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Forecasting the Main Indicators of Insurance Companies

Abstract. Introduction. *Econometric and adaptive models make it possible to predict financial and economic indicators in the short and long term. The most common forecasting models are linear trend models, adaptive Brown, Holt, Holt-Winters, Box-Jenkins, autoregressive and other models. It has been proved that the use of adaptive forecasting models becomes especially relevant in the context of constant changes in the external environment, instability of the economic and political situation.*

Purpose. *The purpose of this article is to substantiate the expediency of using forecasting methods when planning the development of the insurance market and to implement the procedure for forecasting the main indicators of its development using modern methods and techniques.*

Results. *Improving the efficiency of the insurance market is facilitated by the correct organization of its planning and the direct implementation of the planned indicators. Optimality of planning is determined by the degree to which the accuracy of the predicted level of planned indicators is achieved. The methodology and results of the forecast of insurance payments made for the near future can be taken as a basis for drawing up current and strategic plans of insurance companies.*

Conclusions. *It has been established that one of the barriers to the effective development of the insurance market in general and insurance companies in particular is the insufficient level of planning of their activities, especially in terms of forecasting key indicators. The procedure for forecasting the receipt of insurance payments was implemented using modern forecasting methods. The effectiveness of the Brown's adaptive model for short-term planning of insurance premiums is proved. The proposed model was tested for adequacy, on its basis, recommendations were developed for further application in the practice of insurance companies.*

Keywords: *insurance; insurance market; planning; forecasting; econometric model; adaptive model; trend extrapolation.*

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Прогнозування основних показників діяльності страхових компаній

Анотація. *У процесі людської діяльності в фінансово-економічному просторі є певні ризики. З метою їх мінімізації приймається рішення щодо диверсифікації ризику або використання механізмів страхування. Страхування – це ефективний інструмент для зменшення невизначеності однієї сторони, яку називають страхувальником, шляхом передачі особливих ризиків іншій стороні, яку називають страховиком, і яка пропонує відновлення, принаймні частково, економічних збитків, що зазнав страхувальник. Виявлено, що через низький рівень довіри населення до ринку страхування загалом та страхових компаній зокрема, через відсутність знань у галузі страхування, попит на страхові послуги в Україні є достатньо низьким.*

Доведено, що в умовах мінливості зовнішнього середовища в складній соціально-економічній та політичній ситуації в країні ефективність розвитку страхового ринку залежить від оптимальної моделі його планування. Об'єктами планування можуть виступати основні статистичні показники діяльності страхових компаній: валові страхові платежі, валові страхові виплати, страхові платежі, сплачені на перестраховання, загальні активи страхових компаній, власний капітал, грошові кошти, обсяги сформованих страхових резервів, довгострокові фінансові

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інвестиції, поточні фінансові інвестиції тощо. Обґрунтовано, що правильна організація планування та виконання планів забезпечують досягнення страховими компаніями цілей і задач розвитку. Результативність панування досягається за рахунок точності прогнозного рівня планованих показників.

У статті досліджено окремі аспекти прогнозування основних показників діяльності страхових компаній. Розкрито сутність методів та прийомів прогнозування і визначено їх роль у діяльності страхових компаній. Методологічною основою дослідження обрано економетричні методи та методи екстраполяції тенденцій. В якості прийомів прогнозування використано економетричні лінійні моделі та базові адаптивні моделі. Визначено переваги застосування у практичній діяльності розглянутих методів прогнозування. Запропоновані моделі протестовані на адекватність та зроблені висновки щодо можливості їх застосування у практиці планування обсягів надходжень страхових платежів страхових компаній України.

Ключові слова: страхування; страховий ринок; планування; прогнозування; економетрична модель; адаптивна модель; екстраполяція тенденцій.

Formulation of the problem. The insurance market has a significant impact on the sustainable development of the economy in general and its regions in particular. Acting as an important element of the country's financial system, it is the basis for redistribution and risk reduction. One of the barriers to the effective development of the insurance market is the insufficient level of its planning. The objects of planning the development of the insurance market can be the main statistical indicators: gross insurance payments, gross insurance payments, insurance payments paid for reinsurance, total assets of insurance companies, equity, cash, the volume of insurance reserves, long-term financial investments, current financial investments, etc. The effectiveness of planning is determined by the degree of accuracy of the predicted indicators and their comparability with actual data.

