The Insulin Bio Code - Standard Deviation

Lutvo Kurić

Bosnia and Herzegovina

ABSTRACT-This paper discusses cyberinformation studies of the amino acid composition of insulin, in particular the identification of scientific terminology that could describe this phenomenon, ie, the study of genetic information, as well as the relationship between the genetic language of proteins and theoretical aspect of this system and cybernetics. The result of this research show that there is a matrix code for insulin. It also shows that the coding system within the amino acidic language gives detailed information, not only on the amino acid "record", but also on its structure, configuration and its various shapes. The issue of the existence of an insulin code and coding of the individual structural elements of this protein are discussed. Answers to the following questions are sought. Does the matrix mechanism for biosynthesis of this protein function within the law of the general theory of information systems, and what is the significance of this for understanding the genetic language of insulin? What is the essence of existence and functioning of this language?

Is the genetic information characterized only by biochemical, or also by cyberinformation principles? The potential effects of physical and chemical, as well as cybernetic and information ptinciples, on the biochemical basis of insulin are also investigated. This aper discusses new methods for developing genetic technologies, in particular more advanced digital technology based on programming, cybernetics, and informational laws and systems, and how this new technology could be useful in medicine, bioinformatics, genetics, biochemistry, and other natural sciences.

Index Terms: - human insulin; bio code; standard deviation; genetics code; amino acids code;

INTRODUCTION

The biologic role of any given protein in essential life processes, eg, insulin, depends on the positioning of its component amino acids, and is understood by the "positioning of letters forming words". Each of these words has its biochemical base. If this base is expressed by corresponding discrete numbers, it can be seen that any given base has its own program, along with its own unique cybernetics and information characteristics.

Indeed, the sequencing of the molecule is determined not only by distin biochemical features, but also by cybernetic and information principles. For this reason, research in this field deals more with the quantitative rather than qualitative characteristics of genetic information and its biochemical basis. For the purposes of this paper, specific physical and chemical factors have been selected in order to express the genetic information for insulin.Numerical values are them assigned to these factors, enabling them to be measured. In this way it is possible to determine oif a connection really exists between the quantitative ratios in the process of transfer of genetic information and the qualitative appearance of the insulin molecule. To select these factors, preference is given to classical physical and chemical parameters, including the number of atoms in the relevant amino acids, their analog values, the position in these amino acids in the peptide chain, and their frequencies. There is a arge numbers of these parameters, and each of their gives important genetic information. Going through this process, it becomes clear that there is a mathematical relationship between quantitative ratios and the qualitative appearance of the biochemical "genetic processes" and that there is a measurement method that can be used to describe the biochemistry of insulin.

Methods

The sample of an insulin can be represented by two different forms: one is the discrete form and the other is the sequential form. In the discrete form, a insulin is represented by a set of discrete codes or a multiple dimension vector. In the sequential form, an insulin is represent by a series of amino acids according to the order of their position in the chains 1AI0. Therefore, the sequential form can naturally reflect all the information about the sequence order and lenght of an Insulin. The crux is: can we develop a different discrete form to represent an insulin that will allow accomodation of partial, if not all, sequence-order information? Since a protein sequence is usually represented by a series of amino acid codes, what kind of numerical values should be assigned to these codes in order to optimally convert the sequence-order information into a series of numbers for the discrete form representation?

How functioning of biochemistry is determined through cybernetic information principles, will be discussed further in next section. EXPRESSION OF INSULIN CODE MATRIX-1A10

The matrix mechanism of Insulin, the evolution of biomacromolecules and, especially, the biochemical evolution of Insulin language, have been analyzed by the application of cybernetic methods, information theory and system theory, respectively. The primary structure of a molecule of Insulin is the exact specification of its atomic composition and the chemical bonds connecting those atoms.

R6 INSULIN HEXAMER (d1ai02)

The structure 1AI0 has in total 12 chains. Out of these 2 are sequence-unique (Figure 1)

									1	AI0	:A									
G	Т	v	Е	Q	с	с	т	s	I.	с	s	L	Y	Q	L	Е	N	Y	с	N
10	22	19	19	20	14	14	17	14	22	14	14	22	24	20	22	19	17	24	14	17
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21
									1	AI0	:B									
			F	v	Ν	Q	Н	T	С	G	S	Н	L	۷	Е	Α	L			
			23	19	17	20	20	22	14	10	14	20	22	19	19	13	22			
			22	23	24	25	26	27	28	29	30	31	32	33	34	35	36			
				1																
			Y	L	V	С	G	Е	R	G	F	I	Y	т	Ρ	К	т			
			24	22	19	14	10	19	26	10	23	22	24	17	17	24	17			
			37	38	39	40	41	42	43	44	45	46	47	48	49	50	51			
	etc.																			
	Rank = 1,2,3, n.																			

