

Use of Multiple Linear Regression Technology in Controlling the Technical Specifications of Industrial Products: Predicting the Mechanical Specifications of the Products of the Longitudinal Rolling Factory of the LISCo

*Omar Azouza^a, Madi Naser^b

^aDept. of Industrial Engineering and Manufacturing, Faculty of Engineering, Misurata University, Libya

^bDept. of Petroleum and Gas Engineering, Faculty of Engineering, Misurata University, Libya

Keywords:

linear regression
Libyan Iron and Steel Company
industrial products
predicting

ABSTRACT

The main objective of this study is to arrive at the formulation of a model through which to predict the mechanical specifications of the products of the longitudinal rolling mill of the Libyan Iron and Steel Company (LISCo), using the multiple linear regression method, and in a simple, accurate manner through which the quality of the products can be controlled within the specifications approved by the LISCo. Given the proportions of the added chemicals. As a conclusion, the mechanical specifications of the products were predicted from yield stress, tensile strength and elongation ratio, and the error rate in predicting these mechanical specifications was very close to zero. This research also contributes to introducing the LISCo to the importance of prediction models for mechanical properties, which work to reduce the rate of uncertainty, especially with the development of forecasting methods that allow improving the degree of accuracy and help in making sound decisions.

إستخدام تقنية الانحدار الخطي المتعدد في التحكم في المواصفات الفنية للمنتجات الصناعية (التنبؤ بالمواصفات الميكانيكية لمنتجات مصنع الدرفلة الطولية بالشركة الليبية للحديد والصلب)

*عمر اعزوزه¹ و مادي نصر²

¹ قسم الهندسة الصناعية و التصنيع ، كلية الهندسة ، جامعة مصراته ، ليبيا

² قسم الهندسة النفطية ، كلية الهندسة ، جامعة مصراته ، ليبيا

الكلمات المفتاحية:

الانحدار الخطي
الشركة الليبية للحديد والصلب
منتجات صناعية
توقعات

الملخص

سعيًا من خلال هذه الدراسة للوصول إلى صياغة نموذج يمكن من خلاله التنبؤ بالمواصفات الميكانيكية، لمنتجات مصنع الدرفلة الطولية بالشركة الليبية للحديد والصلب، باستخدام طريقة الانحدار الخطي المتعدد، وبأسلوب بسيط، ودقيق يمكن من خلاله ضبط جودة المنتجات في حدود المواصفات المعتمدة لدى الشركة الليبية للحديد والصلب، بمعلومية نسب المواد الكيميائية المضافة حيث تم التنبؤ بالمواصفات الميكانيكية للمنتجات من اجهاد خضوع وقوة شد ونسبة الاستطالة وكانت نسبة الخطأ في التنبؤ بهذه المواصفات الميكانيكية قريبة جداً لصففر. كما يسهم هذا البحث في تعريف الشركة الليبية للحديد والصلب بأهمية نماذج التنبؤ بالخواص الميكانيكية، التي تعمل على التخفيض من نسبة عدم التأكد، خاصة مع تطور أساليب التنبؤ التي تسمح بتحسين درجة الدقة، وتساعد في اتخاذ القرارات السليمة.

1. Introduction:

The industrial and technological development depends largely on the progress in the field of products and their quality, and as a result of

this great industrial development that the world witnessed in all fields, the need arose to know the mechanical properties of materials

*Corresponding author:

E-mail addresses: o.azouza@eng.misuratau.edu.ly, (Madi Naser) m.naser@eng.misuratau.edu.ly

Article History : Received 26 May 2021 - Received in revised form 21 October 2021 - Accepted 08 November 2021

with multiple industrial uses so that these characteristics have high quality specifications in terms of resistance to pressure, durability, and submission Tensile, resistance, and mechanical properties in general, in order to predict them, and for their adoption in various industrial applications. The mechanical properties can be divided into the following [3]:

Yield Stress: Yield stress is known as the yield point, a point on the stress-strain curve [3]. Yield strength is a property related to critical materials, which is exploited by many basic techniques of working materials to reshape materials by pressure (such as: forming, rolling, pressing, bending, and hydraulic shaping) to separate materials by cutting (such as using a machine) or shearing, and to connect elements rigidly to fasteners, the yield load is taken It is the load applied to the center of a spring loaded to straighten its plates [2].

Tensile Strength: It is a measure of how much stress you dance, dances the right way, which then crashes or loses, crashes or outruns, without crashing and can be calculated by the following equation:

$$\sigma = \frac{F}{A}$$

Where:

σ : tensile strength, unit of measurement (newtons/square meter) (N/m²).

F: the amount of force, the force that pulls the ends of the metal together, the unit of measure (newtons) (N).

A: The cross-sectional area of the metal, unit of measure (square meters) (m²).

Elongation Ratio: It is a measure of the ductility of a material and is defined from the following equation [3]:

$$\epsilon = \frac{\Delta L}{L_0} = \frac{L - L_0}{L_0}$$

Where:

ϵ : elongation ratio.

ΔL : the increase in length.

L_0 : is the original length.

L: is the final length.

Prediction/ Forecasting:

Forecasting is the mainstay for the internal planning of the company's policy decisions, as most of the administrative decisions in the facility depend on forecasting directly or indirectly. [4].

Reasons for the rapid spread of the use of forecasting:

Forecasting is based on a number of factors, the most important of which are [4]:

- ✚ Continuous progress in models and methods of forecasting.
- ✚ Increasing the volume and complexity of the administration's work, which made it more difficult to deal efficiently with decision-making without relying on models that facilitated unraveling the ambiguities of the future.
- ✚ The administration has become completely convinced at the present time of the advantages resulting from the use of forecasting, and its feasibility in completing planning processes.