Analysis of recent research and publications. A large number of scientific works of domestic and foreign researchers are devoted to forecasting the performance of the insurance market. Much attention is paid to the practical application of forecasting models in the works of such authors as O. Amosha, N. Bagrov, S. Bozhko, V. Burkinsky, I. Vakhovych, V. Geets, B. Danilishin, Y. Ershov, V. Leontiev, E. Libanova, A. Loch, K. Lewis, A. Melnik, D. Morell, A. Wrights, T. Saati, B. Spencer, G. Tale, D. Winchirs, V. Shevchuk, N. Shabranska, B. Shchukin, E. Janch.

A significant number of leading scientists focused on key issues and general trends in the development of the insurance market, including forecasting its indicators, both in Ukraine and abroad, in particular O. Abakumenko, V. Bazylevych, V. Berg, A. Wagner, N. Vnukova, T. Govorushko, Y. Gorodnichenko, O. Gvozdenko, A. Duka, Y. Zhuravlyov, O. Zolotareva, M. Mnykh, R. Pikus, E. Prokopchuk, G. Puriy, Y. Romanovska, M. Rothschild, P. Samuelson, L. Sarana, T. Saricheva, O. Sukach, V. Turko, V. Fedorenko, V. Furman, V. Shakhov, A. Sholoyko and many others.

Noting the unconditional advantages of the approaches proposed by the authors for planning the activities of the insurance market, it should be noted that further attention needs aspects aimed at forecasting the main indicators of its development.

Formulation of research goals. The purpose of the article is to substantiate the feasibility of using forecasting methods in planning the main indicators of the insurance

market and the implementation of the forecasting procedure based on the use of modern tools.

Presentation of the main research material. The development of insurance is based on the understanding of the need for protection against sudden danger. Modernization of social relations and the participation of various branches of human activity in economic activity, in turn, led to the expansion of insurance coverage. Insurance is an indispensable part of the economic and financial sector. The need for insurance is due to the need to manage the risks that occur in the lives of individuals and economic activities of organizations (legal entities).

The maintenance and development of insurance relations is extremely important for the economy of any socially oriented state. Proper organization of its insurance planning and forecasting contributes to the effective development of the insurance market. Planning as an operational process involves the sequential implementation of three stages: analysis of the current state, forecasting future values of targets and direct planning with mechanisms for their implementation and current adjustment [5].

Forecasting the main indicators of the development of the insurance market and insurance activities of companies aims to solve the following tasks:

- determination of an objective, reasonable future state of development taking into account the available resources
- determination of the volume of key indicators to ensure long-term development;
- determining the size of extreme resource needs (equity, cash) by establishing the smallest and largest of their value;
- identification of sources of resources (equity, cash) to ensure effective development;
- determination, taking into account forecast estimates, areas of resource use to meet development needs;
- identification of factors that affect the studied processes of the future.

One of the possible ways to solve the problem is, in our opinion, the use of statistical forecasting methods that allow obtaining the expected values of the characteristics of the studied indicators.

According to the available data on the volume of insurance payments during 2010-2019 (table 1), based on linear trend models and extrapolation methods, we will build a forecast for three years.

Table 1 **Volumes of insurance payments (contributions, premiums) in 2010-2019, UAH mln**

Years	Volumes of insurance payments (contributions, premiums), UAH mln	Years	Volumes of insurance payments (contributions, premiums), UAH mln
2010	23081,7	2015	29736,0
2011	22693,5	2016	35170,3
2012	21508,2	2017	43431,8
2013	28661,8	2018	49367,5
2014	26767,3	2019	53001,2

Source: compiled by the authors for [2, 6, 7]

Using the application software - analysis package "Data Analysis" we performed a regression analysis of the linear trend model, the results of the analysis are summarized in Table 2.