Number of amino acids = 306; Number of atoms = 5640;

"Fig 1." Number of atoms in insulin A and B chain

Notes: And in that dimension we can find an explanation for the given empirical reality. Aforementioned aminoacids are positioned from number 1 to 51. Numbers 1, 2, 3, n... present the position of a certain aminoacid.

The aforementioned aminoacids are positioned from number 1 to 51. Numbers 1, 2, 3, n... present the position of a certain aminoacid. This positioning is of the key importance for understanding of programmatic, cybernetic and information principles in this protein. The scientific key for interpretation of biochemical processes is the same for insulin as other proteins and sequences in biochemistry. The first aminoacid in this example has 10 atoms, the second one 22, the third one 19, etc. They have exactly these numbers of atoms because there are many codes in the insulin molecule, analog codes, and other voded features. In fact, there is a cybernetic algorithm which it is "recorded" that the firs amino acid has to have 10 atoms, the second one 22, the third one 19, etc. The obvious conclusion is that there is a concrete relationship between

quantitative ratios in the process of transfer of genetic information and qualitative appearance, ie, the characteristics of the organism.

RESULTS

We shall now give some mathematical evidences that will prove that in the biochemistry of insulin there really is programmatic and cybernetic algorithm in which it is "recorded", in the language of mathematics, how the molecule will be built and what will be the quantitative characteristics of the given genetic information.

In genetics standard deviation is a measure of difference for interval and ratio variables between the observed number of atoms (DNA, RNA, protein) and the mean. The sign of deviation, either positive or negative, indicates whether the observation is larger than or smaller than the mean. The magnitude of the value reports how different (in the relevant numerical scale) an observation is from the mean. One of the features of the mean is that the sum of the deviations across the entire set of all observations is always zero.

The arithmetic mean of a list of numbers is the sum of all the members of the genetic sequences divided by the number of items in the list. If the list is a statistical sample, we call the resulting statistic a sample mean. Examples are presented in Eq. [1].

$$\overline{x} = \frac{1}{N} \sum_{i=1}^{N} x_i$$

Eq. [1]. The formula for calculating of the arithmetic mean of a list of numbers is the sum of all the members of the genetic

STANDARD DEVIATION 1

Insulin is composed of aminoacids with various numerical values. This numerical values are in an irregular order. Here is a concrete example:

G,A,C,S,T,N,P,V,E,Q,H,I,L, F,Y,K,R. = 17 aminoacids and 321 atoms;

X = Standard deviation X = (321 : 17) = 18,8823529

TABLE 1

Representation of the negative standard deviations 1.

Standard deviation (-)							
Amino	Number	Standard					
acids	of atoms	deviation					
2	3	4					
G	10	-8,882353					
A	13	-5,882353					
С	14	-4,882353					

	S	14	-4,882353	
	Т	17	-1,882353	
	Ν	17	-1,882353	
	Р	17	-1,882353	
	7 amino	102	(-)30,17647	
0 0 0	00E0. (10 1	0 000252)	= 0000000, (11 10000)	12521

(10-18,882353) = -8,882353; (13 - 18,882353) = -5,882353; (14 - 18,882353) = -4,882353; etc.

TABLE 2

Representation of the positive standard deviations 1.

Standard deviation (+)						
Amino acids	Number	Standard				
	of	deviation				
	atoms					
2	3	4				
V	19	0,117647				
E	19	0,117647				
Q	20	1,117647				
Н	20	1,117647				
I	22	3,117647				
L	22	3,117647				
F	23	4,117647				
Y	24	5,117647				
K	24	5,117647				
R	26	7,117647				
10 amino	219	+30,17647				

(19 - 18,882353) = 0,117647; (20 - 18,882353) = 1,117647; (22 - 18,882353) = 3,117647; etc.

