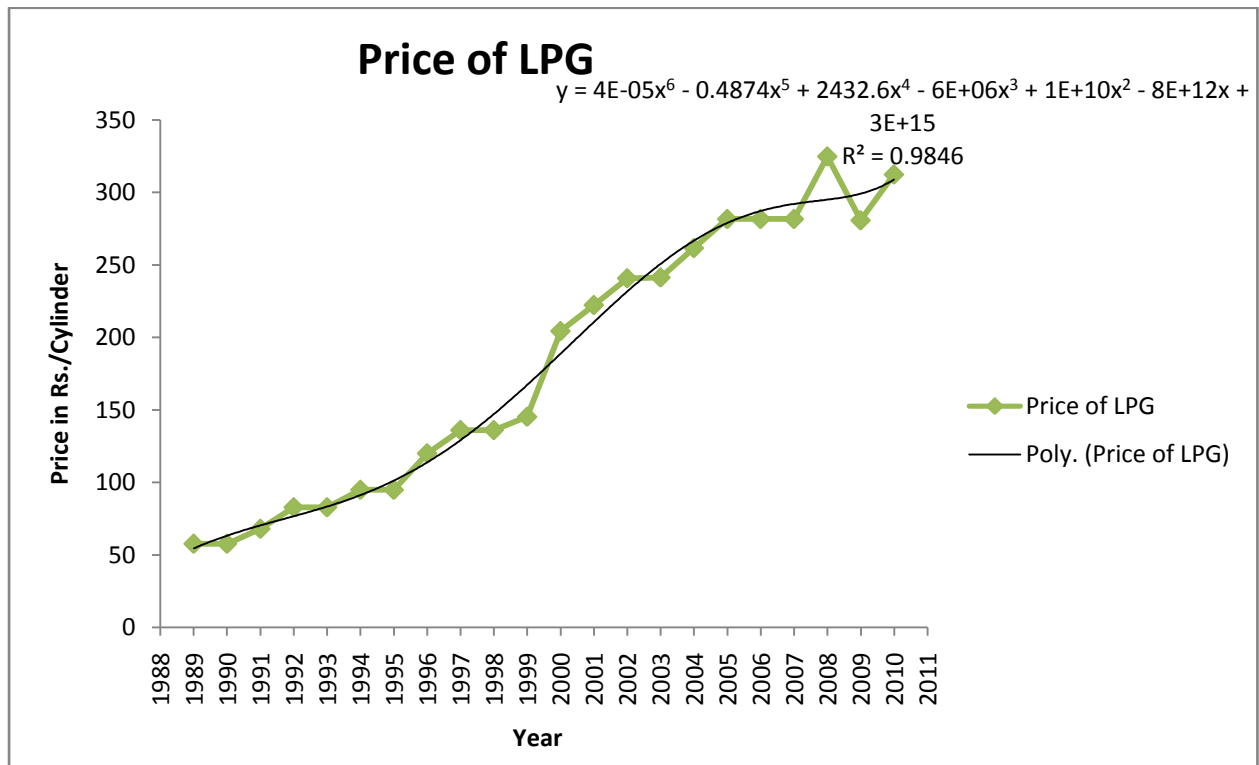
2. Exponential-



3. Linear-



4. Polynomial of degree six-



EQUATIONS:

Hence we obtain the equations of the fit from the graphs as follows:

Trend Line	Equation
Exponential	$y = 1\text{E-}75e^{0.088x}$
Linear	$y = 13.98x - 27782$
Logarithmic	$y = 27964\ln(x) - 21235$
Polynomial	$y = 4\text{E-}05x^6 - 0.487x^5 + 2432.x^4 - 6\text{E}+06x^3 + 1\text{E}+10x^2 - 8\text{E}+12x + 3\text{E}+15$

ESTIMATES:

Using these equations the LPG prices were estimated as follows:

Year	Price of LPG	Estimates			
		Linear	Exponential	Logarithmic	Polynomial
1989	57.6	35.33592	53.54935556	191162.4099	4.81638E+15
1990	57.6	49.32144	58.47526021	191176.4657	4.82286E+15
1991	67.9	63.30696	63.85428958	191190.5144	4.82935E+15
1992	82.75	77.29248	69.72812576	191204.5561	4.83584E+15
1993	82.75	91.27799	76.1422851	191218.5907	4.84234E+15
1994	94.8	105.2635	83.14647091	191232.6183	4.84884E+15
1995	94.8	119.249	90.79495862	191246.6388	4.85536E+15
1996	119.95	133.2345	99.14701635	191260.6524	4.86187E+15
1997	136	147.2201	108.2673642	191274.6589	4.8684E+15
1998	136	161.2056	118.2266757	191288.6584	4.87492E+15
1999	145.2	175.1911	129.1021256	191302.6509	4.88146E+15
2000	204.3083	189.1766	140.9779876	191316.6364	4.888E+15
2001	222.25	203.1621	153.946288	191330.6149	4.89455E+15
2002	240.7231	217.1476	168.1075179	191344.5864	4.9011E+15
2003	241.28	231.1332	183.5714128	191358.551	4.90766E+15
2004	261.6	245.1187	200.4578023	191372.5085	4.91422E+15
2005	281.6	259.1042	218.8975391	191386.4591	4.9208E+15
2006	281.6	273.0897	239.0335126	191400.4028	4.92737E+15
2007	281.6	287.0752	261.0217565	191414.3395	4.93396E+15
2008	324.7333	301.0607	285.0326577	191428.2693	4.94054E+15
2009	280.7	315.0463	311.2522766	191442.1921	4.94714E+15
2010	312.3	329.0318	339.883789	191456.108	4.95374E+15

Here we observe that prices given by the Polynomial and Linear are not realistic hence we do not consider it.

SUM OF SQUARES OF RESIDUALS (SSR):

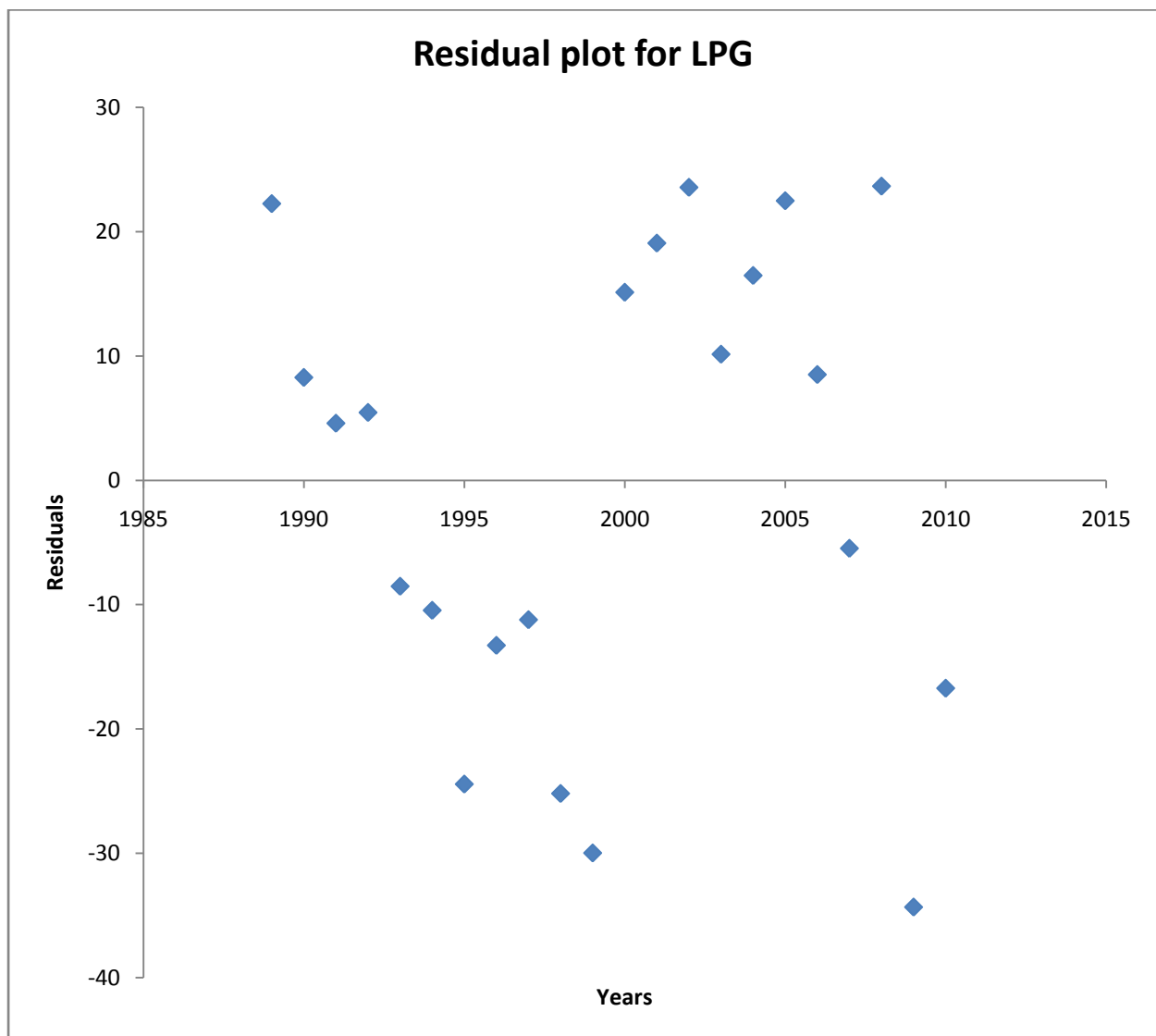
For the other fits we calculate the residuals and their SSR.
The one with least SSR is considered as the best fit.