Forecasting Requirements:

Forecasting depends on a set of requirements, the most important of which are [5]:

- ✚ Interest in and familiarity with various past historical records related to the demand forecasting process.
- ✚ List the factors that affect the sales volume in the past, such as income, advertising, commodity quality and price.
- ✚ Develop a visualization of the future activity of sales.
- ✚ Review and correct predictions and evaluation for future feedback.
- ✚ Determining the dependent demand and the independent demand.
- ✚ Attention and full knowledge of competing and alternative goods and their development.
- ✚ Taking into account the life cycle of the commodity, while forecasting sales, and at what stage of the cycle the commodity will be, where sales are at their peak in a phase of saturation and this phase is characterized by a degree of relative stability in sales, and in this phase it is necessary to seek the assistance of

experts to create new benefits for the commodity, to increase the demand for it, otherwise the commodity will go into decline.

Steps To Prepare The Forecast:

Preparing the forecast is by following these steps [5]:

✚ **First Step:** Determining the purpose of the forecasting process, [5].

Step Two: Collecting historical data, whether on economic trends from government documents, and Figure (1) illustrates how to collect data [5].

Φ mm	Weight kg/m	Area of Grose - Section in Cm ²									
		1	2	3	4	5	6	7	8	9	10
5	0.154	0.196	0.393	0.589	0.785	0.982	1.18	1.37	1.57	1.77	1
6	0.222	0.283	0.566	0.848	1.13	1.41	1.70	1.98	2.26	2.54	2
7	0.302	0.385	0.770	1.15	1.54	1.93	2.31	2.69	3.08	3.46	3
8	0.395	0.503	1.01	0.51	2.01	2.51	3.02	3.52	4.02	4.52	5
10	0.617	0.785	1.57	2.36	3.14	3.93	4.71	5.50	6.28	7.07	7
12	0.888	1.13	2.26	3.39	4.52	5.65	6.79	7.92	9.05	1.02	11
13	1.04	1.33	2.66	3.98	5.31	6.64	7.96	9.29	10.6	11.9	11
14	1.21	1.54	3.08	4.62	6.16	7.70	9.24	10.8	12.3	13.9	1
15	1.58	2.01	4.02	6.03	8.04	10.1	12.1	14.1	16.1	18.1	2
18	2.00	2.54	5.09	7.63	10.2	12.7	15.3	17.8	20.4	22.9	2
19	2.23	2.835	5.67	8.50	11.30	14.2	17.0	19.9	22.7	25.5	2
20	2.47	3.14	6.28	9.42	12.6	15.7	18.8	22.0	25.1	28.3	3
22	2.98	3.80	7.60	11.4	15.2	19.0	22.8	26.6	30.4	34.2	3
24	3.55	4.52	9.04	13.6	18.1	22.6	27.0	31.7	36.2	40.7	4
25	3.85	4.91	9.82	14.7	19.6	24.5	29.5	34.4	39.3	44.2	4
26	4.17	5.31	10.6	13.9	21.2	26.5	31.9	37.2	42.5	47.0	5
28	3.83	6.16	12.3	18.5	24.6	30.8	37.0	43.1	49.3	55.4	6
30	5.55	7.07	14.1	21.2	28.3	35.3	42.4	49.5	56.6	63.6	7
32	6.31	8.04	16.1	24.1	32.2	40.2	48.3	56.3	64.3	72.4	8
34	7.13	9.08	18.20	27.2	36.3	45.4	54.5	63.6	72.6	81.7	9
36	7.99	10.2	20.4	30.6	40.8	50.7	61.2	71.4	81.6	91.8	1
38	8.9	11.3	22.6	33.9	45.2	56.5	67.8	79.1	90.4	102	1

Figure (1) Data Collection Form [5]

✚ **Step Three:** Displaying historical data on a graph as in Figure (2), to determine the extent to which there is a certain pattern of data trend [4].

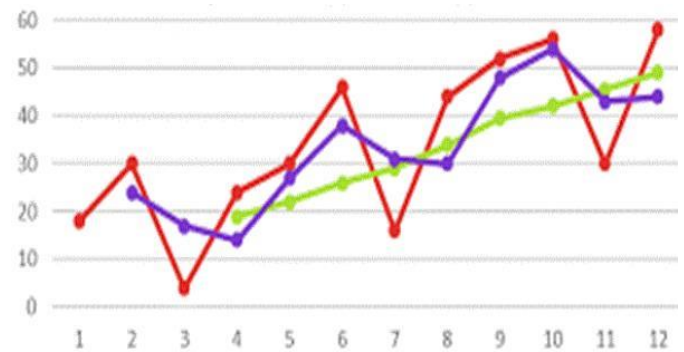


Figure (2) Specific Data Cycle [4]

✚ **Step Four:** Choosing a forecasting model. There are three main methods of forecasting: time-series models, causal models, and qualitative models. In general, we find that time-series models use historical data for the thing to be predicted to predict future demand values. These models depend on the study of the previous behavior of factors over time and we conclude including what will be in the future, and causal models are used for data on independent variables to develop predictions for dependent variables [5].

✚ **Step Five:** Conducting experiments that show the correctness of the methods that were used to predict the real values that appeared during the past period, and the method that produces the smallest average error is usually used, and you use it to predict the coming period. There are common measures to measure the error shown in Figure 3, which are the bias error and the average Absolute deviation and relative error, and when the number of the sample is 30 or more, we can judge that the distribution of samples is a normal distribution, and the distribution represents the estimate of the standard deviation of the community [5].

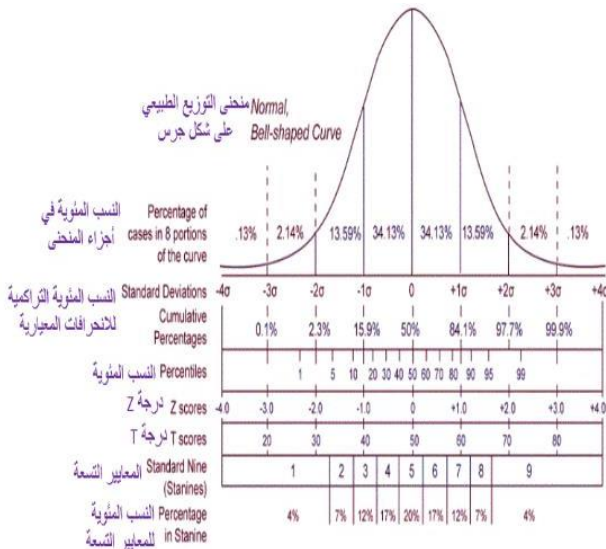


Figure 3 Error Measures [5]

- ✚ **Step Six:** The forecast method is used to predict the values of the dependent variables after their occurrence during the forecast period. It is noted here that statistical methods can be applied to create a reliable level of analysis.
- ✚ **Seventh Step:** Incorporating the effect of information on internal and external factors.
- ✚ **Step Eight:** Follow up on the results of applying the forecast method by recording the actual performance and observing the forecast error.

