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POLYNOMIALS ASSOCIATED WITH EXPONENTIAL REGRESSION

Abstract. Fitting exponentials $a + be^{cx}$ to data by the least squares method is discussed. It is shown how the polynomials associated with this problem can be factored. The closure of the set of this type of functions defined on a finite domain is characterized and an existence theorem derived.

1. Introduction. Suppose we are given integers X_i and real numbers Y_i for $i = 1, \dots, t$. It is convenient if Y_i are integers but not necessary. The X_i and Y_i are not random variables, we are dealing with a numerical assignment. Instead of using $a + b \exp(cx)$ we prefer the model $a + bR^x$ and ask for a, b , and $R > 0$ that would minimize $\sum (a + bR^{X_i} - Y_i)^2$. If there is such a minimum, then the partial derivatives with respect to a, b, R are equal to zero at a point at which the minimum is located and, if $b \neq 0$, we are led to the equations

$$\begin{aligned} at + b \sum R^{X_i} - \sum Y_i &= 0, \\ a \sum R^{X_i} + b \sum R^{2X_i} - \sum Y_i R^{X_i} &= 0, \\ a \sum X_i R^{X_i} + b \sum X_i R^{2X_i} - \sum Y_i X_i R^{X_i} &= 0. \end{aligned}$$

If there exists a solution to this system of equations for some R , then $a, b, -1$ are the coefficients of a linear combination of the columns of the matrix

$$\begin{bmatrix} t & \sum R^{X_i} & \sum Y_i \\ \sum R^{X_i} & \sum R^{2X_i} & \sum Y_i R^{X_i} \\ \sum X_i R^{X_i} & \sum X_i R^{2X_i} & \sum Y_i X_i R^{X_i} \end{bmatrix}$$

yielding the zero vector.

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That can only happen if the column vectors are linearly dependent, for which a necessary and sufficient condition is that the determinant of the matrix is zero for a given R . If the values of X_i are nonnegative integers, the determinant may be written as a polynomial function of one variable. The positive roots of the polynomial give the values for which the sum of squares may be minimal.

We know that a polynomial does not have more roots than its degree. In our case it is three times the maximum of X_i . If X_i take on values 1 through 5, the determinant gives a polynomial of degree fifteen. It is generally known that finding roots of polynomials of high degrees is numerically very difficult.

This method and other details may be found in Gregg *et al.* (1964) (see also the survey paper Pimentel-Gomes (1953)). The origin of the idea is attributed to Pimentel-Gomes and Malavolta (1949).

2. Basic properties. Let $X = (X_1, \dots, X_t)'$ and $Y = (Y_1, \dots, Y_t)'$ be vectors for which we define the determinant $D(R, X, Y)$ as above. We can see immediately that

$$\begin{aligned} D(R, X, Y) &= -D(1/R, -X, Y), \\ D(R, X, Y) &= -D(R, X, -Y), \\ D(R, X, Y) &= D(1/R, -X, -Y). \end{aligned}$$

Let E be the vector consisting of ones and u a positive constant. We also have

$$\begin{aligned} D(R, uE + X, Y) &= R^u D(R, X, Y), \\ D(R, X, uE + Y) &= D(R, X, Y), \\ D(R, X, uY) &= uD(R, X, Y), \\ D(R, uX, Y) &= uD(R^u, X, Y). \end{aligned}$$

We can use these facts to modify X to get a smaller degree of the polynomial. We can subtract, for example, the minimal value of X from all its entries or we can divide all the entries by their greatest common divisor. We can also change the vector Y to keep the coefficients of the polynomial as small as possible.

We conclude that if $Z' = uE' + vY'$, where u, v are constants and $E' = (1, \dots, 1)$, then the polynomials obtained from the determinants $D(R, X, Z) = vD(R, X, Y)$ differ only by a constant factor v . This is obvious but it is less obvious that the converse is also true. If the polynomials $D(R, X, Z)$ and $D(R, X, Y)$ differ only by a constant factor, then $Z' = uE' + vY'$ for some u and v . The proof of this fact is tedious and too long to be presented here.

