

STUDY OF TENSIONAL PROPERTIES OF POLYPROPYLENE YARNS USING THE REGRESSION AND CORRELATION METHOD

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REZUMAT. În cadrul lucrării s-a efectuat un studiu asupra proprietăților tensionale ale firelor polifilamentare din polipropilenă, destinate realizării țesăturilor înguste pentru confecționarea centurilor de siguranță auto. Cu ajutorul programului IBM SPSS 19.0, utilizând metoda regresiei și corelației, se studiază dependența dintre o variabilă rezultativă/dependentă y – forța de rupere, P_r (N) și o variabilă independentă x – lucrul mecanic de deformare la rupere, L (N · m). Metoda de regresie și corelație utilizată în cadrul studiului, a condus la rezolvarea următoarelor aspecte: identificarea existenței legăturii dintre variabila dependentă și variabila independentă; determinarea gradului de intensitate a legăturii și stabilirea sensului și formei legăturii dintre cele două variabile. Regresia ne arată cum (ca formă analitică) o variabilă este dependentă de altă variabilă (sau de alte variabile), iar corelația ne arată gradul în care o variabilă este dependentă de o altă variabilă (sau alte variabile).

Cuvinte cheie: polipropilenă, fire, forță de rupere, metoda regresiei, corelație.

ABSTRACT. In this paper it was done a study about the tensional properties of the polyfilamentary yarns made from polypropylene, used in weaving seat belts made from webbing. With the program IBM SPSS 19.0, using the regression and correlation method, studying the dependency between a resulting/dependent variable y – breaking force, P_r (N) and an independent variable x – breakdown deformation energy, L (N·m). The regression and correlation method, used in this paper, led to the issue of the following aspects: identification of the existence of the connection between the dependent variable and an independent variable; the determination of intensity degree of the connection and establishing the direction and shape of the connection between the two variables. The regression shows us how (the analytical form) a variable is dependent of another variable (or other variables), and the correlation shows us the degree in which a variable is dependent of another variable (or other variables).

Keywords: polypropylene, yarns, breaking force, regression method, correlation.

1. INTRODUCTION

Among the methods that can be applied to determine a relationship of dependency between two quantities of interest, called variables include correlation and regression analysis and used in various fields (psychology, economics, medicine, engineering) [1-4].

It should be noted that neither the regression analysis nor correlation does not reveal a relationship type cause-effect between the variables involved in the model, they only show to what extent are they related one with another. If the analysis is correct, if it turns out that there are solid arguments in this respect, the causality is an issue that can be discussed later [5, 6].

The tensional properties cues define the behavior of the textile products at the tensile stress, summarizes the technological achievement and standardizes,

constituting quality criteria [7, 8]. Characterization products straight / flat cues properties tensional requires peculiarities definition determined by the product structure and metrological conditions test (trial basis, the size of the specimen, the initial state of the specimen, pretension and orientation towards axles technological product) [8].

Propylene is the based monomer from which it is produced polypropylene (PP) by Ziegler-Natta catalysis process. Polypropylene has a good heat resistance, along with low cost, a higher modulus of elasticity and the possibility of obtaining a number of improved variants, this makes the polypropylene to be one of the most used polymers used in the automotive industry [9, 10].

Recent developments in polypropylene sector will provide priority position of this group of plastics materials in automotive applications. High crystallinity polypropylene (HCPP) with improved rigidity,

is now used in applications of automotive interior and engine compartment applications, without the need to add 10% - 20% talc [11]. Due to the registered progresses of the properties of plastic material, bumpers can be produced with a thickness of 2.4 - 2.6 mm, down from about 3.2 mm, thus providing a significant cost and weight reduction [12, 13].

Standard polypropylene is in the form of monofilament and polyfilamentary yarns and fibers. The fibers are obtained by cutting the cable in different lengths depending on the destination, or antistatic treated, wrinkled and painted in mass. High strength polypropylene is in the form of monofilament or polyfilamentary yarns, obtained by spinning water-cooled and air-cooled (more commonly used, but the wires have less resistance) [14]. High strength polypropylene yarns are used in making technical textiles with destinations in various fields such as agriculture, construction and automotives [15-18].