Forecasting Methods and Models:

1. Quantitative Methods:

Quantitative forecasting methods depend on the use of past data to predict the future, and this is consistent with the popular saying, [6].

A. Time Series Analysis:

The main objective of time series analysis is to identify the changes of the time series and to deconstruct this series in order to make an accurate forecast.

B. Regression and Correlation Analysis Models:

Regression analysis is one of the basic statistical methods in predicting the behavior of economic phenomena in the long run. [6].

1. Simple Regression Analysis Models:

The relationship between the dependent variable and the independent variable is linear, and the regression Equation is as follows and as shown in the Figure (4).

$$Y = a + bx + \mu$$

- Y: the dependent variable.
- X: the independent variable.
- μ: the amount of error.
- a,b: coefficients.

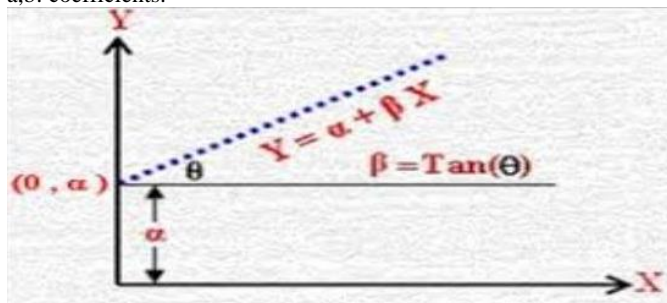


Figure (4) Simple Regression Analysis [6]

2. Simple Time Gap Regression Models:

It often occurs in economic life in the time mismatch between cause and effect, such as the slowing down of the effect of new equipment on the quantity of production, and this issue is called the time delay of the effect of one phenomenon on another, and when taking into account the amount of time units of slowdown, the simple regression model with time gap gives In the following form [7]:

$$\hat{y}_1 = a + bx_{t-L}$$

Where:

Y: the dependent variable.

X: the independent variable.

L: the amount of time units of deceleration (the length of the time gap).

t: time.

a, b: coefficients.

3. Multiple Regression Analysis Models:

Multiple linear regression is one of the advanced statistical methods, which ensures the accuracy of inference in order to improve research results through optimal use of data in finding causal relationships between the phenomena of the research subject, as shown in Figure (5) [8].

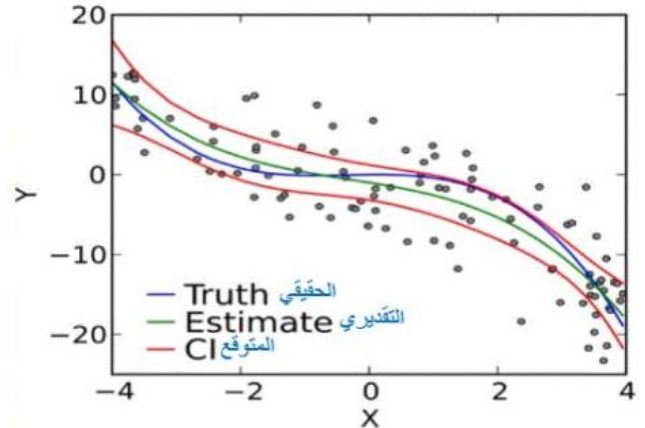


Figure (5) Multiple Regression Analysis [8]

The linear equation for multiple linear regression is [8]:

$$Y = a + B_1X_1 + B_2X_2 + \dots$$

Whereas:

Y: dependent variable

a: constant value

B₁: the slope of the y regression on the first independent variable

B₂: the slope of the y-gradient on the second independent variable

X₁: the first independent variable

X₂: The second independent variable

Multiple linear regression can be used if the following conditions are met [8]:

- ✚ The relationship between the independent variables and the dependent variable must be linear.
- ✚ The data should be normally distributed for the independent variables and the dependent variable.
- ✚ The values of the dependent variable must be of at least ordinal level.

After obtaining the results of the multiple regression equation, we must show whether these coefficients are statistically acceptable. In order to judge the significance of the regression coefficients, we use the test (T) and the level of probability corresponding to it. It is known that the statistical program (SPSS), which is the program of the statistical and social package, and the language (R) which is a software work environment for computer statistics that allows making statistical applications on the one hand and building statistical programs on the other hand. They will automatically extract a T-test and its corresponding probability level. It will also obtain statistics used to know the overall significance of the model and from it (R), (R₂). The first (R) is the simple correlation coefficient, which measures the strength of the relationship between two or more variables, while (R₂) is called the coefficient of determination, which is used to find out the explanatory power of the estimated model (estimated equation) in the case of simple linear regression (one independent variable with one dependent variable). [7].

2. Qualitative Methods:

It is a set of objective methods that are used to make a forecast of demand when historical data on demand is not available, and which depends on methods that invest the wisdom and experience possessed by management [9]

1.2 Problem of Study:

The LISCo is considered one of the largest industrial companies in Libya, and due to its high production costs in order to guarantee the quality of its products, which forced it to conduct many checks and mechanical tests for its products, starting with the casting process in the furnace factories, and ending with testing the mechanical specifications of its final products, and that In order to control the quality of its products according to international standards.

1.3 Research Hypotheses:

- There is no relationship between the independent variables and the dependent variable.
- There is at least one relationship between the independent variables and the dependent variable.

1.4 Importance of the Study:

The importance of forecasting is prominent in the institution, and its importance increases according to the increase in the activity of the institution and the volume of sales and production operations in it, as the forecast situation affects the plan that is developed by the company's management to determine the volume of activity. In order to ensure the quality of product , the company perform mechanical tests which result in consuming time, materials, labour force, which is result in high costs., therefore this study was proposed to examine the possibility of using the prediction method to save the large sums spent on the tests conducted.