(R - T) = (7,117647 - (-) 1,882353) = 9; (R - N) = (R-P) = 9; (F - S) = (F - C) = 9;

(L - T) = (L - N) = (L - P) = 5; (F - T) = (F - N) = (F - P) = 6;

etc.

A data set with a mean of 18,8823529 and a standard deviation of 30,17647.

A low standard deviation indicates that the data points tend to be very close to the mean, whereas high standard deviation indicates that the data is spread out over a large range of values.

Correlation of the negative and positive standard deviation 1

Standard deviation (-) = -30,17647Standard deviation (+) = +30,17647

Thus, there is a mathematical balance of negative and positive standard deviation.

Schematic representation of the negative and positive standard deviations 1 we will show in the fig.2.

Determinants of standard deviation

Mathematical evidence is provided here to prove that in the biochemistry of insulin there really is programmatic and cybernetic algorithm in which it is "recorded", in the language of mathematics, how the

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molecule will be built and what will be the quantitative characteristics of the given genetic information. Example: The standard deviation of the amino acids we can calculate with the help of determinants:

DET
$$||7, 102, 10, 219|| = 513;$$

[513 : (7+10)] = **30,17647**

Standard deviations: -30,17647, +30,17647, DET 30,17647

"Fig 2." Schematic representation of the determinants of standard deviation

As we see, the digital pictures of biochemistry provide us with cybernetic and information interpretation of the scientific facts. Now we have the exact scientific proofs that there is a genetic language that can be described by the theory of systems and cybernetics, and which functions in accordance with certain principles.

STANDARD DEVIATION 2

This protein has 306 amino acids. These amino acids are 5640 atoms.

X = Standard deviation X = (5640:306) = 18,431373

TABLE 3.

Representation of the negative standard deviations 2.

Standard deviation (-)

Amino Number of Standard Number of Aminoacids acids atoms deviation 2 3 4 1 G 10 -8,431373 24 -5,431373 А 13 6 С 14 -4,431372 36 S 14 -4,431372 18 Т 17 -1,431373 18 Ν 17 -1,431373 18 Р 17 -1,431373 6 Sum 102 126

(10-18,431373) = -8,431373; (13-18,431373) = -5,431373; etc.

TABLE 4.

Representation of the positive standard deviations 2.

Standard deviation (+)								
Amino	Number of	Standard	Number of					
acids	atoms	deviation	Aminoacids					
1	2	3	4					
V	19	+0,568627	24					
E	19	+0,568627	24					
Q	20	+1,568627	18					

Н	20	+1,568627	12
1	22	+3,568627	24
L	22	+3,568627	30
F	23	+4,568627	12
Y	24	+5,568627	24
K	24	+5,568627	6
R	26	+7,568627	6
Sum	219		180

DET || 219, 180, 102, 126|| = 9234;

[9234 : (126+180)] = **30,17647;**

 $9234 = (513 + 513 \dots + 513)$

A data set with a mean of 18,431373 and a standard deviation of 30,17647.

Correlation of standard deviation

Standard deviation 1

DET || 7, 102, 10, 219 || = 513;

[513 : (7+10)] = 30,17647

↓

Standard deviation 2

DET ||219, 180, 102, 126|| = 9234 = (513+513+513...+513);

[9234 : (126+180)] = 30,17647

Standard deviations: -30,17647, +30,17647, DET 30,17647

Comment:

$$9234 = (96^2 + 18);$$

9234 = (513 x <mark>18</mark>)

As we see, within the digital pictures in biochemistry, the physical and chemical parameters are in a strict compliance with programmatic, cybernetic and information principles. This can be seen from the following examples:

6

STANDARD DEVIATION 3

The standard deviation is the most common measure of statistical dispersion in genetics, measuring how widely spread the values in a data set are. If many data points are close to the mean, then the standard deviation is small; if many data points are far from the mean, then the standard deviation is large. If all the data values are equal, then the standard deviation is zero.

X = Standard deviation

X = (5640:306) = 18,431373

TABLE 5.

Representation of the negative standard deviations 3.