Table 2 **Assessment of the adequacy of the linear trend model**

Regression statistics	Mean
Multiple correlation coefficient	0,87
Multiple coefficient of determination	0,76
F-criterion	24,84
Standard error	6153,37
Number of observations	10

The results of the calculations given in the table allow us to draw conclusions about the adequacy of the constructed model. According to the Chaddock scale, there is a close relationship between the factors (multiple correlation coefficient is 0,87). The coefficient of determination shows that the variation in the volume of insurance payments by 76% is determined by the time factor, which also determines the accuracy of the regression (compliance of the obtained regression equation with empirical data). The influence of factors not taken into account in the model is 24%.

The linear trend model of the volume of insurance payments is as follows: $Y=13738,41+\phi 3376,53 \cdot t$. The parameter $a_0=13738,41$ characterizes the zero level of the time series, the parameter $a_1=3376,53$ indicates an increase in the amount of insurance payments over time. In particular, the obtained equation determines the annual growth of the studied indicator at the level of UAH 3376,53 million.

Based on the results obtained on the basis of the linear trend model, we determine the forecast values of the studied indicator for the next three years (Table 3).

Table 3 **Determination of theoretical and forecast values of insurance payments based on the model of the linear trend**

Year	Volumes of insurance payments (contributions, premiums), UAH mln	t	Theoretical mean of a number of dynamics $Y=13738,41+3376,53 \cdot t$
2010	23081,7	1	17114,94
2011	22693,5	2	20491,47
2012	21508,2	3	23868
2013	28661,8	4	27244,53
2014	26767,3	5	30621,06
2015	29736,0	6	33997,59
2016	35170,3	7	37374,12
2017	43431,8	8	40750,65
2018	49367,5	9	44127,18
2019	53001,2	10	47503,71
Forecast according to the linear trend model			
2020		50880,24	
2021		54256,77	
2022		57633,3	

Source: calculated by authors

Let's determine the forecast indicators of the volume of insurance payments and analyze a number of dynamics by extrapolation methods. The use of these methods is appropriate for small amounts of information. To determine

the forecast indicators, the elements of the series are divided into two periods: according to the first period, conditionally called "prehistory", a model is built on the basis of which the forecast for the second period is made:

Based on the data of the first six years (base year and the first 5 years), we calculate: the average absolute increase: $\bar{\Delta Y} = \frac{Y_k - Y_0}{k-1} = \frac{29736 - 23081,7}{6-1} = 1330,86$ UAH. mln. – shows that the average amount of insurance payments increases by this amount annually; the average annual growth rate: $\bar{K}_p = \sqrt[k-1]{\frac{Y_k}{Y_0}} = \sqrt[6-1]{\frac{29736}{23081,7}} = 1,05$ або 105% - shows

that the amount of insurance payments increases by 5% annually.

Next, we make a forecast for the next period. The obtained calculations are compared with the actual data. The forecast indicators of the volume of insurance payments, determined by the average absolute growth and the average growth rate are given in table 4.

Table 4 Projected values of insurance payments

Year	Actual value, UAH, mln.	Predicted value	
		by average absolute increase, UAH, mln.	by average growth rate, UAH, mln.
2010	23081,7	28098,16	28105,67
2011	22693,5	29429,02	29510,95
2012	21508,2	30759,88	30986,50
2013	28661,8	32090,74	32535,83
2014	26767,3	33421,6	34162,62

Source: calculated by authors

Based on the forecast, we calculate the long-term average annual growth rate based on the second period using the base year.

$$\bar{K}_p = \sqrt[k-1]{\frac{Y_k}{Y_0}} = \sqrt[6-1]{\frac{53001,2}{26767,3}} = 1,146 \text{ or } 114,6\%, \text{ that is, in the}$$

second period there is also an annual increase in revenues from insurance payments of 15%.

The forecast of receipts of insurance payments in 2020-2022 will make:

$$Y_{2020} = 53001,2 \cdot 1,146 = 60739,38 \text{ UAH, mln.}$$

$$Y_{2021} = 53001,2 \cdot (1,146)^2 = 69431,6 \text{ UAH. mln.}$$

$$Y_{2022} = 53001,2 \cdot (1,146)^3 = 80031,8 \text{ UAH, mln.}$$

Thus, the forecast of insurance payments is based on trends that have developed in the "prehistory". However, it should be noted that these methods (average absolute growth and growth rate) have certain disadvantages. The main one is that these methods are based on the extreme levels of the series and do not take into account intermediate levels, which are sometimes decisive in determining the trend of changes in indicators. This is the reason for the decrease in the accuracy of the forecast and a significant discrepancy between the results obtained by the linear trend model and extrapolation methods. Also, given that the insurance market is developing in conditions of uncertainty and constant change, the possibility of applying traditional models based on the extrapolation of stationary processes is significantly limited.