1.5 Objective of the Study:

We seek through this study to reach the formulation of a model through which the mechanical specifications of the products of the longitudinal rolling mill of the LISCo can be predicted using the multiple linear regression method, and in a simple and accurate manner through which the quality of the products can be controlled within the specifications approved by the LISCo.

1.6 Study Methodology:

In order to achieve the objective of the study, the process of collecting data and information was relied on the following:

- Collecting data, recording, and observations through field visits.
- Applying demand forecasting using multiple linear regression method.
- Analyze the results that realize the aim of the study and my lead to further studies in the area.

2. The LISCo:

The LISCo in Libya is considered a relatively new industry. It is considered one of the largest industrial companies in Libya. It contains 10 production factories and 5 basic facilities. In addition to workshops, warehouses and laboratories, the company also has a wire drawing plant in Tripoli, where the annual design capacity of the company's factories is estimated at (1.32 million tons) of liquid steel and (1.1 million tons) of sponge iron, which has been increased to 1.7 million tons of liquid steel and (1.75 million tons) of hot-moulded sponge iron, ^[1].

2.1 Production Units:

The company includes the following production units ^[1]:

- ✚ Direct reduction factory (3 units).
- ✚ Steel factory (2,1).
- ✚ Rolling for bars and skewers factory (4 lines).
- ✚ Light and medium sectors rolling factory.
- ✚ Hot rolling factory.
- ✚ Cold rolling factory.
- ✚ Galvanizing line.
- ✚ Paint line.

2.2 Supporting Units:

The company includes many auxiliary units and facilities, the most important of which are ^[1]:

- ✚ Port and storage yard for pellets.
- ✚ Electricity and water desalination plant.
- ✚ An oxygen and compressed air factory.
- ✚ A lime factory.
- ✚ Central workshop.
- ✚ A training center.
- ✚ Quality control laboratories.
- ✚ A resort and a sea club.

In the context of the company's concern and its tireless quest to

respond to the customer's demands, and to impose its position in the international markets among the international iron and steel companies, it has worked to adopt the International Quality Standard (ISO 9001/2000) system and formed a Quality Council for it, to ensure the effectiveness of its application, and to achieve the desired goals, which qualified The LISCo was awarded the 12th European International Quality Award in 1998. It also obtained the Certificate of Implementing the Total Quality System 9001/2000 (ISO) on March 25, 2002.

2.3. Factories:

2.3.1. New Bar Rolling Factory:

The factory includes one new line for the production of bars with a design capacity of (800 thousand tons) of steel reinforcement, sizes (8 mm to 40 mm), where the factory uses bars of size (130 mm-150 mm). It is worth mentioning that those who operated the factory are national cadres within a month June of the year 2017, and Figure (6) shows the Rolling Bars Factory ^[1].



Figure (6) Bar Rolling Production Factory ^[1]

2.3.2. Hot Strip Rolling Factory:

The annual design capacity of the factory is (580,000 tons) of hot rolled coils, which are shown in Figure (7), and the commercial operation income is on 05/10/1990.



Figure (7) Products of the Hot Strip Rolling Factory ^[1]

2.3.3. Cold Rolling Factory:

The commercial operation of the factory began on January 20, 1990, with an annual design production capacity of (140,000 tons) of cold-rolled coils, as shown in Figure (8).



Figure (8) Products of the Cold Strip Rolling Factory ^[1]

2.3. Mechanical Properties of the Material:

The concept of mechanical properties of a material is how it behaves

when loaded, where the elasticity factor of a material affects the amount of deflection under load, and the strength of a material determines the stresses it can withstand before it breaks, where the ductility of a material also plays an important role in determining when a material breaks when it is loaded beyond the limits Their flexibility, and because every mechanical system is subjected to loads during operation, it is important to understand how the materials that make up those mechanical systems behave^[2].

3. Study Status:

The importance of forecasting is due to the fact that the company's long-term existence depends on the existence of a continuous demand for its goods or services, and this demand is linked in some way to the general level of economic activity. All management activities must be planned in advance, as well as all management decisions must be anticipated in the light of forecasts future prospects for this activity. And we seek through this study to reach models through which to predict the mechanical specifications of the products of the longitudinal rolling mill of the LISCo, using the multiple regression method, and in a simple and accurate way through which the quality of the products can be adjusted according to the specifications approved by the LISCo, where we took 198 A sample from the LISCo, to prove the accuracy of the prediction results and compare them with the real results of each of the yield stress, tensile strength and elongation ratio. The stratified sample was selected which is groups (DRI reduced iron group, FeMn group, Ferro silicon group FeSi group, Calcium oxide Lime group, Calcium fluoride group DOL, Carbon group C, Silicon group Si, Manganese group Mn), due to the availability of 198 samples. With great accuracy, Figure (9) shows the samples used in preparing and testing the prediction model.

NO	DRI	FeMn	FeSi	Lime	Dol.	C	Si	Mn	Yield Effort	Tensile Strength	Elongation
1	0.93647	0.00745	0.00287	0.04363	0.00319	34	23	81	448	747	20
2	0.93779	0.00729	0.00281	0.04897	0.00313	31	19	75	429	668	17
3	0.66688	0.00729	0.00281	0.04168	0.00313	30	20	74	431	681	18
4	0.93579	0.00753	0.0029	0.0484	0.00323	32	21	70	450	673	21
5	0.95863	0.00729	0.00281	0.0521	0.00313	30	22	72	475	771	15
6	0.9244	0.00886	0.00342	0.05065	0.0038	30	19	80	384	619	20
7	0.93579	0.00753	0.0029	0.04302	0.00323	31	18	77	412	653	19
8	0.59473	0.04248	0.02294	0.08496	0	29	22	76	457	700	16
9	0.84966	0.00959	0.0037	0.05482	0	32	21	75	382	583	26
10	0.9289	0.00834	0.00322	0.05955	0.00357	29	23	79	452	732	16
11	0.75023	0.007	0.0027	0.05002	0.003	33	24	70	428	692	17
12	0.8651	0.00631	0.00243	0.03514	0.0027	30	26	74	442	717	20
13	0.94875	0.00722	0.00278	0.04125	0.00309	32	25	76	419	650	19
14	0.94087	0.00693	0.00267	0.04457	0.00297	35	18	74	442	712	22
15	0.69409	0.00715	0.00276	0.0347	0.00306	33	26	70	444	729	17
16	0.9329	0.00787	0.00303	0.0562	0.00337	34	23	72	394	651	20
17	0.890403	0.007695	0.002968	0.054963	0.003298	34	21	74	370	613	21
18	0.863461	0.009094	0.003508	0.045472	0.003898	32	19	68	412	622	21
19	0.916953	0.007294	0.002813	0.042722	0.003126	31	22	71	408	673	25

Figure (9-a) Samples Used In Preparing and Testing the Prediction Model.