3. Factoring $D(R, X, Y)$ as a polynomial in R . The fact that $(R-1)^3$ divides $D(R, X, Y)$ when the components of X are equidistant, is mentioned in Pimentel-Gomes (1953) and is attributed to Nogueira (1950). It is straightforward but tedious to extend it to the general case.

If we let $R = 1$, we see that the value of the determinant is zero. Even though the value $R = 1$ does not have any suitable interpretation, this root enables us to factor the polynomial. We can try to investigate the determinant as a function of R and calculate its derivative at $R = 1$. As a reward we get a surprise, this derivative is zero no matter what X and Y look like. We even get a bigger surprise when we find out that the second derivative is always zero at $R = 1$ regardless of X and Y . It took a while to calculate the third derivative at $R = 1$; it is, after skipping the obvious subscripts,

$$3 \sum X \left(\sum X^2 \sum YX + \sum X^3 \sum Y - \sum X \sum YX^2 \right) + 3t \left(\sum X^2 \sum YX^2 - \sum X^3 \sum Y \sum X \right) - 3 \sum X^2 \sum X^2Y.$$

The Taylor expansion centered at $R = 1$ of the polynomial $P(R)$ of some degree M is

$$P(R) = P(1) + P'(1)(R - 1) + P''(1)(R - 1)^2/2 + P'''(1)(R - 1)^3/6 + \dots + P^{(M)}(1)(R - 1)^M/M!.$$

Since $P(1) = P'(1) = P''(1) = 0$, $(R - 1)^3$ is a factor of $P(R)$. It is not only the decrease in the degree of the polynomial that we achieve, we also discard multiple roots which can cause trouble in numerical calculations.

4. Examples. We give three examples to illustrate various situations.

Example 1		Example 2		Example 3	
y	x	y	x	y	x
0	0	1	0	0	0
0	1	2	1	5	1
0	2	3	2	2	2
1	3	1	3	7	3

Only four observations were considered in each example and data made up to suit the given purpose. We cannot present data with a larger number of observations because the degrees of polynomials would be prohibitive.

The purpose of these examples is to show that difficulties a routine use of statistical software may cause are for real. It is obvious that the usual numerical methods of finding minima will fail if the minima are not there.

EXAMPLE 1. We get the polynomial $2R^3 + 3R^2 + 4R + 1$. Positive coefficients mean no positive root and no $a, b, R > 0$ at which a local minimum would be located. The infimum of the sum of squares is zero since we can pick $R = N$, $a = 0$, $b = 1/N^3$, where N is a positive integer, and the sum of squares is

$$(0 - 1/N^3)^2 + (0 - N/N^3)^2 + (0 - N^2/N^3)^2 + (1 - N^3/N^3)^2 < 3/N^2.$$

For every ε there is an R such that the sum of squares is less than ε , thus its infimum is zero. For no positive value of R can this infimum be attained, because $a + bR^x$ is strictly monotone for $R > 0$, disregarding $R = 1$ and $b = 0$.

EXAMPLE 2. This example yields the polynomial $2R^3 + 7R^2 + 6R + 5$. When we try an iterative method, it is a gross error to believe that the parameters converge to a minimal solution. After repeated starts we can see that they diverge even though the sum of squares is stable. It is interesting to try to fit a straight line to this set of data by the least squares method. The sum of squares is equal to 4.2. No matter how hard we try, we never get a smaller sum of squares using $a + bR^x$, even the best guesses gave a little bit more than 4.2.

EXAMPLE 3. The polynomial we get is

$$P(R) = R^3 - 3R^2 - 3R + 1 = (R + 1)(R^2 - 4R + 1),$$

the equation $R^2 - 4R + 1 = 0$ has $R_1 = 2 + \sqrt{3}$ and $R_2 = 2 - \sqrt{3}$ as its roots. It can be shown that there are no local minima at R_1 or R_2 . It can also be seen that local minima are obtained in the closure of but outside the set of functions $a + bR^x$.