2. MATERIALS AND METHODS

Materials used in this study were polypropylene yarns of linear density of 1100 dtex, designed to achieve webbing for manufacturing automotive safety belts. Checking tensile properties of polypropylene yarns was done by dynamometric tests on samples with standardized dimensions (500mm) in accordance with EN 13934-1 standard ISO / 2002 [19], the dynamometer H5K – Tinius Olsen T calculation software QMAT textiles, TEXTILEXPRT research Laboratory, department of Textile Engineering and Design.

The obtained experimental data were statistical processed using IBM SPSS 19.0 software using regression and correlation method.

In regression and correlation analysis using SPSS software the following steps are:

- identification of the existence of the connection through the graphical method which involves construction of the graphical correlation curve;
- setting the direction and form of the connection by analytical methods to study the connections (determining the regression model, estimating model parameters and test their significance).

3. RESULTS AND DISCUSSIONS

After processing the data in SPSS and by regression and correlation analysis the following results were obtained on statistical variables: descriptive statistics indicators (Descriptive Statistics); estimated coefficients (Estimates); confidence intervals of the coefficients (Confidence intervals); calculating statistics R, R² and ANOVA table (Model fit); change of the determination coefficient and testing change significance adding each block of variables (R squared change). In the Residual zone occurs an analysis of residues to decide on their normality and diagnose outliers [20, 21].

The statistical parameters calculated for each variable are shown in Table 1. (output Descriptive Statistics), in which is observed the homogeneity of values of the two variables.

Analyzing the data in Table 2, which contains information about each variable analyzed independently, indicator Skewness used in distribution analysis of a series of data to indicate deviation of empirical distribution in relation to a symmetrical distribution around the mean and indicator Kurtosis used in distribution analysis of a series of data to indicate the degree of flattening or sharpening a distribution, may be observed that:

- dependent variable "Pr (N), breaking force" is characterized by an average of 65.79 and variance of 26.95; Skewness (-0.300) <0 indicates that the distribution is tilted to the right, with more extreme values to the left; Kurtosis (-0.612) <3 indicates a platikurtic distribution, a distribution flatter than a normal one, with values spread out over a longer interval around the mean. The probability for extreme values is less than for a normal distribution.

- independent variable L (N.m), breakdown deformation energy is characterized by a mean value of 2.38 and variance of 0.382; Skewness (0.784) > 0 - the distribution is tilted to the left, with more extreme values to the right; Kurtosis (1.145 <3) - leptokurtic distribution, a distribution sharper than a normal one; having more concentrated around the mean values and thicker tails parallels what high probabilities for extreme values.

Table 1. Descriptive Statistics

	N	Min	Max	Mean	Std. Deviation	Variance	Skewness		Kurtosis	
							Statist.	Std. Error	Statist.	Std. Error
Pr [N]	50	54.3	74.5	65.79	5.19137	26.950	-0.300	0.337	-0.612	0.662
L [N.m]	50	1.22	4.31	2.382	0.61800	0.382	0.784	0.337	1.145	0.662

Table 2. Model Summary and Parameter Estimates. Dependent Variable: Pr [N]

Equation	Model Summary					Parameter Estimates			
	R Square	F	df1	df2	Sig.	Constant	b1	b2	b3
Linear	0.725	126.777	1	48	0.000	48.744	7.154		
Logarithmic	0.786	176.661	1	48	0.000	50.762	17.974		
Quadratic	0.813	102.185	2	47	0.000	31.704	21.057	-2.658	
Cubic	0.819	69.237	3	46	0.000	45.830	3.505	4.172	-0.833
Exponential	0.714	120.052	1	48	0.000	50.489	0.110		

The independent variable is L [N.m].