20	0.94711	0.00745	0.00287	0.01383	0.00319	31	22	72	418	653	18
21	0.93906	0.00715	0.00276	0.01429	0.00306	32	21	75	437	697	17
22	0.91695	0.00729	0.00281	0.01563	0.00313	31	21	70	394	639	20
23	0.60376	0.01208	0.00466	0.05175	0.00618	29	21	86	433	655	26
24	0.98945	0.00761	0.00294	0.02936	0.00326	31	20	77	374	564	21
25	0.98957	0.00753	0.0029	0.03012	0.00323	35	25	75	410	658	20
26	0.65127	0.01018	0.00275	0.05495	0.00305	34	14	72	387	595	22
27	0.96618	0.02196	0.01186	0.24155	0.01318	37	18	66	379	593	21
28	0.9711	0.00819	0.00316	0.0585	0.00351	30	19	69	370	554	26
29	0.93779	0.00729	0.00281	0.04168	0.00313	35	21	74	388	609	20
30	0.93594	0.0113	0.00436	0.04841	0.00484	35	25	73	380	626	20
31	0.98802	0.00865	0.00333	0.0457	0.00371	25	11	66	365	546	24
32	0.98989	0.00729	0.00281	0.0521	0.00313	32	20	74	403	651	19
33	0.95683	0.00761	0.00294	0.05437	0.00326	29	18	81	399	597	18
34	0.98934	0.00769	0.00297	0.05496	0.0033	35	14	75	409	676	16
35	0.78927	0.01036	0.00266	0.04933	0.00296	34	19	77	376	608	21
36	0.73921	0.00761	0.003153	0.04383	0.003261	31	15	98	358	559	22
37	0.98953	0.007292	0.003125	0.041667	0.003125	43	15	73	401	633	21

Figure (9-b) Samples Used In Preparing and Testing the Prediction Model.

38	0.70409	0.00761	0.00285	0.04757	0.00285	26	23	62	444	627	16
39	0.98612	0.01053	0.00335	0.04787	0.00287	26	18	68	343	537	23
40	0.97166	0.02227	0.00607	0.08097	0.00607	25	18	75	393	573	24
41	0.34568	0.02963	0.00741	0.07407	0.00741	28	17	78	371	556	22
42	0.98644	0.00986	0.0037	0.04932	0.0037	24	12	98	406	660	19
43	0.70323	0.01125	0.00422	0.07032	0.00422	33	18	1,01	407	681	26
44	0.98806	0.00869	0.00326	0.04886	0.00326	35	16	87	400	626	20
45	0.68985	0.01054	0.00259	0.04791	0.00287	38	2	84	426	664	18
46	0.98872	0.00752	0.00376	0.05373	0.00322	31	24	72	350	542	23
47	0.98684	0.00987	0.00329	0.05482	0.00329	36	19	84	326	499	25
48	0.97959	0.01633	0.00408	0.06803	0.00408	33	17	109	366	565	22
49	0.95238	0.03741	0.0102	0.10204	0.0102	33	21	76	364	584	22
50	0.96691	0.00859	0.00301	0.05372	0.00322	24	26	74	348	537	22
51	0.98253	0.01365	0.00382	0.05459	0.00328	37	21	87	385	608	23
52	0.97986	0.01633	0.00381	0.05444	0.00327	33	18	91	397	598	20
53	0.98793	0.00878	0.00329	0.05488	0.00329	29	13	90	411	626	23

Figure (9-c) Samples Used In Preparing and Testing the Prediction Model.

54	0.98446	0.01244	0.00311	0.05181	0.00311	32	24	79	405	636	19
55	0.98446	0.01244	0.00311	0.05181	0.00311	38	22	78	392	630	20
56	0.98691	0.01007	0.00302	0.05035	0.00302	40	18	80	384	619	20
57	0.9838	0.01273	0.00347	0.05787	0.00347	38	17	89	413	679	25
58	0.98542	0.01106	0.00352	0.06033	0.00302	29	14	88	397	633	20
59	0.98571	0.01084	0.00345	0.04929	0.00296	36	20	82	376	591	27
60	0.98196	0.01494	0.0031	0.05337	0.0032	33	12	89	355	632	25
61	0.98214	0.01488	0.00298	0.05852	0.00298	28	17	97	329	500	25
62	0.86735	0.00714	0.00306	0.05612	0.00306	35	20	101	424	637	20
63	0.76684	0.01244	0.00311	0.05181	0.00311	38	21	81	429	629	19
64	0.9768	0.01954	0.00366	0.06105	0.00366	33	18	91	390	631	20
65	0.97622	0.02003	0.00375	0.06258	0.00375	33	18	81	409	645	24
66	0.98874	0.00783	0.00343	0.04895	0.00294	35	29	83	417	652	18
67	0.98307	0.01365	0.00328	0.05461	0.00328	28	21	80	381	566	22
68	0.9781	0.01752	0.00438	0.05839	0.00438	32	11	110	380	578	22
69	0.78254	0.00906	0.00247	0.04119	0.00247	35	21	97	407	610	20
70	0.768443	0.01332	0.003074	0.05123	0.003074	35	18	92	422	650	19
71	0.795229	0.012922	0.002982	0.049702	0.002982	34	21	89	413	627	19

Figure (9-d) Samples Used In Preparing and Testing the Prediction Model.