Year	Price of LPG	Linear		Exponential	
		Residual	SSR	Residual	SSR
1989	57.6	22.26408	495.689	4.050644	16.40772
1990	57.6	8.278558	68.53453	-0.87526	0.76608
1991	67.9	4.593041	21.09603	4.04571	16.36777
1992	82.75	5.457525	29.78457	13.02187	169.5692
1993	82.75	-8.52799	72.72665	6.607715	43.6619
1994	94.8	-10.4635	109.485	11.65353	135.8047
1995	94.8	-24.449	597.7549	4.005041	16.04036
1996	119.95	-13.2845	176.4791	20.80298	432.7641
1997	136	-11.2201	125.8897	27.73264	769.0991
1998	136	-25.2056	635.3211	17.77332	315.8911
1999	145.2	-29.9911	899.4657	16.09787	259.1416
2000	204.3083333	15.13172	228.969	63.33035	4010.733
2001	222.25	19.08787	364.3469	68.30371	4665.397
2002	240.7230769	23.57543	555.801	72.61556	5273.019
2003	241.28	10.14684	102.9583	57.70859	3330.281
2004	261.6	16.48132	271.634	61.1422	3738.368
2005	281.6	22.49581	506.0613	62.70246	3931.599
2006	281.6	8.510288	72.42501	42.56649	1811.906
2007	281.6	-5.47523	29.97813	20.57824	423.4641
2008	324.7333333	23.67259	560.3914	39.70068	1576.144
2009	280.7	-34.3463	1179.666	-30.5523	933.4416
2010	312.3	-16.7318	279.9524	-27.5838	760.8654
	Total-		7384.41		32630.73

The SSR of Linear is the minimum. Hence Linear fit is the best fit for the LPG prices.

RESIDUAL PLOT:

After fitting the appropriate Trend Line to the oil prices, the prices were estimated for the collected data using the equations. Then the residuals were obtained as the difference of the observed prices and the estimated prices. Then a graph of the year and the corresponding residual were plotted to obtain a Residual Plot to study the goodness of the fit.



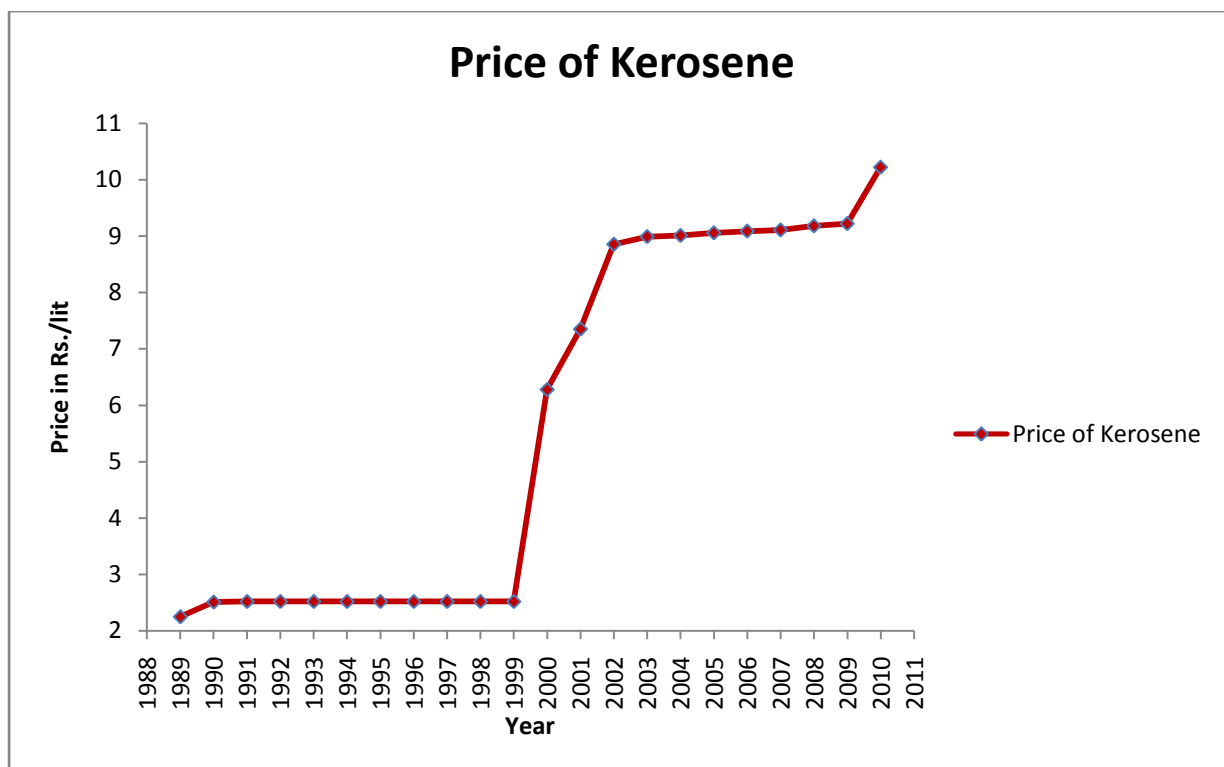
CONCLUSION:

From the graphs we can conclude that the Linear fit for LPG is good as the Residual Plot does not show any pattern.

KEROSENE:

DATA:

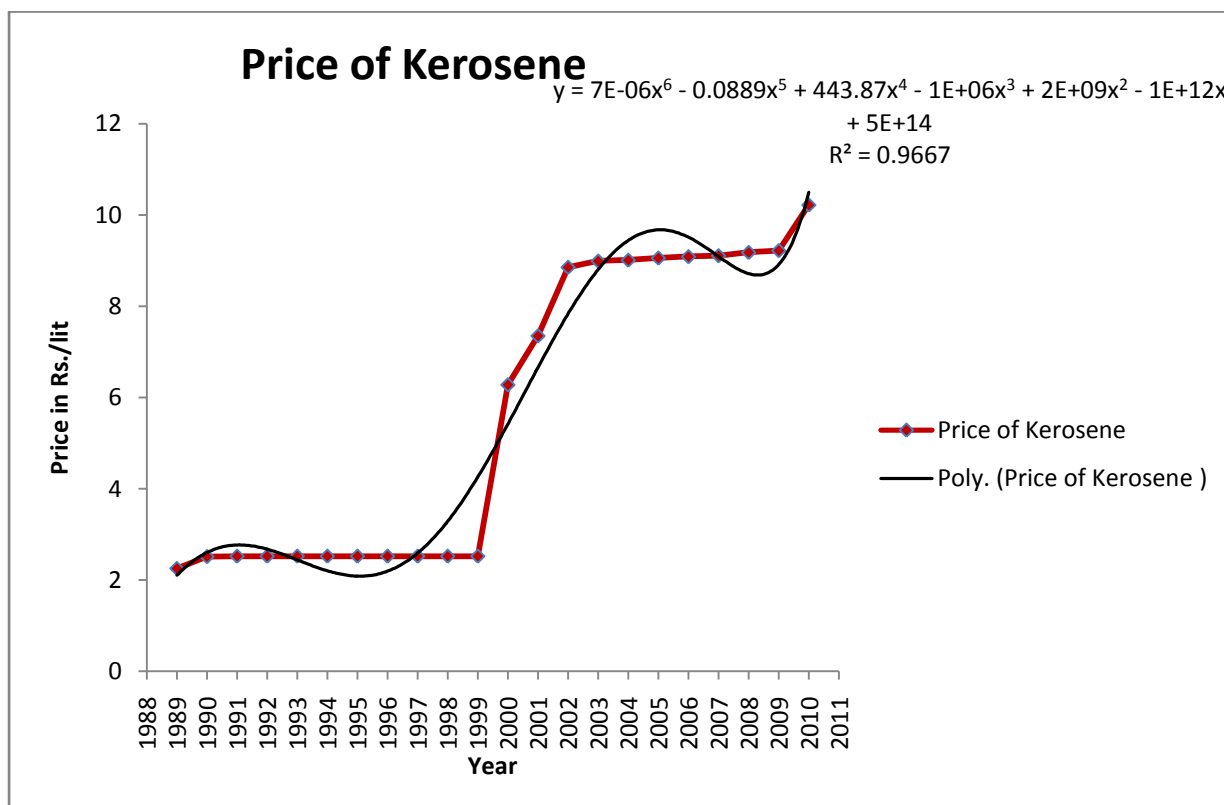
Year	Price of Kerosene
1989	2.25
1990	2.51
1991	2.52
1992	2.52
1993	2.52
1994	2.52
1995	2.52
1996	2.52
1997	2.52
1998	2.52
1999	2.52
2000	6.278333
2001	7.35
2002	8.854615
2003	8.99
2004	9.01
2005	9.056667
2006	9.0875
2007	9.1075
2008	9.18
2009	9.22
2010	10.22