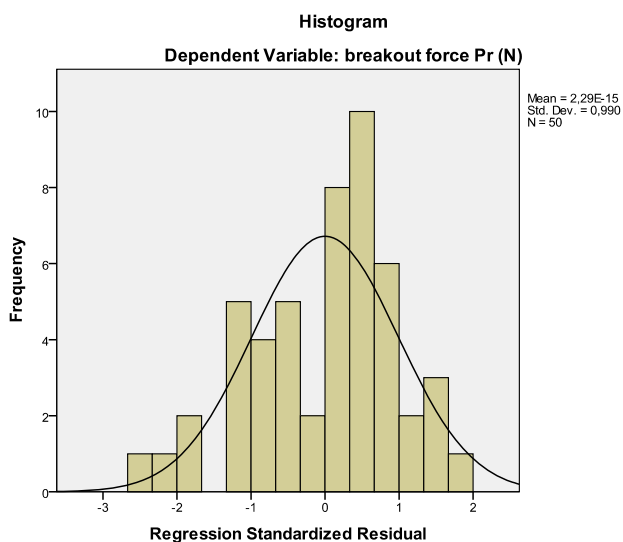


Fig. 1. The histogram and the normal distribution curve.

Graphic representation as histograms (Fig. 1) is another way of checking the normality of distribution by viewing graphical differences between empirical and theoretical distribution (Gauss Curve). It finds that the distribution is below a normal curve to the left having a slight elongation ($sk < 0$) and in terms of sharpness there is a platikurtic tendency ($k < 3$) meaning that the results are scattered toward the mean. The values of the two indicators for assessing the normality of distribution were determined in Table 1.

Choosing the best regression and correlation model for the two studied variables, using SPSS, can be achieved by interpretation of the most important results of the output presented in Table 2 and the Scatterplot - Fig. 2.

From Table 2 it is observed that for each regression model we find in the Model Summary columns the determination report (R Square) and its testing with test F and in Parameter Estimates columns equation regression coefficients. Based on the R Square from the Summary Model, it will

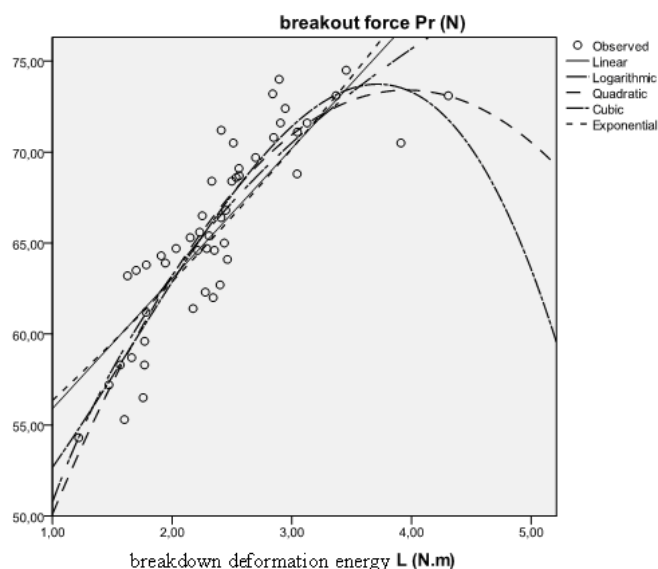


Fig. 2. Diagrama de dispersie sau Scatterplot.

decide the best regression model, finally, the more R Square is closer to value 1 the more the regression model it's most appropriate.

Analyzing the data in Table 2, we see that the highest value for R Square of all regression models are registered to function Cubic - by grade 3 (R Square = 0.819), the function Quadratic - function Grade 2 (R Square = 0.813).

For the Cubic model we analyze the most important results of the output presented in Tables 3, 4 and 5

Table 3. Model Summary

R	R Square	Adjusted R Square	Std. Error of the Estimate
0.905	0.819	0.807	2.281

The independent variable is L [N.m].