5. Closure. We have dealt with some examples first to make it clear that the set of functions $a + bR^x$ is not closed in the sense that a convergent sequence of functions from this set need not have a limit in this set.

Let us consider the closure of the set of functions $a + bR^x$ defined on a finite set. If there is, in this closure, another function with the sum of squares smaller than the sum of squares for any of the functions $a + bR^x$, then there is no minimal solution among the functions $a + bR^x$. It is conceivable that there might even be some local minima with the sum of squares greater than the infimum but the infimum could be attained by a function in the closure different from $a + bR^x$. It would be nice to know what kind of functions form this closure. We could then find the function with the minimal sum of squares in the closure and the question would be resolved.

6. Characterization of the closure. Let us assume that we have t points $0 = x_1 < \dots < x_t$ and let us consider the set of all functions defined

at these points. All these points and function values do not have to be integers any more. It is essential that, without loss of generality, we may indeed assume $x_1 = 0$ and $F(x_1) = 0$, since we can always add a constant or use a substitution to translate the original problem into one with $x_1 = 0$ and $F(x_1) = 0$ and backwards.

We take a Euclidean metric for the values of the functions as a distance $d(F, G) = (\sum (F(x_i) - G(x_i))^2)^{1/2}$ where F and G are two functions defined at x_i for $i = 1, \dots, t$. The distance $d(F, G)$ corresponds to the sum of squares of differences in an obvious manner.

By definition, a function F is in the closure of a set of functions if there is a sequence of functions in this set converging to F with respect to the distance $d(F, G)$. Since the number of points $0 = x_1 < \dots < x_t$ is finite, it is easy to see that the convergence of a sequence of functions with respect to the distance $d(F, G)$ is equivalent to the pointwise convergence. This fact is important because the pointwise convergence is easier to work with.

Assume that F is in the closure of functions $a + bR^x$ defined on $0 = x_1 < \dots < x_t$. Then there is a sequence $a_N + b_N R_N^{x_i}$ converging to $F(x_i)$ for every $i = 1, \dots, t$. The aim is to show what properties F has.

We may start with a special but illustrative case first. The case of $b_N = 0$ may be treated by discarding $a_N + b_N R_N^{x_i}$ with $b_N = 0$ if there are finitely many members with $b_N = 0$. In the case of an infinite subsequence with $b_K = 0$, it does not matter what R_K is; if $b_K = 0$ then $a_K + b_K R_K^x = a_K$. Since $F(x_1) = 0$, we get $\lim a_K = 0$ and, as a result, $F(x_i) = \lim(a_K + b_K R_K^{x_i}) = \lim a_K = 0$ for all i , that is, $F(x)$ must be a constant function. It obviously does not matter what the rest of the sequence $a_N + b_N R_N^x$ may look like, it converges and it does so to $F(x)$. This is the basic idea that the proofs are based on.

If we now assume that $b_N \neq 0$ for all N , the proof will depend on the assumptions about R_N and, accordingly, the classification of the limits of $a_N + b_N R_N^x$ will be made. Our basic assumption is that $R_N > 0$ for all N , no assumptions are made at this point regarding a_N or $b_N \neq 0$ and the questions of convergence of a_N or b_N are studied only when they are needed to determine the convergence of $a_N + b_N R_N^x$ as functions.

After the case that R_N is unbounded we get the fact that R_N is bounded and, therefore, we can pick a subsequence R_K that is convergent. Various possibilities of limits of this subsequence are then discussed.