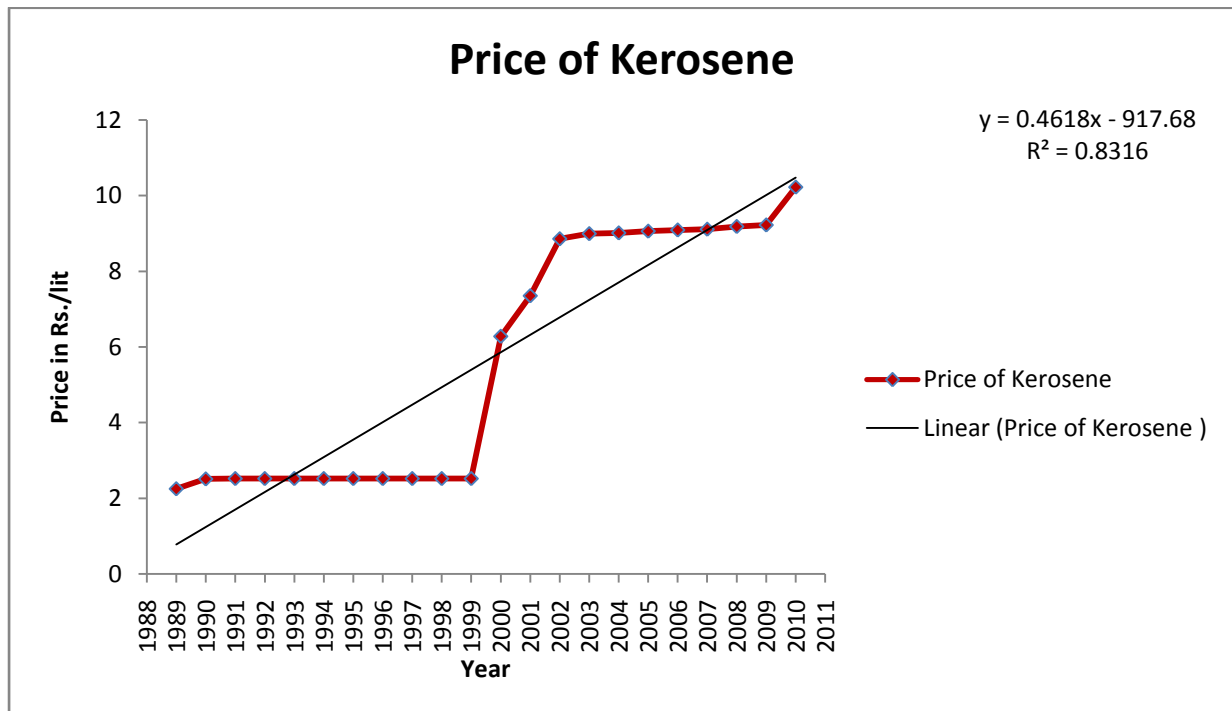
TREND LINE:

We fit the following Trend Lines to the above data:

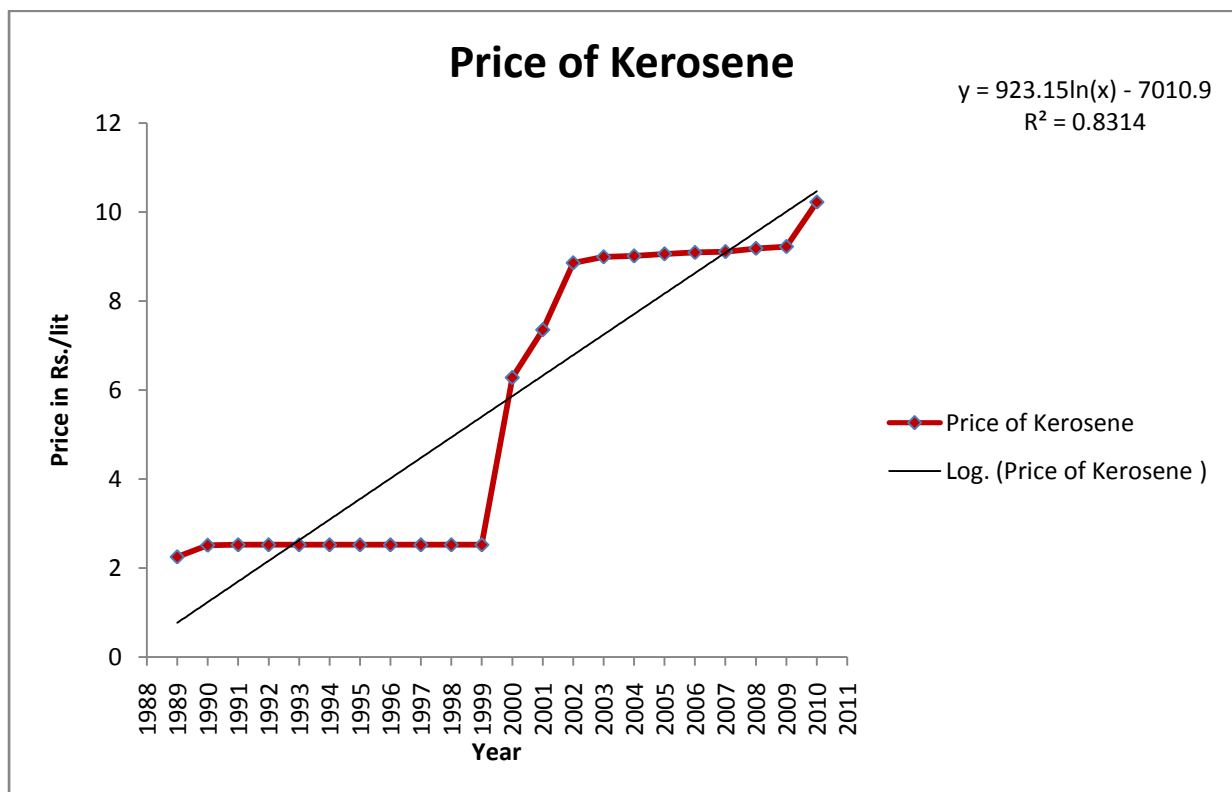
1. Polynomial of degree six-



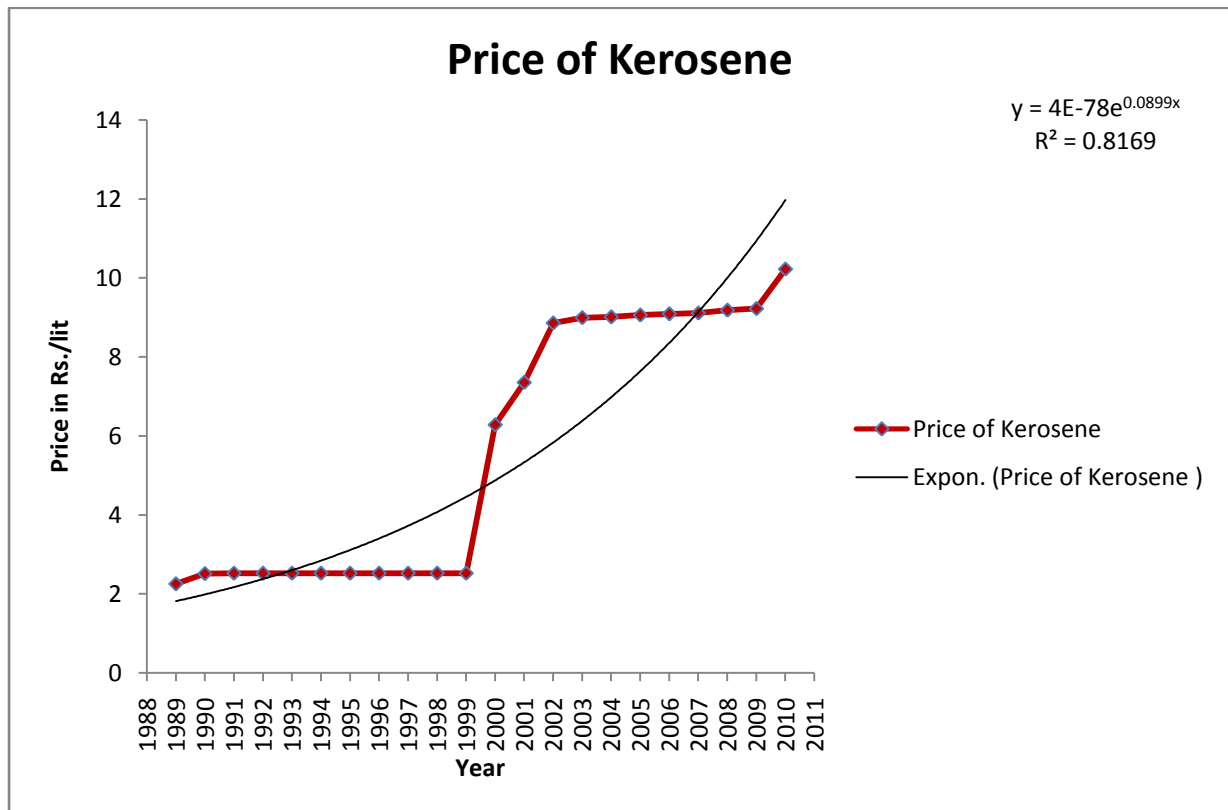
2. Linear-



3. Logarithmic-



4. Exponential-



EQUATIONS:

Hence we obtain the equations of the fit from the graphs as follows:

Trend Line	Equation
Exponential	$y = 4E-78e^{0.089x}$
Linear	$y = 0.461x - 917.6$
Logarithmic	$y = 923.1\ln(x) - 7010$
Polynomial	$y = 7E-06x^6 - 0.088x^5 + 443.8x^4 - 1E+06x^3 + 2E+09x^2 - 1E+12x + 5E+14$

ESTIMATES:

Using these equations the Kerosene prices were estimated as follows:

Year	Price of Kerosene	Estimates			
		Linear	Logarithmic	Exponential	Polynomial
1989	2.25	1.30199711	0.778442536	0.302866863	3.1944E+15
1990	2.51	1.765983046	1.240212577	0.33105789	3.1979E+15
1991	2.52	2.229735882	1.701982619	0.361872954	3.2014E+15
1992	2.52	2.693255851	2.16375266	0.395556302	3.2048E+15
1993	2.52	3.156543188	2.625522702	0.432374916	3.2083E+15
1994	2.52	3.619598127	3.087292743	0.472620629	3.2118E+15
1995	2.52	4.082420899	3.549062785	0.516612436	3.2153E+15
1996	2.52	4.545011738	4.010832827	0.564699027	3.2188E+15
1997	2.52	5.007370876	4.472602868	0.617261546	3.2223E+15
1998	2.52	5.469498545	4.93437291	0.674716616	3.2258E+15
1999	2.52	5.931394977	5.396142951	0.737519637	3.2293E+15
2000	6.278333	6.393060403	5.857912993	0.806168401	3.2328E+15
2001	7.35	6.854495054	6.319683034	0.881207032	3.2363E+15
2002	8.854615	7.315699161	6.781453076	0.963230303	3.2398E+15
2003	8.99	7.7766729	7.2432231	1.0528883	3.2433E+15

3		53	17	49	15
200 4	9.01	8.2374166 61	7.7049931 59	1.1508918 18	3.2468E+ 15
200 5	9.05666 7	8.6979305 15	8.1667632 01	1.2580175 09	3.2504E+ 15
200 6	9.0875	9.1582147 43	8.6285332 42	1.3751145 22	3.2539E+ 15
200 7	9.1075	9.6182695 74	9.0903032 84	1.5031109 94	3.2574E+ 15
200 8	9.18	10.078095 24	9.5520733 25	1.6430214 53	3.2609E+ 15
200 9	9.22	10.537691 96	10.013843 37	1.7959548 61	3.2645E+ 15
201 0	10.22	10.997059 97	10.475613 41	1.9631234 01	3.268E+1 5

Here we observe that prices given by the Polynomial is not realistic hence we do not consider it.

SUM OF SQUARES OF RESIDUALS (SSR):

For the other fits we calculate the residuals and their SSR. The one with least SSR is considered as the best fit.

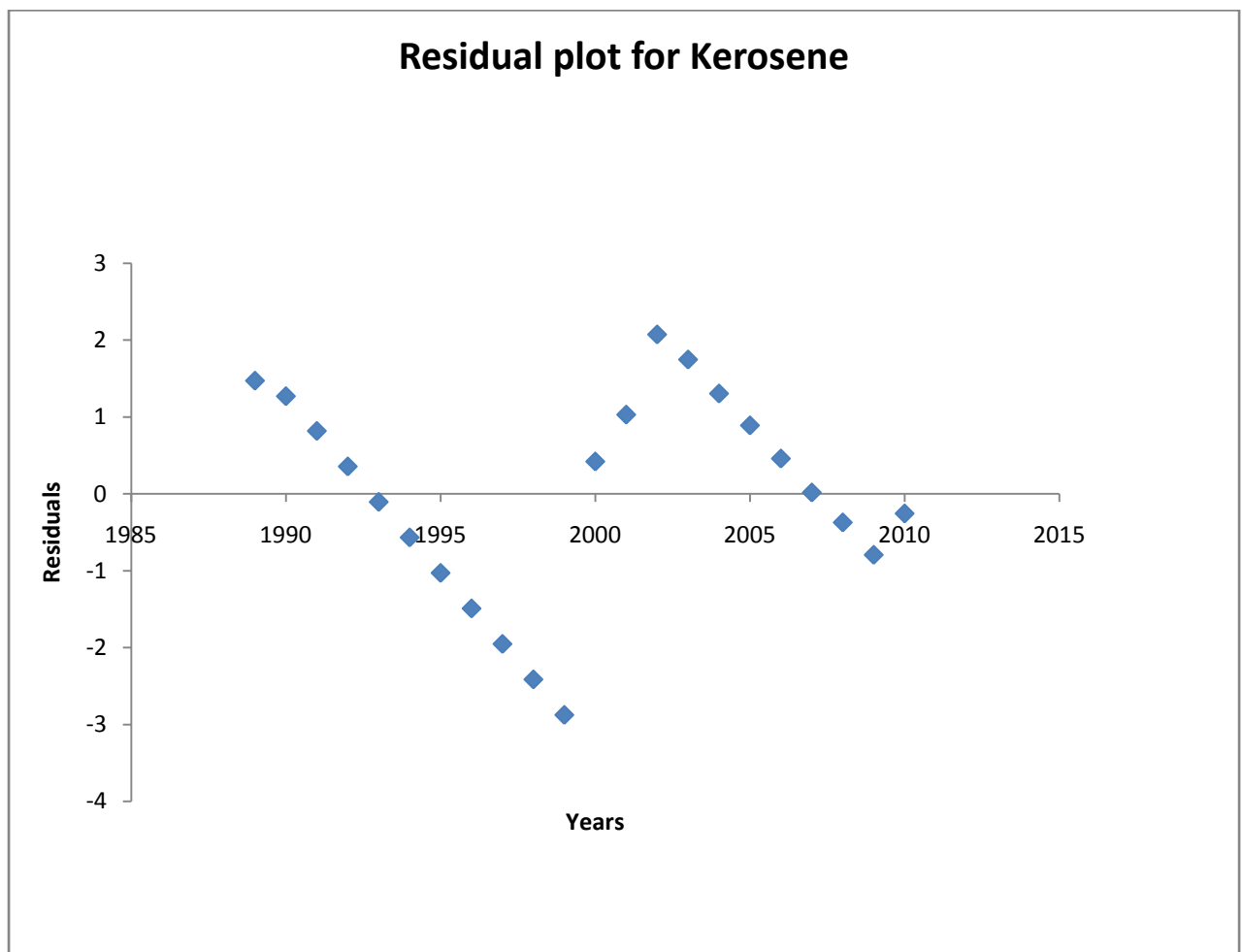
Year	Price of Kerosene	Linear		Logarithmic		Exponential	
		Residual	SSR	Residual	SSR	Residual	SSR
1989	2.25	0.948003	0.898709479	1.471557	2.165481	0.302867	1.947133137
1990	2.51	0.744017	0.553561228	1.269787	1.61236	0.331058	2.17894211
1991	2.52	0.290264	0.084253258	0.818017	0.669152	0.361873	2.158127046
1992	2.52	-0.17326	0.03001759	0.356247	0.126912	0.395556	2.124443698
1993	2.52	-0.63654	0.405187231	0.10552	0.011135	0.432375	2.087625084
1994	2.52	-1.0996	1.20911604	0.56729	0.321821	0.472621	2.047379371
1995	2.52	-1.56242	2.441159065	1.02906	1.05897	0.516612	2.003387564
1996	2.52	-2.02501	4.100672539	1.49083	2.222583	0.564699	1.955300973
1997	2.52	-2.48737	6.187013875	1.9526	3.812658	0.617262	1.902738454
1998	2.52	-2.9495	8.699541669	2.41437	5.829197	0.674717	1.845283384
1999	2.52	-	11.6376	-	8.2721	0.737	1.78248

99		3.411 39	1569	2.876 14	98	52	0363
20 00	6.278 333	- 0.114 73	0.01316 2301	0.420 42	0.1767 53	0.806 168	5.47216 4933
20 01	7.35	0.495 505	0.24552 5151	1.030 317	1.0615 53	0.881 207	6.46879 2968
20 02	8.854 615	1.538 916	2.36826 3144	2.073 162	4.2980 02	0.963 23	7.89138 5082
20 03	8.99	1.213 327	1.47216 2523	1.746 777	3.0512 29	1.052 888	7.93711 1651
20 04	9.01	0.772 583	0.59688 5015	1.305 007	1.7030 43	1.150 892	7.85910 8182
20 05	9.056 667	0.358 736	0.12869 1627	0.889 903	0.7919 28	1.258 018	7.79864 9158
20 06	9.087 5	- 0.070 71	0.00500 0575	0.458 967	0.2106 5	1.375 115	7.71238 5478
20 07	9.107 5	- 0.510 77	0.26088 5558	0.017 197	0.0002 96	1.503 111	7.60438 9006
20 08	9.18	- 0.898 1	0.80657 5055	- 0.372 07	0.1384 39	1.643 021	7.53697 8547
20 09	9.22	- 1.317 69	1.73631 2103	- 0.793 84	0.6301 87	1.795 955	7.42404 5139
20 10	10.22	- 0.777 06	0.60382 22	- 0.255 61	0.0653 38	1.963 123	8.25687 6599
	Total		44.4841 3292		38.22 989		103.994 7279

The SSR of Logarithmic is the minimum. Hence Logarithmic fit is the best fit for the Kerosene prices.