Table 3. Model Summary contains information concerning the correlation coefficient and standard error of the estimation. Note the correlation coefficient R (0.905) and determination coefficient

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R Square (0.819) shows that between breakdown deformation energy to breakage and breakout force there is a link type parabolic Grade 3 (Fig. 3), direct and strong meaning 81.9% of the variance of the dependent variable (breaking strength) is explained by the regression equation.

In Table 4. ANOVA the important information is statistical test F which tests the global significance of independent variables (only one in the case of simple regression, the procedure acquires substance when multiple regression). In column Sig. critical probability of the test is displayed, Sig <0.05 confirms that the parabolic relationship Grade 3 between the two considered variables is significant,

rejecting the irrelevant signification hypothesis of the independent variables in favor of the hypothesis regression model, is significant. They say that the test is a significant test of R Square.

The values shown in Table 5 column Unstandardized Coefficients B, help us write the equation regression for our model

$$y = a + bx + cx^2 + dx^3 \quad (1)$$

Where: a = (Constant); b = mechanical work; c = mechanical work **2; d = mechanical work **3

So the equation becomes:

$$y = 45.830 + 3.505 \cdot x + 4.172 \cdot x^2 - 0.833 \cdot x^3 \quad (2)$$

Table 4. ANOVA

	Sum of Squares	df	Mean Square	F	Sig.
Regression	1081.134	3	360.378	69.237	0.000
Residual	239.431	46	5.205		
Total	1320.565	49			

The independent variable is L [N.m].

Table 5. Coefficients

	Unstandardized Coefficients		Standardized Coefficients	t	Sig.
	B	Std. Error	Beta		
L [N.m]	3.505	14.942	0.417	0.235	0.816
L [N.m] ** 2	4.172	5.725	2.639	0.729	0.470
L [N.m] ** 3	-0.833	0.695	-2.294	-1.199	0.237
(Constant)	45.830	12.390		3.699	0.001

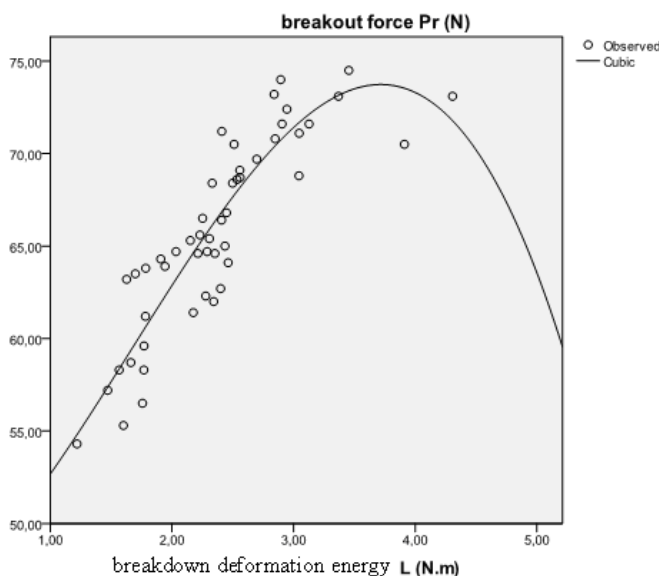


Fig. 3. Scatter Plot Chart for cubic model.

If we are testing coefficients regression equation using t test for the hypothesis that they would be significantly different from zero, it is observed that the significance threshold for the coefficients b, c and d is greater than 0.05, therefore these coefficients are not reliable to build a regression

equation to define clearly a significant link parabolic grade 3 for the two variables.

In conclusion to choose the most appropriate regression model must simultaneously meet the three conditions of the table:

- high value R-Square;

- in the ANOVA analysis A value for F Sig less than 0.05;
- Sig values of test t for the independent variable of the equation coefficients to be below 0.05.

For the Quadratic model we will analyze the most important results of the output presented in Tables 6; 7; 8.

Table 6. Model Summary

R	R Square	Adjusted R Square	Std. Error of the Estimate
0.902	0.813	0.805	2.292

The independent variable is L [N.m].