Permanent changes in the external environment are most successfully taken into account in short-term forecasting, where the important dynamics of the studied indicator at the end of the observation period, rather than the trend of its development on average during the study period. Therefore, there is a need for such predictive models that would reflect the qualitative changes that occur in the patterns of development of the studied processes. The situation is exacerbated by the lack of a priori information about the nature of such changes. Given

the above, we believe that the accuracy of forecast estimates can be ensured by adaptive models. Unlike other forecasting models, adaptive ones are able to take into account the evolution of dynamic characteristics when displaying the current state of the object under study [3].

Adaptive forecasting is developing in two directions: complicating the structure of forecasting models and improving adaptive mechanisms. At present, issues related to the problem of expanding the scope of application of the principles of adaptation to solving problems of long-term planning are becoming relevant. Adaptive models can be built using a small data set and are suitable for short-term forecasting, which is used in conditions of variability and instability of the environment.

Adaptive models and methods have in their structure a built-in mechanism that responds to changes in the studied indicator. Estimation of parameters of basic adaptive models is carried out on the first few observations; on its basis the forecast which is compared with actual data is made. In the next step, the model is adjusted according to the magnitude of the forecast error and then used to predict the next level; the procedure is repeated until all observations are exhausted.

Thus, the model at each step takes into account changes in the external and internal environment, adapts to changes and at the end of the period reflects the current trend. The forecast is the result of extrapolation of the latest trend [4].

We use a standard prediction algorithm by the method of double exponential Brown smoothing. Basic estimates of the model parameters are obtained by the method of least squares. To estimate the parameters in the Brown model, the prediction step $k=1$ and the smoothing parameter $\alpha=0,7$ were chosen. The optimal value of the discount parameter β was determined using numerical optimization methods, provided that it is in the range from 0 to 1. The results of the evaluation of the parameters of the Brown model are given in table 5.

Table 5 The results of the evaluation of the parameters of the Brown model

t	y(t)	a ₀	a ₁	$\hat{y}_p(t)$	$\varepsilon(t)$
		13738,41	3376,53	17114,94	
1	23081,7	20157,99	6419,58	26577,57	5966,76
2	22693,5	27700,61	7542,62	35243,23	2202,03
3	21508,2	34039,73	6339,12	40378,85	-2359,8
4	28661,8	41101,66	7061,93	48163,59	1417,27
5	26767,3	46198,17	5096,51	51294,68	-3853,76
6	29736	49121,27	2923,1	52044,37	-4261,59
7	24844,2	45654,12	-3467,15	42186,97	-12529,9
8	43431,8	43554,36	-2099,76	41454,6	2681,15
9	49367,5	44127,16	572,8	44699,96	5240,32
10	53001,2	47503,68	3376,52	50880,2	5497,49

Source: calculated by authors

Thus, for the last step we obtained an adaptive model of Brown, which has the form:

$$y_p(n+k)=47503,68+3376,52 \cdot k$$

To calculate the point forecast estimates for the following periods (2020-2022) in the above formula we substitute the value k=1, 2, 3:

2020 p.: $y_p(11)=47503,68+3376,52 \cdot 1=50880,2$ UAH, mln

2021 p.: $y_p(12)=47503,68+3376,52 \cdot 2=54256,72$ UAH, mln

2022 p.: $y_p(13)=47503,68+3376,52 \cdot 3=57633,24$ UAH, mln

Compliance of the model with the adequacy criteria determines its quality. Adequacy is characterized by compliance with statistical properties and the degree of proximity to actual data (accuracy). The model is considered adequate if the series of residues($\varepsilon(t)$) corresponds to the properties of randomness, independence of successive levels of the series and normality of the distribution [1]. The results of the study of the adequacy of the adaptive model for forecasting the volume of insurance payments are given in table 6.