72	0.72131	0.01311	0.00328	0.05464	0.00328	31	20	89	399	594	21
73	0.9865	0.01038	0.00312	0.05192	0.00312	33	16	96	429	680	19
74	0.63102	0.01283	0.00321	0.05348	0.00321	24	19	72	415	628	20
75	0.73892	0.01182	0.00296	0.04926	0.00296	33	13	96	411	629	20
76	0.98305	0.01356	0.00339	0.0678	0.00339	34	22	110	438	730	17
77	0.73988	0.01387	0.00347	0.0578	0.00347	33	17	92	415	696	19
78	0.65502	0.01419	0.00328	0.06004	0	31	14	86	359	545	22
79	0.72821	0.01231	0.00308	0.05128	0.00308	30	111	100	429	663	20
80	0.55894	0.01118	0.00305	0.05081	0.00305	36	19	94	413	655	19
81	0.98912	0.00791	0.00297	0.04946	0.00297	33	17	92	413	633	20
82	0.9894	0.00757	0.00303	0.04139	0	29	19	79	401	643	19
83	0.98958	0.00729	0.00313	0.04167	0.00313	38	19	85	407	635	19
84	0.7448	0.01095	0.00329	0.04381	0.00329	32	12	66	433	668	19
85	0.76004	0.00869	0.00326	0.04343	0.00326	33	25	79	386	566	22
86	0.98649	0.01085	0.00286	0.04932	0	30	20	83	412	611	19
87	0.86482	0.03186	0.01229	0.22758	0.01365	27	21	87	364	557	23
88	0.98937	0.007695	0.002968	0.054963	0.003298	31	22	81	383	657	20
89	0.989678	0.007449	0.002873	0.053208	0.003193	28	24	84	494	535	20

Figure (9-e) Samples Used In Preparing and Testing the Prediction Model.

90	0.98633	0.01178	0.00289	0.04284	0.00321	34	21	81	380	603	20
91	0.98979	0.00737	0.00284	0.04212	0.00316	35	18	76	425	738	18
92	0.98468	0.01204	0.00328	0.04376	0.00328	33	20	92	413	657	21
93	0.93458	0.0113	0.00277	0.05135	0.00308	36	18	78	384	601	20
94	0.98548	0.01166	0.00286	0.05298	0.00318	30	25	81	386	586	20
95	0.98501	0.01178	0.00321	0.05353	0.00321	33	15	83	376	599	21
96	0.70855	0.00729	0.00281	0.0521	0.00313	32	23	81	400	702	19
97	0.98979										

Table of 161 samples used in preparing and testing the prediction model. Columns include ID, and multiple columns of numerical values.

Figure (9-j) Samples Used In Preparing and Testing the Prediction Model.

Table of 178 samples used in preparing and testing the prediction model. Columns include ID, and multiple columns of numerical values.

Figure (9-k) Samples used in preparing and testing the prediction model.

Table of 198 samples used in preparing and testing the prediction model. Columns include ID, and multiple columns of numerical values.

Figure (9-m) Samples used in preparing and testing the prediction model.

3.1. Forecasting Models:

To use the prediction equation for multiple linear regression analysis, it is necessary to know whether the independent variables have an effect on the dependent variable or not, and for this we calculated the correlation rate R, which shows if there is a relationship between the independent variables and the successive variable.

3.1.1. Yield Effort Prediction Model:

After confirming the existence of a correlation, we can use the prediction equation to analyze the multiple linear regression, and Figure (10) shows the results of the multiple linear regression analysis and the statistical analysis of the yield effort prediction model.

Figure (10) Results of the multiple linear regression analysis with the yield stress prediction model. Includes SUMMARY OUTPUT, Regression Statistics, ANOVA, and Coefficients tables.

Figure (10) Results of the multiple linear regression analysis with the yield stress prediction model

Through Figure (11), it is necessary to know whether there is a relationship between the independent variables (chemicals) and the dependent variable (submission stress), by means of the correlation coefficient R where its value was 0.57 and this confirms the existence of a strong correlation between the independent variables and the dependent variable, while R2 The coefficient of determination indicates the influence of the independent variables on the dependent variable, where its value was 0.32, and this indicates the strength of

the effect. As for Adjusted (R2), the corrected coefficient of determination is the same as R2, but they differ from each other that Adjusted (R2) increases only when adding an independent variable that has an effect on the dependent variable and after knowing that there is a relationship, we test the hypothesis by tabular F test where if F The tabular F selects the null hypothesis H0: there is no relationship between the independent variables and the dependent variable. And if the tabular F > F selects the hypothesis H1: there is at least one relationship between the independent variables and the dependent variable, and this test is considered a comprehensive test, so we conduct another test to find out which of the independent variables have an effect on the dependent variable individually by T and P test -Value, where if the value of T > 2, then this independent variable has an effect on the dependent variable, and if T < 2, then this variable has no effect, and as for the P-Value test, if P-Value > 0.05, this variable has no effect on the dependent variable However, if P-Value < 0.05, this independent variable has an effect on the dependent variable.

3.1.2. Results of the Yield Effort Prediction Model:

These results are 42 samples obtained from the yield stress prediction model and were compared with the real stress as shown in Figure (11), where it was found that the mean error is close to zero.

Figure (11-a) results of the yield stress prediction model and its comparison with the real stress. Table with columns: OBSERVATOR, PREDICTED YIELD EFFORT, RESIDUALS, STANDARD RESIDUALS, PERCENTILE, and YIELD EFFORT.

Figure (11-a) results of the yield stress prediction model and its comparison with the real stress.

Figure (11-b) results of the yield stress prediction model and its comparison with the real stress. Table with columns: OBSERVATOR, PREDICTED YIELD EFFORT, RESIDUALS, STANDARD RESIDUALS, PERCENTILE, and YIELD EFFORT.

Figure (11-b) results of the yield stress prediction model and its comparison with the real stress.

Figure (11-c) results of the yield stress prediction model and its comparison with the real stress. Table with columns: OBSERVATOR, PREDICTED YIELD EFFORT, RESIDUALS, STANDARD RESIDUALS, PERCENTILE, and YIELD EFFORT. Includes Total Error and Average Error.