RESIDUAL PLOT:

After fitting the appropriate Trend Line to the oil prices, the prices were estimated for the collected data using the equations. Then the residuals were obtained as the difference of the observed prices and the estimated prices. Then a graph of the year and the corresponding residual were plotted to obtain a Residual Plot to study the goodness of the fit.



CONCLUSION:

The Residual Plot of Diesel shows a certain pattern, which means that the fit is not very good. But due to the limitations of Microsoft Excel, other fits, which could be better, cannot be used for the estimation. Hence we proceed with Logarithmic Fit as the best fit for Kerosene Prices.

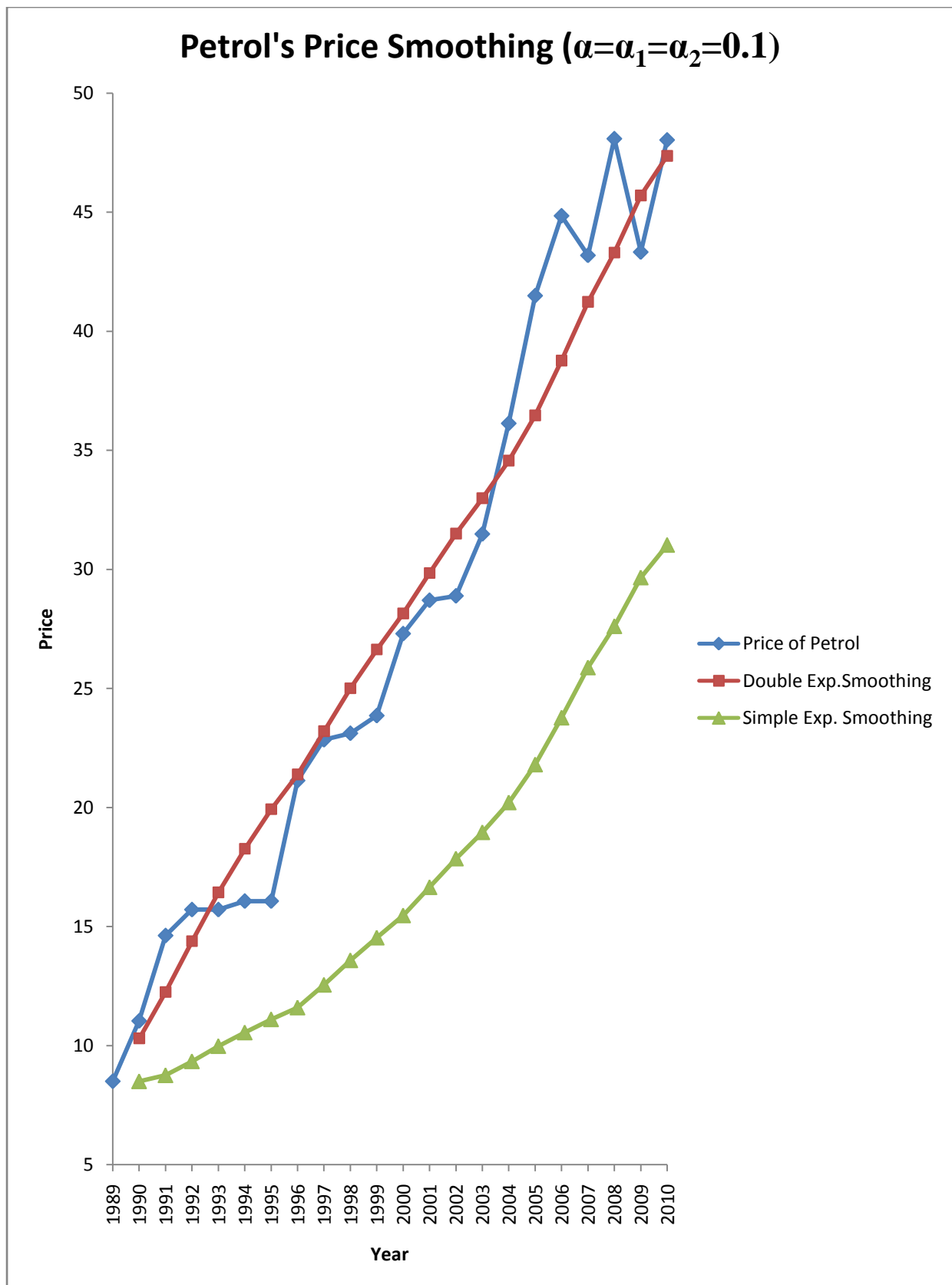
DOUBLE EXPONENTIAL SMOOTHING

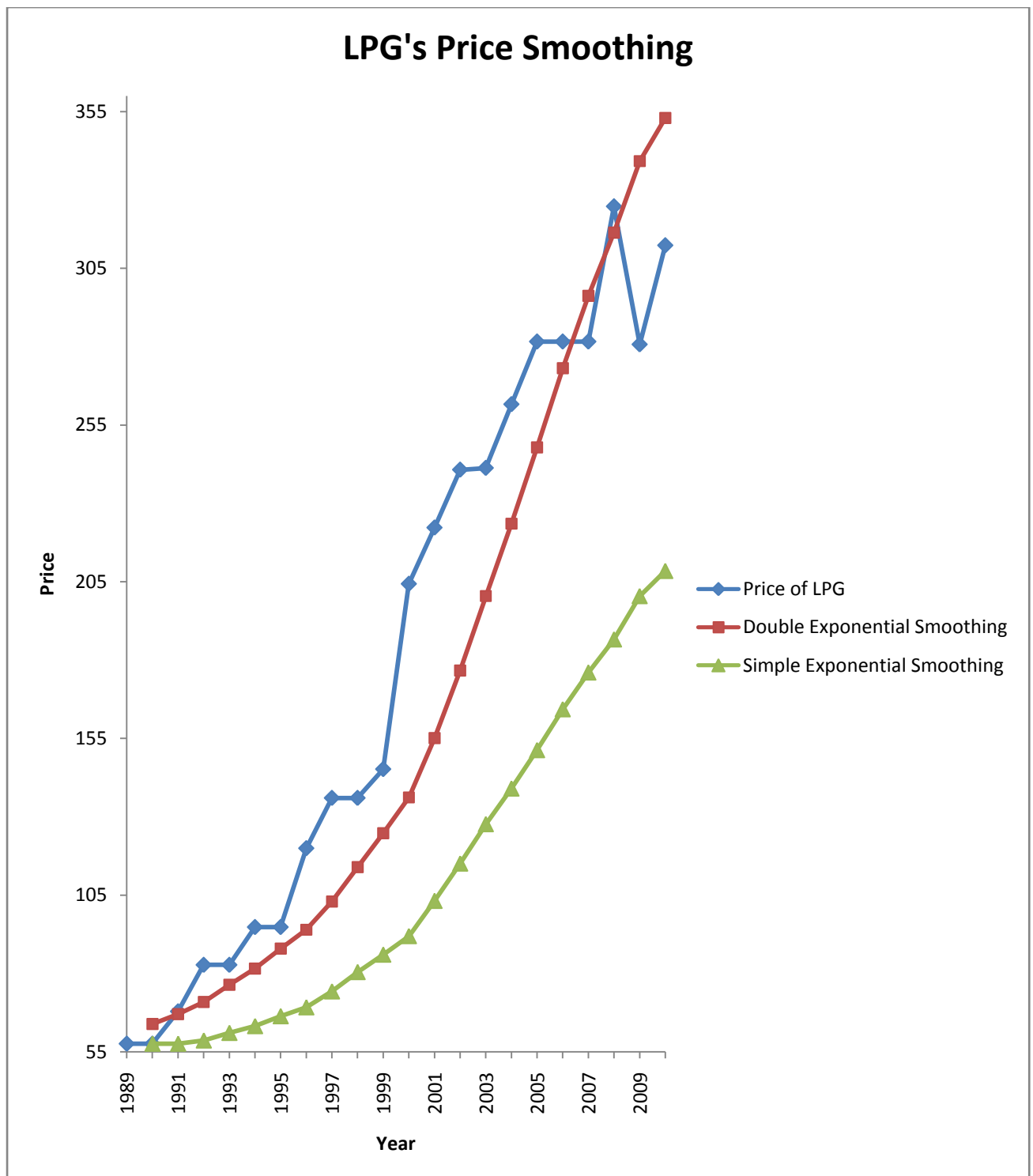
The Prices of Petrol & LPG have a Linear increasing trend & no seasonal variation, hence Double Exponential Smoothing Procedure can be applied to Petrol and LPG's Prices to produce smoothed values of Prices. Simple Exponential Smoothing (SES) requires time series to be stagnant; therefore Double Exponential Smoothing (DES) is applied below.