Table 6 The results of the assessment of the adequacy of Browns model

T	y(t)	$\varepsilon(t)$	Pivot points	$\varepsilon(t)^2$	$\varepsilon(t-1)$	$(\varepsilon(t)-\varepsilon(t-1))^2$
1	23081,7	5966,76	-	35602224,9	-	-
2	22693,5	2202,03	1	4848936,121	5966,76	14173191,97
3	21508,2	-2359,8	0	5568656,04	2202,03	20810292,95
4	28661,8	1417,27	1	2008654,253	-2359,8	14266257,78
5	26767,3	-3853,76	0	14851466,14	1417,27	27783757,26
6	29736	-4261,59	0	18161149,33	-3853,76	166325,3089
7	24844,2	-12529,9	1	156998895,2	-4261,59	68365280,99
8	43431,8	2681,15	0	7188565,323	-12529,9	231376042,1
9	49367,5	5240,32	1	27460953,7	2681,15	6549351,089
10	53001,2	5497,49	1	30222396,3	5240,32	66136,4089
	Σ	-0,05	5	302911897,3	-	383556635,9

Source: calculated by authors

The randomness of the levels of a number of residues is checked on the basis of the criterion of turning points, according to which each level of the series is compared with two adjacent ones. Next, the sum of turning points is determined and its value is compared with the inequality:

$$p > \left[\frac{2 \cdot (N-2)}{3} - 1,96 \cdot \sqrt{\frac{16 \cdot N-29}{90}} \right] \quad [1]$$

According to our calculations, the sum of turning points is 5, which is greater than 3 (calculations for inequality). Therefore, the above inequality holds, and a number of

residues is considered to correspond to the property of chance.

The independence of the residues (absence of autocorrelation) is checked by the Darbin-Watson criterion, according to which the coefficient d is determined. According to the calculations, the value $d'=0,7$ ($d < 2$), which indicates the absence of autocorrelation and independence of the residues.

The conformity of the residues to the normal distribution law is determined by the R/S criterion, namely:

$R/S = \frac{\epsilon(t)_{\max} - \epsilon(t)_{\min}}{S}$ with critical levels that for $N=10$ correspond to the interval (2,75; 3,85). The value of R/S calculated by us is equal to 3,28 and falls into the specified interval, which indicates the compliance of a number of residues with the normal distribution law and the admissibility of constructing confidence intervals for the forecast.

Thus, the calculations obtained by the adaptive model of Brown allow us to conclude about the adequacy of the model and its suitability for practical use in real economic conditions.

In order to determine the effectiveness of the forecasting methods used and to select the best forecast, the results of the calculations are summarized in a common table (Table 7).

Table 7 **Estimated values of insurance payments are calculated using the methods used**

Years	Linear trend model	Methods of extrapolation of tendencies	Brown's adaptive model
2020	50880,24	60739,38	50880,2
2021	54256,77	69431,6	54256,72
2022	57633,3	80031,8	57633,3

Source: calculated by authors

Thus, if we compare the three methods and the models built on them, it should be noted that only two of them most accurately reflect the real processes: a linear trend model and an adaptive Brown model. However, it is important to remember that the Brown model takes into account changes in the middle of the time series. In most cases, this model provides the most adequate forecast values, because the smoothing factors take into account the crisis periods that occur in the economy.

Conclusions. It was found that one of the barriers to the effective development of the insurance market in general and insurance companies in particular are the insufficient level of planning of their activities, especially in terms of forecasting key indicators. In the course of the research the

optimality of planning is substantiated, it is shown that it can be achieved due to the accuracy of the forecast level of the planned indicators. In view of this, it was proposed to use modern forecasting methods depending on the purpose and forecasting period. The expediency of using adaptive methods is substantiated, because they are able to take into account the evolution of dynamic characteristics when reflecting the current state of the object under study under conditions of environmental variability. The effectiveness of Brown's adaptive model for short-term planning of insurance premium receipts is proved. The proposed model is tested for adequacy, based on its recommendations for further application in the practice of insurance companies.

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