Figure (11-c) results of the yield stress prediction model and its comparison with the real stress.

samples, the total prediction for the yield effort was 17117, and the average error was -8.12049E-14, which is very close to zero, which indicates the accuracy of the prediction. As for the negative sign, this indicates that the prediction is greater than the real one. We subtract this value from all predictions until we get an average error equal to zero, and the average error is the result of subtracting the prediction from the real one, adding all the values of the subtraction process and dividing it by the number of samples (42 samples). Prediction was performed for only 42 samples because Microsoft Office Excel did not accept more than this size.

3.1.3. Tensile Strength Prediction Model:

After confirming the existence of a correlation, we can use the prediction equation to analyze the multiple linear regression, and Figure (12) shows the results of the multiple linear regression analysis of the tensile strength prediction model.

SUMMARY OUTPUT						
Regression Statistics						
Multiple R	0.997128759					
R Square	0.994265762					
Adjusted R Square	0.963873418					
Standard Error	54.22239823					
Observations	42					
ANOVA						
	df	SS	MS	F	Significance F	
Regression	8	17332575.67	2166571.959	736.9120758	6.57E-35	
Residual	34	99962.32797	2940.66847			
Total	42	17432638				
Coefficients						
		Standard Error	t Stat	P-value	Lower 95%	Upper 95%
Intercept	64.0203392	63.8262404	1.003844811	0.3229264	-65.69012	193.7307829
DRI	2212.043288	3952.269399	-0.5596994	0.579360193	-10244.02	5819.934498
FaSi	6357.878607	8495.264275	0.748402687	0.459363378	-10906.58	23622.33279
Lime	-119.6940869	584.7570177	-0.204690296	0.839034554	-1308.063	1068.675152
Dol.	-1232.957154	8756.876433	-0.140798738	0.888898825	-19029.07	16563.1569
C	4.216962061	2.289887732	1.841557628	0.074277936	-0.436652	8.878573884
Si	10.05367571	2.294171246	4.382289079	0.00010691	5.391359	14.71599663
Mn	3.469175352	0.920850068	3.766582749	0.000620895	1.59758	5.340771111

Figure (12) results of the multiple linear regression analysis of the tensile strength prediction model.

Through the results, it must be known whether there is a relationship between the independent variables (chemicals) and the dependent variable (tensile strength), through the correlation coefficient R, where its value was 0.997, and this confirms the existence of a strong correlation between the independent variables and the dependent variable. As for the R2 coefficient of determination, it indicates the effect of the independent variables on the dependent variable, where its value was 0.9994, and this indicates the strength of the effect, and the Adjusted (R2) coefficient of determination is the same as the meaning of R2, but they differ from each other that Adjusted (R2) increases only when adding an independent variable It has an effect on the dependent variable and after knowing that there is a relationship, we test the hypothesis by using the F tabular test where if $F < F$ tabular selects the null hypothesis H0: there is no relationship between the independent variables and the dependent variable. And if the tabular $F > F$ selects the hypothesis H1: there is at least one relationship between the independent variables and the dependent variable, and this test is considered a comprehensive test, so we conduct another test to find out which of the independent variables have an effect on the dependent variable individually by T and P test -Value, where if the value of $T > 2$, then this independent variable has an effect on the dependent variable, but if $T < 2$, then this variable has no effect, and as for the P-Value test, if $P\text{-Value} > 0.05$, this variable has no effect on the dependent variable, If the $P\text{-Value} < 0.05$, this independent variable has an effect on the dependent variable.

3.1.4. Results of the Tensile Strength Prediction Model:

These results are 42 samples obtained from the tensile strength prediction model and were compared to the real tensile strength as shown in Figure (13), where it was found that the mean error is close to zero.

RESIDUAL OUTPUT			
Observation	Predicted قوة الشد	Residuals	Standard Residuals
1	708.1988264	38.80117357	0.795337254
2	634.0081868	33.99181317	0.696756125
3	619.9045078	61.09549221	1.252320911
4	640.8586352	32.14136484	0.658826074
5	650.5047386	120.4952614	2.469883294
6	645.6264589	-26.62645891	-0.545782841
7	631.4085979	21.59140206	0.442575439
8	686.913766	13.08623401	0.268238521
9	656.3976541	-73.39765409	-1.504487708
10	677.5308601	54.46913993	1.116495513
11	663.3158914	28.68410864	0.587960057
12	693.9831872	23.01681284	0.471793171
13	703.6638175	-53.66381748	-1.099988205
14	638.1818323	73.81816773	1.513107297
15	681.6208495	47.37915054	0.971166592
16	675.1227637	-24.12276372	-0.49446269
17	659.4309267	-46.43092666	-0.951730124
18	610.3628864	11.63711355	0.238534794
19	649.7066605	23.29333951	0.477461349
20	658.5194648	-5.519464804	-0.113136681

Figure (13-a) Results of the tensile strength prediction model and their comparison with the real tensile strength.

RESIDUAL OUTPUT			
Observation	Predicted قوة الشد	Residuals	Standard Residuals
21	662.6081195	34.39188054	0.704956612
22	639.4265376	-0.426537592	-0.008743066
23	660.7476296	-5.747629617	-0.117813549
24	656.5640492	-92.56404919	-1.897355929
25	716.7018275	-58.70182751	-1.20325614
26	560.2204891	34.77951092	0.712902167
27	609.4748475	-16.4748475	-0.337697517
28	609.7101182	-55.71011821	-1.141932792
29	668.3872533	-59.38725332	-1.217305835
30	703.0552519	-77.05525189	-1.57946028
31	500.2729844	45.72701557	0.937301521
32	647.7709228	3.229077211	0.066188859
33	636.8178465	-39.81784651	-0.816176775
34	603.0751032	72.92489683	1.494797242
35	636.5196068	-28.51960678	-0.584588137

Figure (13-b) Results of the tensile strength prediction model and their comparison with the real tensile strength.

