

ORIE 4820: Spreadsheet-Based Modeling and Data Analysis

Trend Estimation

Spring 2013

The purpose of this lab exercise is to give your project team an opportunity to *design a set of demand projection methods* that will be of use to a Liquair-Pro salesperson in determining a price quote for an *existing customer*.

The template file for this exercise is *trend-estimation.xlsm*. Download a copy of the file from the course Blackboard site and *save it on your computer*.

Topics/Tools used to construct the template file:

- *Parsing text* and *converting text to numbers*: **right, value**
- *Using array functions* to select values from a restricted set: **if**
- *Transforming data* for use in regression models: **ln**
- *Performing dynamic linear regression*: **linest**
- *Evaluating the goodness of fit* of simple regression models: **rsq, steyx**
- *Using regression model output to make projections*

Background:

The workbook *trend-estimation.xlsm* consists of a single worksheet that contains customer information and up to seven years' worth of historical quarterly demand data for five existing Liquair-Pro customers (G – K), as well as quarterly data for a potential new customer (L) that is thinking of switching to Liquair-Pro from its current supplier. Already completed on the worksheet is a simple regression tool that determines the *best fit linear trend line* and the *best fit exponential trend line* for the selected customer's demand data at the annual level. These trend lines are projected forward to *provide point estimate demand projections for the next five years*.

The linear and exponential regression models were constructed using the steps in Sections 1-3 below. Walk through these steps with your team to ensure you understand how the projections are being made. Then tackle the questions at the bottom of page 3.

Section 1: Summarizing and Transforming Data

Data validation has been used on cell L5 so that the user may select any of the customers from a dropdown list. In the *Selected Customer Demand* table:

- (1) Column K has already been populated with the corresponding column of quarterly demand data using the **vlookup** and **match** functions. Note that a zero (0) will result if the source data element is blank.
- (2) *Populate the Year column* (column L) by *extracting the rightmost four characters* from the corresponding cell in column J using the **right** function and then *converting the result to a numeric value* using the **value** function. For instance, cell L16 should contain “=VALUE(RIGHT(J16,4))”.

- (3) **Populate the First and Last Year cells** (L8 and L9), making sure to *only consider those years where demand is greater than 0* (recall that you can do this by using **if** as an array function). Once this is achieved, populate cell L7 accordingly.

In the *Best Linear Fit* table:

- (4) Note that column P has been populated to *display only those years having nonzero demand*.
(5) Summarize the annual demands in column Q using the **sumif** function.

In the *Best Exponential Fit* table:

- (6) Note that column X has been populated to *display only those years having nonzero demand*.
(7) Summarize the annual demands in column Y using the **sumif** function.
(8) Take the natural logarithm of these values in column Z using **ln**. We can now use *simple linear regression* to get the (transformed) coefficients of the best fit exponential curve.

Section 2: Dynamic Simple Linear Regression

Using the existing infrastructure on the worksheet, we will build a dynamic regression tool that is faster and more convenient than using Excel's static regression tool.

- (1) **Regression Coefficients and Standard Error of Coefficients:** The array function **linest** uses the "least squares" method to calculate a straight line that best fits the designated *Y* and *X* data (i.e., the response and explanatory values that you give it), and it *returns an entire matrix of information*, including the regression line coefficients (i.e., the regression line slope \hat{b} and intercept \hat{a}) and their corresponding standard errors. **Linest** can be used for either simple or multiple linear regression. In the *Best Linear Fit* and *Best Exponential Fit* tables, you will use the **linest** function to populate the designated cells with the regression line coefficients and their corresponding standard errors by entering it as an *array formula*. For instance, in the *Best Linear Fit* table:

- (a) Highlight the four cells R25:S26.

- (b) Type

"=LINEST(OFFSET(\$Q\$16,0,0,\$L\$7,1),OFFSET(\$O\$16,0,0,\$L\$7,1),TRUE,TRUE)"

and press **Ctrl-Shift-Enter**. It is *essential* that you specify the third and fourth arguments of the **linest** function as **TRUE**.

Note: The **linest** function cannot handle blanks or zeroes. We are using the **offset** function here to ensure that only actual data are passed into the **linest** function. You can also find the regression line coefficients \hat{b} and \hat{a} , respectively, using the functions **slope** and **intercept**. However, you need to use the **linest** function to get the standard errors of these coefficients, which are needed if you want to construct confidence intervals for these parameters.

- (c) Do the same for the *Best Exponential Fit* table, using the log demand vector as the first input vector to the **linest** function.

- (2) **R Square and Standard Error of Estimate for the linear regression line:** Cells R28 and R29 should contain, respectively, the *R Square* value and the standard error of estimate associated with the linear regression line. You can use the Excel functions **rsq** and **steyx**, respectively, to do this. Be sure to enter the demand vector as the first vector, and time (*t*) as the second vector. Do the same for the *Best Exponential Fit* table.
- (3) **Fitted Demand and Residual columns:** Cells R16:R22 and AA16:AA22 should contain the *fitted* annual demand values corresponding to each year in the dataset (i.e., the predicted values according to the regression coefficients). Cells S16:S22 and AB16:AB22 should contain the associated residual errors. For example, in the *Best Exponential Fit* table, given the *t* and *Y_t* values in columns W and Y, columns AA and AB should contain, respectively:

$$\hat{Y}_t = e^{\hat{a}} \cdot e^{\hat{b}t}$$

$$e_t = Y_t - \hat{Y}_t$$

- (4) **Annual Demand Prediction:** Cells R34:R38 and Z34:Z38 should contain, respectively, the predicted annual demand for 2012-2016 based on the regression coefficients for the best linear fit and the best exponential fit.
- (5) **Standard Error of Estimate for the original annual demand data (*s_e*):** In cells S34 and AA34, to get the standard error of estimate associated with the annual demand data, you can use the fact that this value is (essentially) the *standard deviation of the residual errors*, except that we use *n* – 2 in the denominator (instead of *n* - 1), since the residuals are based on the estimation of 2 parameters (*a* and *b*). That is:

$$s_e = \sqrt{\frac{\sum e_i^2}{n-2}}$$

- (6) **Annual Growth Rate and its Standard Error:** Cell Z30 should contain the *annual growth rate r* of the exponential curve corresponding to the regression line. That is:

$$r = e^{\hat{b}} - 1$$

Finally, we will approximate the *standard error of the growth rate* in cell Z31 by taking the same transformation on the standard error of \hat{b} (in cell Z26).

Get together with your project team and consider the following questions:

- (A) *For which customers do the linear and/or exponential trend tools appear to be helpful in making demand projections? Why or why not? What are the risks and limitations of using these projections?*
- (B) *What other demand projection methods do you feel may be useful for Liquair-Pro salespeople to have at their disposal?*
- (C) *Assume you want to provide the capability for a Liquair-Pro salesperson to easily incorporate a set of demand projections for an existing customer into a pricing tool. How will you design and implement this process?*