7. The case of R_N unbounded. If the sequence R_N is not bounded, there must be a subsequence $a_K + b_K R_K^x$, convergent to $F(x)$ at x_1, \dots, x_t , for which R_K goes to infinity. We could even pick R_K increasing. At $x_1 = 0$, the value of $a_K + b_K R_K^x$ is equal to $a_K + b_K$ for any R . If $a_K + b_K R_K^0$ converges to $F(x_1) = 0$, $\lim(a_K + b_K) = 0$ must hold. Since R_K converges

to infinity and $x_t > 0$, b_K must go to zero, otherwise there would be $\varepsilon > 0$ such that $|b_K| > \varepsilon$ for infinitely many K , $|b_K|(R_K^{x_t} - 1)$ would not be bounded and $a_K + b_K R_K^{x_t} = a_K + b_K + b_K R_K^{x_t} - b_K = (a_K + b_K) + b_K(R_K^{x_t} - 1)$ would not be bounded, thus $\lim b_K = 0$ must hold and we get $\lim a_K = 0$.

Now we can see that $\lim b_K R_K^{x_t} = F(x_t)$ and $\lim b_K R_K^{x_i}$ is, for $i < t$, equal to $\lim b_K R_K^{x_t} / (R_K^{x_t - x_i})$. Since the numerator has a finite limit $F(x_t)$ and the denominator goes to infinity, the limit of the fraction is zero.

8. Subsequence R_K going to zero. Let there exist a subsequence R_K converging to zero. As before, it is clear that $\lim(a_K + b_K) = F(0) = 0$. It follows from the boundedness of $a_K + b_K R_K^x$ that there is a B_1 such that

$$|a_K + b_K + b_K R_K^x - b_K| = |a_K + b_K R_K^x| < B_1$$

for all K and from $\lim(a_K + b_K) = 0$ it follows that there is a B_2 for which $|b_K(R_K^x - 1)| < B_2$. Therefore $|b_K| < B_2 / (1 - R_K^x)$ starting from some K , thus b_K is a bounded sequence, because $\lim R_K^x = 0$. Thus $b_K R_K^x$ converges to zero for $x > 0$ and we must have $\lim a_K = F(x_2) = \dots = F(x_t)$.

9. Subsequence with a positive limit other than one. Let there be a subsequence R_K with $\lim R_K = R$, where R is positive, finite, and different from one. We again have $\lim(a_K + b_K) = 0$ and we know that $\lim(a_K + b_K R_K^{x_t}) = F(x_t)$. Since $a_K + b_K R_K^{x_t} = a_K + b_K + b_K R_K^{x_t} - b_K = a_K + b_K + b_K(R_K^{x_t} - 1)$ converges to $F(x_t)$, we have $\lim b_K(R_K^{x_t} - 1) = F(x_t)$. We know, because of continuity, that $\lim(R_K^{x_t} - 1) = R^{x_t} - 1$, and the following will also hold:

$$\lim b_K = \lim b_K(R_K^{x_t} - 1) / (R_K^{x_t} - 1) = F(x_t) / (R^{x_t} - 1).$$

Set $b = \lim b_K$. Then we can see that $a = \lim a_K = -b$. Thus $a_K + b_K R_K^x$ converges to $a + bR^x$.

10. Subsequence with limit one. (a) If there is a subsequence R_K with all elements one, then $a_K + b_K R_K^x = a_K + b_K$. We know that the limit of this sequence is zero, which implies that $F(x_i) = 0$ for all i .

(b) Let each R_K be different from one. Set $U_K(x) = a_K + b_K R_K^x$. Then $U_K(0) = a_K + b_K$. Define $D_K(x) = U_K(x) - U_K(0)$. Then we get $D_K(x) = b_K(R_K^x - 1)$. This difference will be treated separately and its limit will be added to $\lim(a_K + b_K) = 0$. We will need the following limit for R going to one while x and x_t are fixed:

$$\lim_{R \rightarrow 1} (R^x - 1) / (R^{x_t} - 1) = x / x_t.$$

Since $a_K + b_K R_K^{x_t}$ converges to $F(x_t)$ or $\lim U_K(x_t) = F(x_t)$ and at the same time $\lim U_K(0) = F(x_1) = 0$, we have $\lim D_K(x_t) = F(x_t)$. Thus

$$\lim D_K(x_t) = \lim b_K(R_K^{x_t} - 1) = F(x_t).$$

We can use the limit we found beforehand to get

$$\lim D_K(x) = \lim b_K(R_K^x - 1) = \lim b_K(R_K^{x_t} - 1) \frac{R_K^x - 1}{R_K^{x_t} - 1}.$$

That gives $\lim D_K(x) = xF(x_t)/x_t$ and shows that the limit of $D_K(x)$ exists for $x > 0$ and is proportional to x .