Amino	Number of	Standard	Number of	(2x4)	(3x4)		
acids	atoms	deviation	aminoacids				
1	2	3	4	5	6		
G	10	-8,431373	24	240	-202,352952		
А	13	-5,431373	6	78	-32,588238		
С	14	-4,431372	36	504	-159,529392		
S	14	-4,431372	18	252	-79,764696		
Т	17	-1,431373	18	306	-25,764714		
Ν	17	-1,431373	18	306	-25,764714		
Р	17	-1,431373	6	102	-8,588238		
Sum	102	-	126	1788	-534,3529		
(10-18,431373) = 8,431373; $(13-18,431373) = 5,431373;$ etc.							

Standard deviation (-)

TABLE 6.

Representation of the positive standard deviations 3.

Amino	Number of	Standard	Number of	(2x4)	(3x4)
acids	atoms	deviation	aminoacids		
1	2	3	4	5	6
V	19	+0,568627	24	456	+13,647048
E	19	+0,568627	24	456	+13,647048
Q	20	+1,568627	18	360	+28,235286
Н	20	+1,568627	12	240	+18,823524
I	22	+3,568627	24	528	+85,647048
L	22	+3,568627	30	660	+107,05881
F	23	+4,568627	12	276	+54,823524
Y	24	+5,568627	24	576	+133,647048
K	24	+5,568627	6	144	+33,411762
R	26	+7,568627	6	156	+45,411762
Sum	219		180	3852	534,3529

Standard deviation (+)

Table 5. and 6. Overview of negative and positive standard deviations for R6 insulin HEXAMER (d1ai02), showing some of the quantitative characteristic of the insulin molecule and the exact mathematical balance between its components. The values show some of the quantitative characteristics of the molecule of insulin.

Standard deviation (-) = -534,3529Standard deviation (+) =+534,3529

DET || 126, 180, 1788, 3852 || = 163 512;

$[163\ 512:(126+180)] = 534,3529$

Standard deviations: -534,3529, +534,3529,

DET 534,3529

Schematic representation of the negative and positive standard deviations 3 we will show in the fig.3.

STANDARD DEVIATION 4

Schematic representation of the negative and positive standard deviations 4 we will show in the fig.10. and 11.

> X = Standard deviation X = (5640:306) = 18,431373

> > TABLE 7.

Representation of the negative standard deviations 4.

	Standard of deviation (-)							
Amino	Number of	Number of	(3x4)	Standard				
acids	atoms	Aminoacids		deviation				
				(-)				
2	3	4	5	6				
Т	17	18	306	-25,764706				
N	17	18	306	-25,764706				
F	23	12	276	-55,764706				
S	14	18	252	-79,764706				
G	10	24	240	-91,764706				
Н	20	12	240	-91,764706				
R	26	6	156	-175,764705				
K	24	6	144	-187,764706				
Р	17	6	102	-229,764706				
А	13	6	78	-253,764706				
10	181	126	2100	-1217,647059				

Ctauland of dealers

TABLE 8.

	Standard deviation (+)								
Amino	Number of	Number of	(3x4)	Standard					
acids	atoms	Aminoacids		deviation					
				(+)					
2	3	4	5	7					
Q	20	18	360	+28,235294					
V	19	24	456	+124,235294					
Е	19	24	456	+124,235294					
С	14	36	504	+172,235294					
Ι	22	24	528	+196,235294					
Y	24	24	576	+244,235294					
L	22	30	660	+328,235294					
7	140	180	3540	+1217,647059					

Representation of the positive standard deviations 4.

DET || 10, 2100, 7, 3540 || = 20700;

 $[20\ 700:(10+7)] = 1217,647059;$

Standard deviations: -1217,647059, +1217,647059, DET 1217,647059

Schematic representation of the negative and positive standard deviations 4 we will show in the fig.4.

STANDARD DEVIATION 5

In genetics standard deviation is a measure of difference for interval and ratio variables between the observed number of atoms (DNA, RNA, protein) and their position in the chains 1AI0. Example: Position of amino acids in the chains 1AI0-(Rank) = 46971;

X = Standard deviation X = (46971 : 17) = 2763;

TABLE 9.

Representation of the negative standard deviations 5.

Standard deviation (-)							
Amino	Number	Rank	Standard deviation				
acids	of atoms						
2	3	4	5				
N	17	2673	-90				
S	14	2601	-162				
Q	20	2565	-198				
F	23	1932	-831				
Н	20	1872	-891				
Р	17	1059	-1704				
K	24	1065	-1698				
R	26	1023	-1740				
A	13	975	-1788				
9 amino	174	15765	-9102				

 $(10-18,882353) = -\overline{8,882353}; (13 - 18,882353) = -5,882353; (14 - 18,882353) = -4,882353; \text{etc.}$

TABLE 10.