The mathematical model is appropriate, because it is respected the condition for $F_C > F_T$

$$F_c = 102.185 > F_{(2,47)} = 3.23$$

So the best regression equation for the two variables is Quadratic regression and has the following form:

$$y = 31,704 + 21,057x - 2,658x^2 \quad (3)$$

Table 7 ANOVA

	Sum of Squares	df	Mean Square	F	Sig.
Regression	1073.651	2	536.826	102.185	0.000
Residual	246.914	47	5.253		
Total	1320.565	49			

The independent variable is L [N.m].

Table 8 Coefficients

	Unstandardized Coefficients		Standardized Coefficients		t	Sig.
	B	Std. Error	Beta			
L [N.m]	21.057	3.009	2.507		6.999	0.000
L [N.m] ** 2	-2.658	0.566	-1.681		-4.694	0.000
(Constant)	31.704	3.857			8.220	0.000

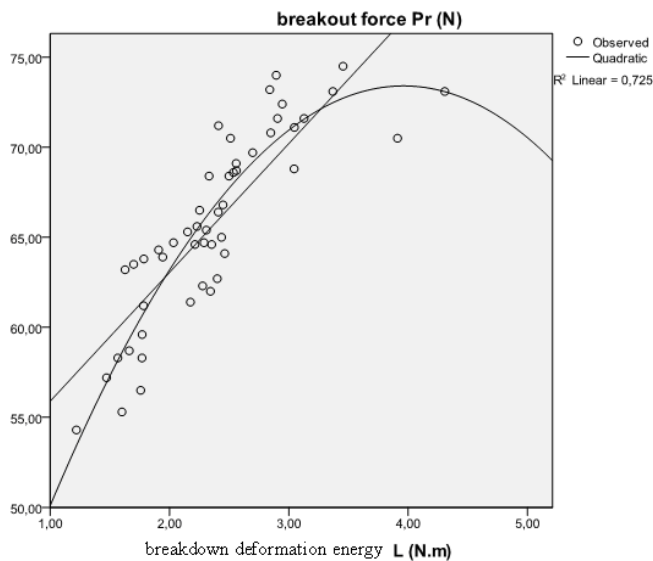


Fig. 4. Diagrama Scatterplot for the Quadratic model.

4. CONCLUSIONS

1. The simple regression is a prediction procedure, based on the correlation between two quantitative variables.

2. The precision prediction is given by the correlation coefficient between variables. As R is higher, so the prediction of the values of a variable (called criterion) starting from the values of the other variables (called predictor) is better. At the limit, when $R = 1$, the prediction is perfect.

3. The regression model is expressed graphically as a straight line whose route through the point cloud minimizes the distances between the line points and those of the correlation scatterplot.

4. The Skewness indicator value $(-0.300) < 0$ in the case of the dependent variable, indicates the fact that the distribution is tilted to the right, with more extreme values to the left, and if the independent variable Skewness $(0.784) > 0$ - the distribution is tilted to the left having more extreme values to the right.

5. The Kurtosis indicator value $(-0.612) < 3$ in the case of the independent variable, indicates a platikurtic distribution, a distribution flatter than a normal one, with values spread out over a longer interval around the mean, and in the case of a Kurtosis independent variable $(1.145 < 3)$ indicates a leptokurtic distribution, a distribution sharper than a normal one ; having more concentrated values around the mean and thicker tails which means high probabilities for extreme values.

6. The best regression and correlation model in this study was Quadratic model, because it meets simultaneously three conditions: high value for the coefficient of determination, R Square (0.813); the value from the ANOVA analysis for F Sig < 0.05 and the values for Sig. t for the coefficients of the independent variable in the regression equation are below the significance threshold $\alpha = 0.05$.

7. The determination coefficient of R^2 shows that the variant of the dependent variable breakout force is explained by the variant of the independent variable breakdown deformation energy in proportion of 81.3%.

8. The existence of the strong and positive connection between breaking force and breakdown deformation energy of the polypropylene yarn is given by the correlation coefficient $R = 0.902$.

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