As the limit of $a_K + b_K R_K^x$ is zero for $x_1 = 0$ and the value of the function F is the sum of its value at zero and the difference, we conclude that F is a linear function mx , where $m = F(x_t)/x_t$.

11. Completion of characterization. Now we can work our way back and see that the closure consists of functions we get by adding a constant or by linear substitution. As a result, we can describe the functions in the closure in the following manner.

THEOREM. *If there are $t \geq 3$ points $x_1 < \dots < x_t$, then the closure of the set of functions $a + bR^x$, where $0 < R \neq 1$, defined at these points is formed by the following functions:*

- (1) *the functions constant except at the first point x_1 , that is, $F(x_1)$ is arbitrary, $F(x_2) = F(x_3) = \dots = F(x_t)$;*
- (2) *the functions constant except the last point x_t , that is, $F(x_1) = F(x_2) = \dots = F(x_{t-1})$, arbitrary $F(x_t)$;*
- (3) *$a + mx$;*
- (4) *$a + bR^x$;*

Proof. What is left to prove is that all the functions of the four types are in the closure. Let $F(x_i) = 0$ for $i < t$ and $F(x_t) = 1$. As in Example 1, let $a = 0$, $b = 1/N^{x_t}$, and $R = N$, N positive integer. We get the sum of squares

$$(0 - N^{x_1}/N^{x_t})^2 + \dots + (0 - N^{x_{t-1}}/N^{x_t})^2 + (1 - N^{x_t}/N^{x_t})^2 < (t - 1)/N^{2(x_t - x_{t-1})}.$$

As $x_t - x_{t-1} > 0$, by choosing N sufficiently large the sum of squares can be made arbitrarily small. The general case can be obtained by choosing suitable a and b .

When $F(x_1) = 1$ and $F(x_i) = 0$ for $i > 1$, we proceed in the similar manner. Let $a = 0$, $b = 1/N^{x_1}$, and $R = 1/N$, N natural, and write down the sum of squares of deviations.

Let $F(x) = q + mx$, $m \neq 0$. Choose any sequence R_N convergent to one with all R_N different from one. The sequences a_N and b_N can be obtained from the equations

$$\begin{aligned} a_N + b_N R_N^0 &= q, \\ a_N + b_N R_N^{x_t} &= q + mx_t. \end{aligned}$$

The determinant of the system is $R_N^{x_t} - 1 \neq 0$ and we obtain the sequence

$a_N + b_N R_N^x$. Define $D_N(x) = U_N(x) - U_N(0) = b_N(R_N^x - 1)$ and compute

$$\lim D_N(x) = \lim b_N(R_N^{x_t} - 1) \frac{R_N^x - 1}{R_N^{x_t} - 1} = mx_t(x/x_t) = mx,$$

because $b_N R_N^{x_t} - b_N = a_N + b_N R_N^{x_t} - (a_N + b_N R_N^0) = q + mx_t - q = mx_t$ by definition of a_N and b_N .

12. Existence of a minimal solution in the closure. The closure of the set of functions $a + bR^x$ has been described and what is left is the proof of the existence of a minimal solution in this closure.