RESIDUAL OUTPUT			
Observation	Predicted قوة الشد	Residuals	Standard Residuals
36	662.8197844	-103.8197844	-2.128073321
37	643.6364033	-10.63640333	-0.218022473
38	593.1373675	33.86263247	0.69410821
39	578.3803834	-41.38038338	-0.848205286
40	580.9595089	-7.959508881	-0.163152126
41	545.2677191	10.73228094	0.21998775
42	616.2162295	43.78377048	0.897469344
Total Error =		34.96950514	
Average Error =		0.832607265	
26917.03049			
Average Tensile Strength = 640.881678			

Figure (13-c) Results of the tensile strength prediction model and their comparison with the real tensile strength.

In Figure (13), after performing the prediction process for 42 samples, the average error was 0.832, which is very close to zero, which indicates the accuracy of the prediction.

3.1.5. Elongation Ratio Prediction Model:

After confirming that there is a correlation, we can use the prediction equation to analyze the multiple linear regression, and Figure (14) shows the results of the multiple linear regression analysis and the statistical analysis of the elongation ratio prediction model.

Table with 18 columns and 63 rows of numerical data for Figure 17-d.

Figure (17-d) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 18 rows of numerical data for Figure 17-k.

Figure (17-k) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 63 rows of numerical data for Figure 17-e.

Figure (17-e) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 18 rows of numerical data for Figure 17-m.

Figure (17-m) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 63 rows of numerical data for Figure 17-f.

Figure (17-f) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 18 rows of numerical data for Figure 17-n.

Figure (17-n) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 63 rows of numerical data for Figure 17-g.

Figure (17-g) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 18 rows of numerical data for Figure 17-o.

Figure (17-o) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 63 rows of numerical data for Figure 17-h.

Figure (17-h) Testing the prediction model and calculating the mean error in predicting the tensile strength.

Table with 18 columns and 18 rows of numerical data for Figure 18-a.

Figure (18-a) Testing the prediction model and calculating the mean error in predicting the elongation ratio.

Table with 18 columns and 21 rows of numerical data for Figure 18-b.

Figure (18-b) Testing the prediction model and calculating the mean error in predicting the elongation ratio.

Subjection	Quality Mechanical Products		Quality of Elongation Ratio
	Stress	Tensile Strength Quality	
OK		OK	OK
OK		OK	OK
OK		OK	OK
OK		OK	OK
OK		OK	OK
OK		OK	OK
OK		OK	OK
OK		OK	OK
OK		OK	OK
OK		OK	OK

Figure (30) Matching Model.

Conclusion:

From the results and prediction models of this study, we can conclude the following:

1. The LISCo's lack of knowledge of the importance of prediction models for mechanical properties that reduce uncertainty.
2. Through the use of the model in forecasting, prediction values were obtained that are very close to the real results of the mechanical properties of longitudinal rolling iron for iron and steel plant.
3. Obtaining an error rate close to zero in the process of predicting the mechanical properties of longitudinal rolling iron for the iron and steel plant, and this increases its accuracy.
4. Such model lead to save time and contribute in cost reduction as compared to the traditional tests that used to ensure the quality of mechanical properties.
5. The possibility of modifying the chemical composition of the mixture to obtain the required mechanical specifications before the manufacturing process and the production of out-of-spec products.

Recommendations:

Through this research, we can recommend the following points:

1. Emphasis on the importance of applying the forecasting process in decision-making as a method for building the future and reducing costs for the organization.
2. Emphasis on the importance of statistical modeling using forecasting modeling techniques in calculating expectations.
3. We recommend paying attention to databases and previous statistical information in order to serve the planning process satisfactorily.
4. The higher the efficiency of the predictive operations of the LISCo, the closer the picture will be to the desired reality requested by the institution.
5. Developing statistical and planning training methods and methodology for preparing plans and formulating decisions in order for the LISCo to keep pace with recent developments in the administrative, economic and social sciences.
6. The use of programs for the forecasting process, such as the mechanical specifications of the company's products in various other products, which allows the company to increase its production by reducing the percentage of rejected products that are out of specifications.
7. The use of this model, which was prepared to predict the mechanical specifications of the company's longitudinal rolling products, and whose predictive ability has been proven through that the average error is almost equal to zero, before the manufacturing process, which leads to an increase in the company's productivity, reduces the percentage of rejected and raises the quality of its products.

References:

- [1]- الوحدات الإنتاجية والمصانع الرئيسية والفرعية، مجلة الصلب الليبي، إصدار الشركة الليبية للحديد والصلب، أغسطس 2020 م.
- [2]- الدكتورة عبير عتوم، الخصائص الميكانيكية للمادة، دار النشر العلوم والفيزياء، الأردن، 20 أكتوبر 2020 م.

- [3]- وليد خليفة، منحنى الإجهاد والانفعال، مقاومة الشد وإجهاد الخضوع ونسبة الاستطالة، مركز المعرفة، لبنان، 2015 م.
- [4]- محمد عبيدات، هاني الضمور، شفيق حداد، إدارة المبيعات والتنبؤ، دار وائل، عمان، الأردن، 2013 م.
- [5]- عمر صخري، اقتصاد المؤسسة، نماذج التنبؤ، ديوان المطبوعات الجامعية، الجزائر، 2016 م.
- [6]- إيهاب الحسن، أنواع نماذج التنبؤ الكمية والوصفية، قسم الإدارة الصناعية، كلية الإدارة والاقتصاد، جامعة بابل، العراق، 2011 م.
- [7]- سمرين سمير خليل، استخدام نموذج التنبؤ في تحليل تداول أسهم البنوك المدرجة في بورصة عمان، كلية الأعمال، جامعة الشرق الأوسط الأردن 2009 م.
- [8]- أحمد سعودي، تحليل الانحدار الخطي المتعدد، كلية العلوم الإنسانية والاجتماعية، جامعة محمد بوضياف، المسيلة، الجزائر 2019 – 2020 م.
- [9]- طهراوي فريد، نمذجة الظواهر الاقتصادية، اقتصاد كمي، جامعة البويرة، الجزائر 2020 م.
- حنان بن عوالي، التنبؤ بالطلب، كجزء مكمل من التخطيط الاستراتيجي، كلية العلوم الاقتصادية والتجارية، جامعة الشلف، الجزائر 2017 م.