For both the time series mentioned above the value of smoothing constant is 0.1 for Simple Exponential Smoothing. For Double Exponential Smoothing $\alpha_1=\alpha_2=0.1$ for Petrol's Price Smoothing & $\alpha_1=0.1$, $\alpha_2=0.4$ for LPG's Price Smoothing.

Year	Price of Petrol	DES	SES	Price of LPG	DES	SES
1989	8.5	—	—	57.6	—	—
1990	11.035	10.3025	8.5	57.6	63.8875	57.6
1991	14.62	12.2515	8.7535	67.9	67.03125	57.6
1992	15.71	14.387785	9.34015	82.75	70.925375	58.63
1993	15.71	16.43266365	9.977135	82.75	76.3880725	61.042
1994	16.0666667	18.2658278	10.5504215	94.8	81.5589774	63.2128
1995	16.0666667	19.92935059	11.10204602	94.8	87.9474326	66.37152
1996	21.13	21.38789426	11.59850808	119.95	93.9711451	69.214368

1997	22.84	23.204337 95	12.55165727	136	102.946 64	74.2879312
1998	23.115	25.006493 9	13.58049155	136	113.951 721	80.4591380 8
1999	23.856	26.637019 31	14.53394239	145.2	124.738 224	86.0132242 7
2000	27.2983 333	28.150781 99	15.46614815	204.30 83	136.184 548	91.9319018 4
2001	28.7	29.848877 24	16.64936667	222.25	155.122 025	103.169545
2002	28.8838 462	31.505840 87	17.85443	240.72 31	176.645 039	115.077590 5
2003	31.4826 667	32.989272 8	18.95737162	241.28	200.426 181	127.642139 1
2004	36.1283 333	34.569177 53	20.20990112	261.6	223.519 054	139.005925 2
2005	41.49	36.471250 01	21.80174434	281.6	247.857 878	151.265332 7
2006	44.845	38.769469 41	23.77056991	281.6	273.112 504	164.298799 4
2007	43.185	41.234122 17	25.87801292	281.6	296.181 167	176.028919 5
2008	48.0833 333	43.305818 44	27.60871163	324.73 33	316.359 718	186.586027 5
2009	43.3233 333	45.707953 56	29.6561738	280.7	339.168 691	200.400758 1
2010	48.03	47.370028 97	31.02288975	312.3	352.954 686	208.430682 3





CONCLUSION:

It can be seen that Single Exponential Smoothing does not follow the data well when there is a trend in the data hence this drawback of Simple Exponential Smoothing is overcome by Double Exponential Smoothing.

ESTIMATION OF FUEL PRICES

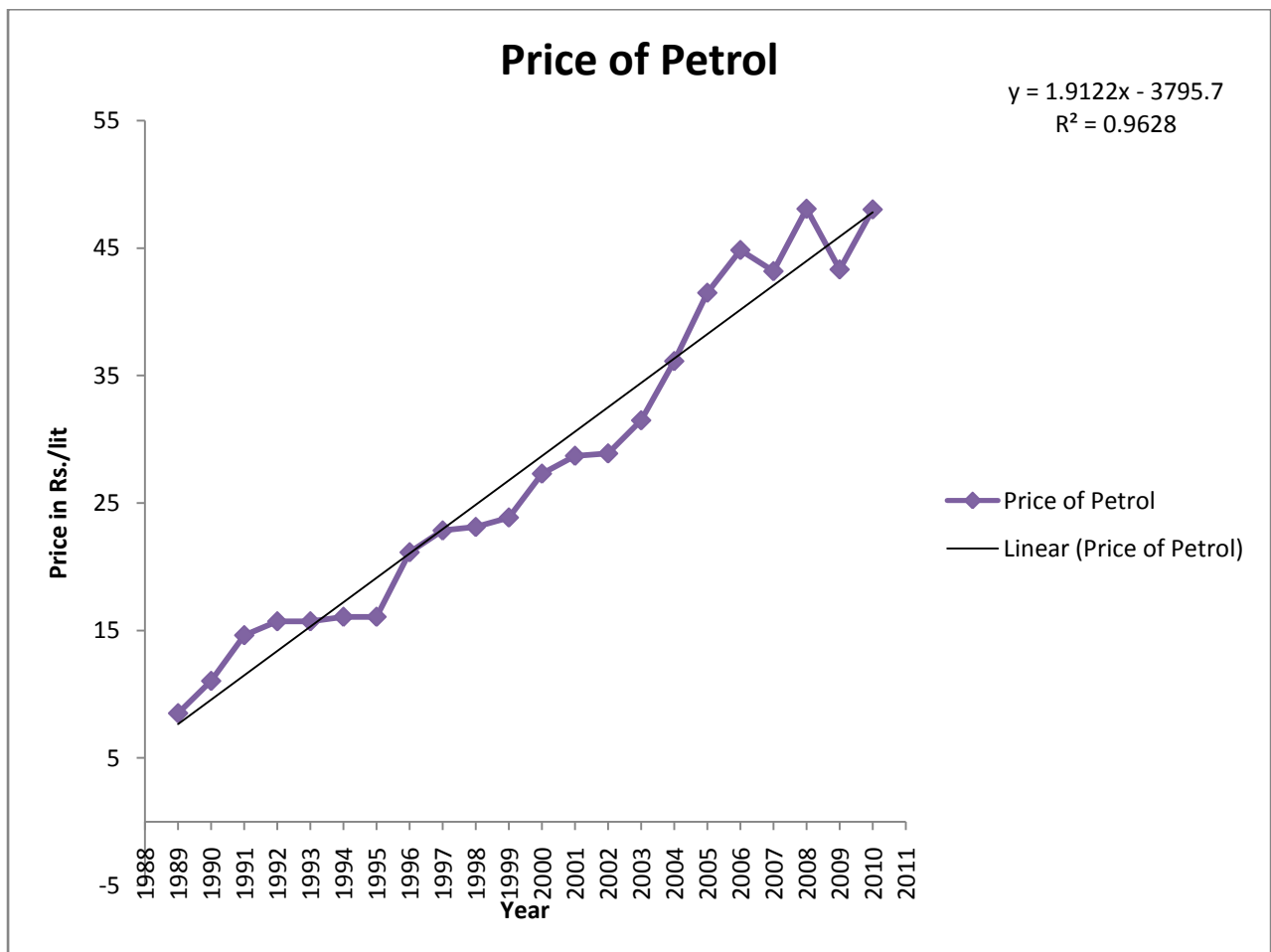
For estimation purpose we select an equation which gives a realistic value of price of the fuel. Selecting an equation which has a good R^2 value does not always give realistic value of the price. Hence we choose the following equations for the following fuels:

<i>FUEL</i>	<i>Equation Used</i>
Petrol	Linear
Diesel	Logarithmic
LPG	Linear
Kerosene	Logarithmic

PETROL

For estimation of Petrol prices in the following years after 2011 we fit a Linear function to the data of Petrol prices.