THEOREM. *Let (y_i, x_i) for $i = 1, \dots, t$ be pairs of real numbers with at least three x 's distinct. Among the functions $F(x)$ in the closure of functions of the type $a + bR^x$ defined at x_1, x_2, \dots, x_t there exists a function $F_0(x)$ that minimizes the sum of squares $\sum_{i=1}^t (F(x_i) - y_i)^2$.*

Proof. We consider (y_i, x_i) for $i = 1, \dots, t$. We may have $x_i = x_j$ for some $i \neq j$, which is what statisticians call repeated observations. We can choose one fixed function $a_0 + b_0 R_0^x$ and determine the sum of the squares of the deviations from y_i . By a closed ball B_0 in the set of functions defined on the x_i for $i = 1, \dots, t$ we will mean the subset of those functions for which $\sum (F(x_i) - y_i)^2 \leq \sum (a_0 + b_0 R_0^{x_i} - y_i)^2$. We are not dealing exactly with a closed ball as usually defined in topology because we have to consider repeated observations. It is clear that this ball B_0 is nonempty, closed and bounded, because the sum of squares is a continuous function and the image under this mapping is a closed interval. One of its endpoints is $\sum (a_0 + b_0 R_0^{x_i} - y_i)^2$; the other endpoint is the sum of the squares of the deviations $\sum (F_{\min}(x_i) - y_i)^2$ of the function F_{\min} , the values of which are arithmetical averages of y_i calculated for each value x_i , for these averages minimize the sum of squares of deviations at each value of x_i separately.

If there are repeated values of x_i , we may not have exactly a ball. This is especially the case when for some $i \neq j$ the values of y_i and y_j may be different, whereas x_i and x_j are the same. On the other hand, if, for some i , x_i is different from all other x_j , $j \neq i$, we have $F_{\min}(x_i) = y_i$.

To show that the sums of squares of deviations take on all the values in the interval we set $F_C(x_i) = F_{\min}(x_i)$ for $i > 1$ while $F_C(x_1) = F_{\min}(x_1) + C$ for $i = 1$. Thus, by picking the right $C \geq 0$, we get any desired sum of squares of deviations from $\sum (F_{\min}(x_i) - y_i)^2$ to infinity.

The intersection of the closed ball B_0 with the closure of the set of functions $a + bR^x$ is nonempty, closed and bounded, thus it is a nonempty and compact set and we know that a continuous mapping, such as the sum of squares of deviations, as a mapping of F , takes on its infimum on a set like that at some point F_0 .

13. Applications and comments. Another model may be given by the function $b \exp(cx) = bR^x$ defined on a finite set. To find out how the closure of the set of functions of this type looks like, we would have to discuss three cases: R_K unbounded, R_K going to zero, and R_K going to some finite positive limit. $R = 1$ is acceptable for it defines a constant function.

Other frequently used models derived from $a + bR^x$ are the logistic function and the Gompertz function, the former being just the reciprocal of $a + bR^x$, the latter the exponential function of $a + bR^x$. It is clear that in both cases the closure can be described in a similar manner. It is just that one has to be careful about the division by zero.

Since it is hard to find all the positive real roots of polynomials of large degrees, users of statistical software will use iterative methods to minimize the sum of squares. But we now know that the result has to be compared against minimal sums of squares of other types of functions in the closure of the set of exponential ones.

To determine if the reciprocal of a linear function yields a better sum of squares than the logistic one we have to deal with two nonlinear regression problems. The graphical visualization of the reciprocal of the logistic function is of great value to see how close it is to the straight line.

References

- J. V. Gregg, C. H. Hossell and J. T. Richardson (1964), *Mathematical Trend Curves: an Aid to Forecasting*, Oliver and Boyd, Edinburgh.
- I. R. Nogueira (1950), *Sobre una propiedad da equação utilizada para a interpolação da lei de Mitscherlich*, Anais Esc. Sup. Agr. Luiz de Queiroz 7, 105–108.
- F. Pimentel-Gomes (1953), *The use of Mitscherlich's regression law in the analysis of experiments with fertilizers*, Biometrics 9, 498–516.
- F. Pimentel-Gomes and E. Malavolta (1949), *Aspectos matematicos e estatisticos da lei de Mitscherlich*, Anais Esc. Sup. Agr. Luiz de Queiroz 6, 193–229.

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