We choose this fit because it gave realistic values of prices and its Sum of Squares of Residual is Minimum for this fit. Hence we get the following graph:



The equation of this Exponential function is

$$y = 1.912x - 3795$$

Till 2011 the observed Petrol price and the estimated prices are as follows

Year	Observed Prices	Estimated Price
1989	8.5	7.6658
1990	11.035	9.578
1991	14.62	11.4902
1992	15.71	13.4024
1993	15.71	15.3146
1994	16.06666667	17.2268
1995	16.06666667	19.139
1996	21.13	21.0512
1997	22.84	22.9634
1998	23.115	24.8756
1999	23.856	26.7878
2000	27.29833333	28.7
2001	28.7	30.6122
2002	28.88384615	32.5244
2003	31.48266667	34.4366
2004	36.12833333	36.3488
2005	41.49	38.261
2006	44.845	40.1732

2007	43.185	42.0854
2008	48.08333333	43.9976
2009	43.32333333	45.9098
2010	48.03	47.822

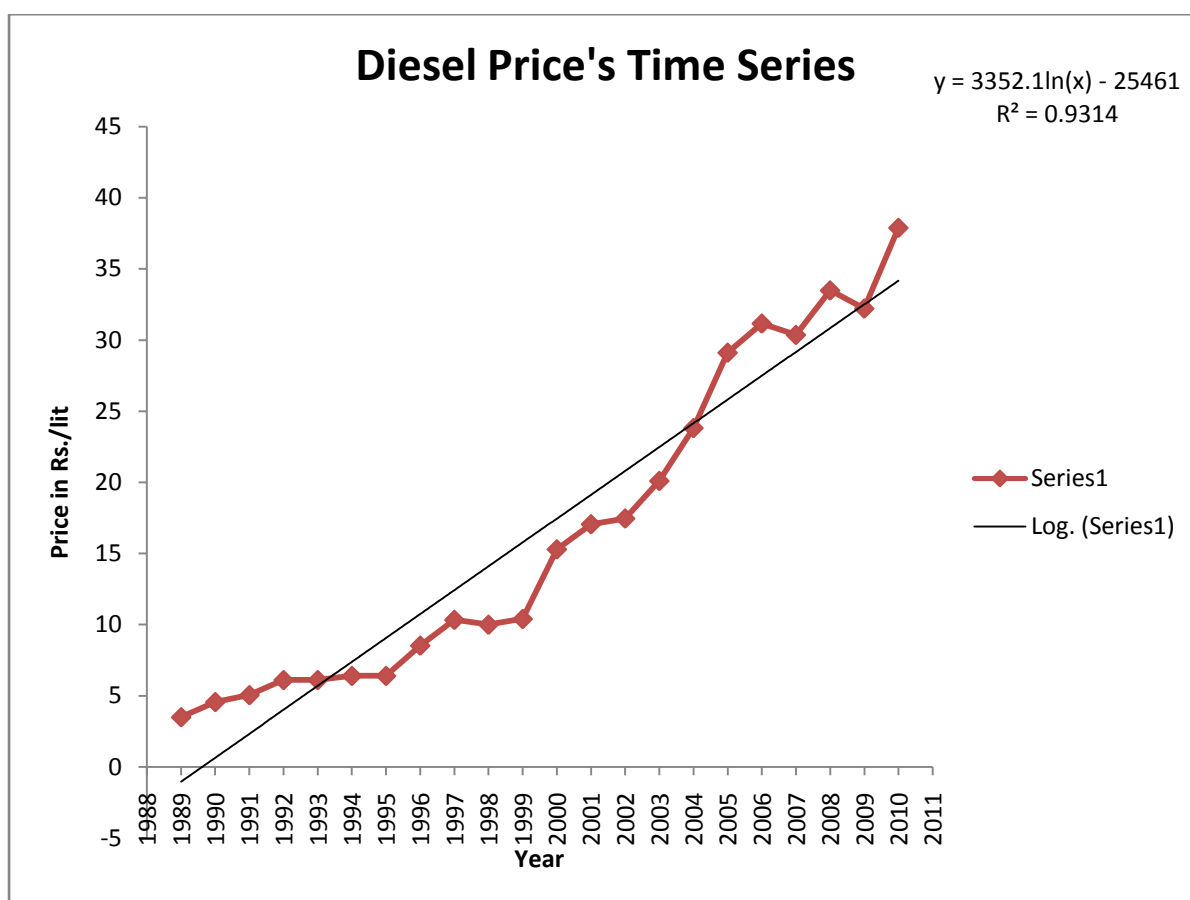
For the successive years the Petrol prices are predicted to be as follows:

<i>Year</i>	<i>Estimated price (in Rs.)</i>
2011	49.7342
2012	51.6464
2013	53.5586
2014	55.4708
2015	57.383
2016	59.2952
2017	61.2074
2018	63.1196
2019	65.0318
2020	66.944

DIESEL:

For estimation of Diesel prices in the following years after 2011 we fit a Logarithmic function to the data of Diesel prices.

We choose this fit because it gave realistic values of prices and its Sum of Squares of Residual is Minimum for this fit. Hence we get the following graph:



The equation of this Linear function is

$$y = 3352.\ln(x) - 25461$$

Till 2011 the observed Diesel price and the estimated prices are as follows:

Year	Observed prices	Estimated prices (in Rs.)
1989	3.5	-1.26184
1990	4.565	0.42300
1991	5.05	2.107003
1992	6.11	3.790157
1993	6.11	5.472465
1994	6.4	7.15393
1995	6.4	8.834552
1996	8.53	10.51433
1997	10.34	12.19327
1998	10	13.87105
1999	10.4	15.54863
2000	15.295	17.22504
2001	17.06	18.90063
2002	17.46384615	20.57537
2003	20.10266667	22.24928
2004	23.82166667	23.92235
2005	29.11666667	25.59459

2006	31.165	27.26599
2007	30.365	28.93656
2008	33.49	30.6063
2009	32.21666667	32.27521
2010	37.89	33.94328

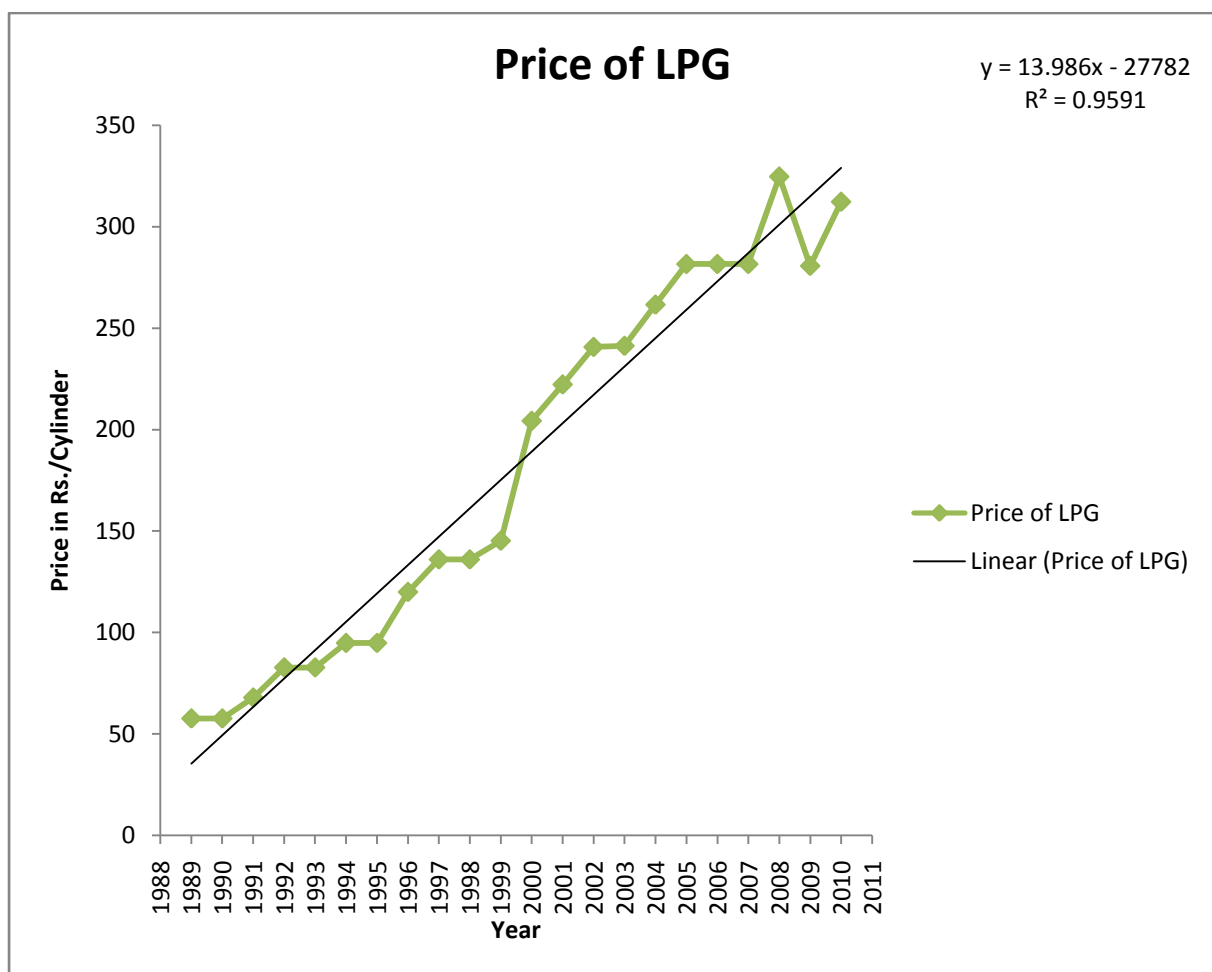
For the successive years the Diesel prices are predicted to be as follows:

<i>Year</i>	<i>Estimated price (in Rs.)</i>
<i>2011</i>	<i>35.61053052</i>
<i>2012</i>	<i>37.27694865</i>
<i>2013</i>	<i>38.94253874</i>
<i>2014</i>	<i>40.60730163</i>
<i>2015</i>	<i>42.27123812</i>
<i>2016</i>	<i>43.93434905</i>
<i>2017</i>	<i>45.59663522</i>
<i>2018</i>	<i>47.25809746</i>
<i>2019</i>	<i>48.91873658</i>
<i>2020</i>	<i>50.5785534</i>

LPG:

For estimation of LPG prices in the following years after 2011 we fit a Linear function to the data of LPG prices.

We choose this fit because it gave realistic values of prices and its Sum of Squares of Residual is Minimum for this fit. Hence we get the following graph:



The equation of this Linear function is

$$y = 13.98x - 27782$$

Till 2011 the observed LPG price and the estimated prices are as follows:

Year	Observed Prices	Estimate Prices
1989	57.6	36.154
1990	57.6	50.14
1991	67.9	64.126
1992	82.75	78.112
1993	82.75	92.098
1994	94.8	106.084
1995	94.8	120.07
1996	119.95	134.056
1997	136	148.042
1998	136	162.028
1999	145.2	176.014
2000	204.3083333	190
2001	222.25	203.986
2002	240.7230769	217.972
2003	241.28	231.958
2004	261.6	245.944
2005	281.6	259.93
2006	281.6	273.916

2007	281.6	287.902
2008	324.7333333	301.888
2009	280.7	315.874
2010	312.3	329.86

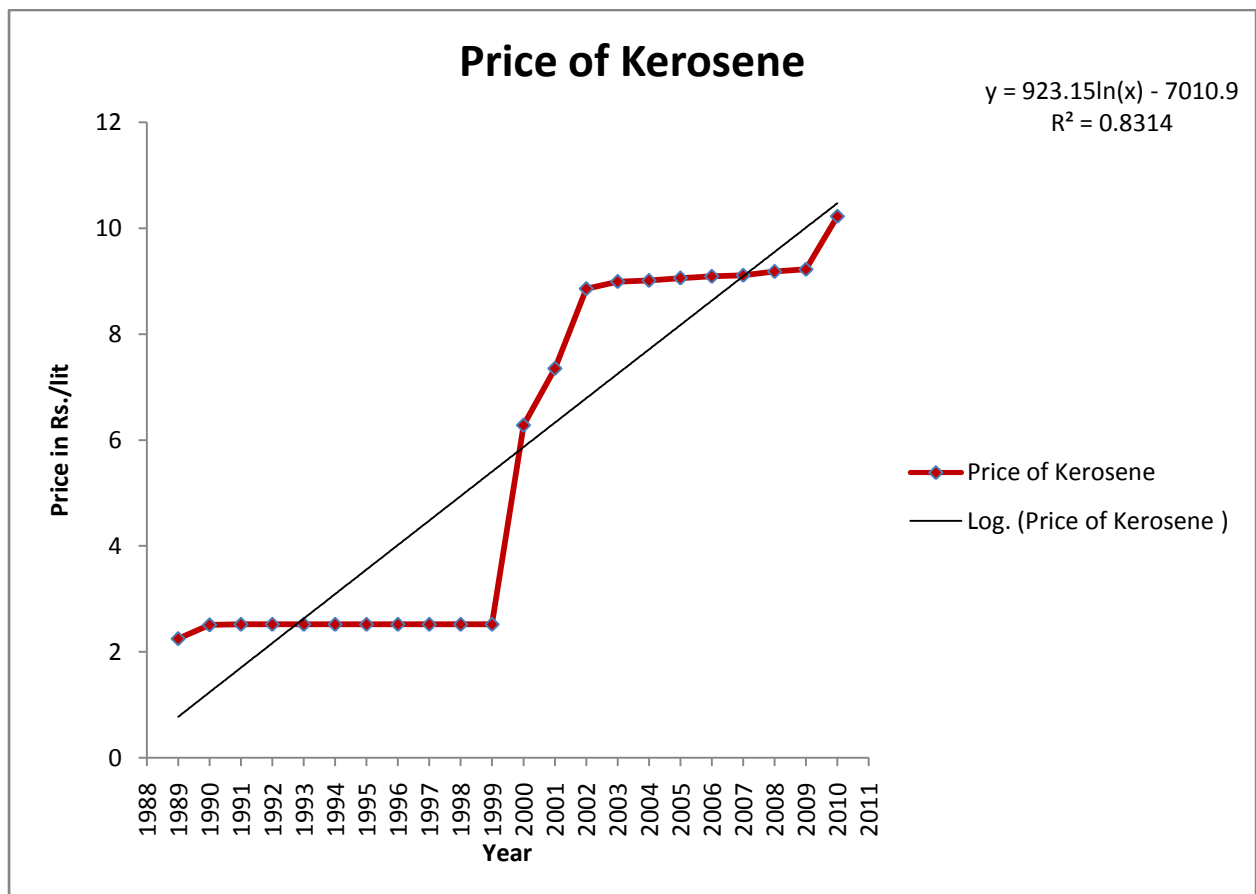
For the successive years the LPG prices are predicted to be as follows:

<i>Year</i>	<i>Estimated price (in Rs.)</i>
2011	343.846
2012	357.832
2013	371.818
2014	385.804
2015	399.79
2016	413.776
2017	427.762
2018	441.748
2019	455.734
2020	469.72

KEROSENE:

For estimation of Kerosene prices in the following years after 2011 we fit a Logarithmic function to the data of Kerosene prices.

We choose this fit because it gave realistic values of prices and its Sum of Squares of Residual is Minimum for this fit. Hence we get the following graph:



The equation of this Logarithmic function is

$$y = 923.1\ln(x) - 7010$$

Till 2011 the observed Kerosene price and the estimated prices are as follows:

Year	Observed Prices	Estimate Prices
1989	2.25	0.781766474
1990	2.51	1.245777542
1991	2.52	1.709555497
1992	2.52	2.173100573
1993	2.52	2.636413004
1994	2.52	3.099493024
1995	2.52	3.562340865
1996	2.52	4.024956761
1997	2.52	4.487340943
1998	2.52	4.949493643
1999	2.52	5.411415094
2000	6.278333333	5.873105526
2001	7.35	6.334565171
2002	8.854615385	6.795794259
2003	8.99	7.25679302
2004	9.01	7.717561684
2005	9.056666667	8.178100482
2006	9.0875	8.638409641

2007	9.1075	9.098489391
2008	9.18	9.558339961
2009	9.22	10.01796158
2010	10.22	10.47735447

For the successive years the Kerosene prices are predicted to
be as follows:

<i>Year</i>	<i>Estimated price (in Rs.)</i>
2011	10.93651887
2012	11.395455
2013	11.85416308
2014	12.31264335
2015	12.77089602
2016	13.22892134
2017	13.68671951
2018	14.14429077
2019	14.60163535
2020	15.05875345

CONCLUSION

CONCLUSION

All the Oil prices show an increasing trend.

Based on the increasing pattern, we observe that the Petrol and LPG prices show a Linear trend while Diesel and Kerosene show a Logarithmic trend.

Also as prices of Petrol and LPG have a Linear increase and no seasonal variation, Double Exponential Smoothing can be applied to them to obtain smoothed values of prices.

Using the selected fits we estimate the oil prices for the successive ten years that is from 2011 to 2020. Hence, by studying the trends in the changes in Oil Prices from 1989-2010, the oil prices for the years 2014 and 2015 are estimated to be as follows:

Fuel	Estimated Prices for 2014 (Rs.)	Estimated Prices for 2015 (Rs.)
Petrol	55.47	57.38
Diesel	40.60	42.27
LPG	385.804	399.79
Kerosene	12.31	12.77

The estimates are approximate, due to the limitations of the data set. Hence the actual prices may or may not differ from estimated values.

SCOPE AND LIMITATIONS

SCOPE

The statistical analysis of Oil Prices can be made on a larger extent by using larger data for a greater number of years. Having studied the previous patterns in changes of Oil Prices, the study enables the approximate prediction of Average Annual Oil Prices for years 2011 - 2020.

If the data set is large enough and consists of monthly, weekly or daily Oil Prices, the use of similar statistical techniques could be extended to predicting the Oil Prices for forthcoming years, on an Annual, Monthly, Weekly and / or Daily basis with Sufficient Accuracy.

LIMITATIONS

- The Residual Plots of Diesel and Kerosene show a certain pattern, which means that their respective fits are not very good. But due to the limitations of Microsoft Excel, other fits, which could be better, cannot be used for estimation.
- The trend line equations obtained in Microsoft Excel seem to be incorrect. They do not provide realistic values of Oil Prices for a given year, even for the trend line having the best R-Squared Value.
- The data represents the Oil Prices in Delhi, Only. They are not the average Oil Prices in the Nation.
- The data of Oil Prices has been obtained for the period 1989-2010 only. There were major changes in Oil Prices after 2010, which have not been studied and taken into consideration, while estimating the forecast values for Oil Prices. This may result discrepancies between the Forecasted and Observed Values.
- The Oil Prices in the data are Annual Averages. Oil Prices on a Monthly, Weekly and Daily basis could not be obtained. Hence, Prediction is limited to Annual Prices only. Oil Prices cannot be predicted on a Monthly, Weekly and Daily basis.

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