Representation of the positive standard deviations 5.

Standard deviation (+)								
Amino acids	Number of atoms	Rank	Standard deviation					
2	3	4	5					
Т	17	2937	+174					
I	22	3570	+807					
E	19	3642	+879					
V	19	3648	+885					
G	10	3750	+987					
Y	24	3762	+999					
L	22	4635	+1872					
C	14	5262	+2499					
8 amino	147	31206	+9102					

Standard deviation = -9102

Standard deviation = +9102

In this example, there is a mathematical balance of the standard deviation 5.

DET || 9, 15765, 8, 31206 || = 154 734;

[154 734 : (9+8)] = **9102**;

Correlation determinants

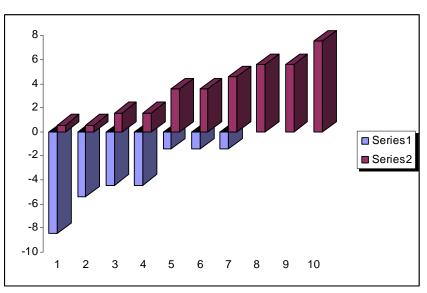
Determinants = 9234, 513, 154734, 163512; DET || 9234, 513, 154734, 163512 || = 1 430 491 266; 1430491266 = (513 + 513 + 513... + 513);

In this example there is also a mathematical balance between positive and negative values of the standard deviations. This balance is one of important quantitative characteristics of all processes in biochemistry. Each numerical value has its standard deviation. The research we carried out have shown that standard deviation are one of quantitative characteristics in biochemistry. Standard deviation is , actually, a discrete code that protects and guards genetic information coded in bio-chemical processes. This a recently discovered code, and more detailed knowledge on it is yet to be discovered.

From the previous examples we can see that this protein really has its quantitative characteristics. It can be concluded that there is a connection between quantitative characteristics in the process of transfer of genetic information and the qualitative appearance of given genetic processes.

DISCUSSION

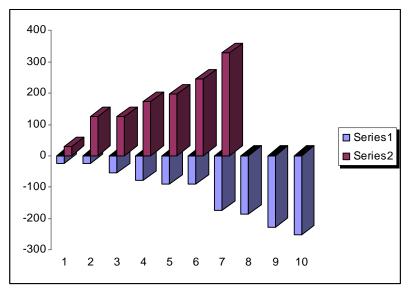
The results of our research show that the processes of sequencing the molecules are conditioned and arranged not only with chemical and biochemical lawfulness, but also with program, cybernetic and informational lawfulness too. At the first stage of our research we replaced nucleotides from the Amino Acid Code Matrix with numbers of the atoms and atomic numbers in those nucleotides. Translation of the biochemical language of these amino acids into a digital language may be very useful for developing new methods of predicting protein sub-cellular localization, membrane protein type, protein structure secondary prediction or any other protein attributes. Since the concept of Chou's pseudo amino acid composition was proposed [1-2], there have been many efforts to try to use various digital numbers to represent the 20 native amino acids in order to better reflect the sequence-order effects through the vehicle of pseudo amino acid composition. Some investigators used complexity measure factor [3], some used the values derived from the cellular automata [4-7], some used hydrophobic and/or hydrophilic values [8-16], some were through Fourier transform [17-18], and some used the physicochemical distance [19]. The author [34-42] is devoted to provide a digital code for each of 20 native amino acids. These digital codes should more complete and better reflect the essence of each of the 20 amino acids. Therefore, it might stimulate a series of future work by using the author's digital codes to formulate the pseudo amino acid composition for predicting protein structure class [20-22], subcellular location [23,24], membrane protein type [9,25], enzyme family class [26,27], GPCR type [28, 29], protease type [30], protein-protein interaction [31], metabolic pathways [32], protein quaternary structure [33], and other protein attributes. It is going to be possible to use a completely new strategy of research in genetics in the future. However, close observation of all these relationships, which are the outcomes of periodic laws (more specifically the law of binary coding), stereo-chemical and digital structure of proteins.



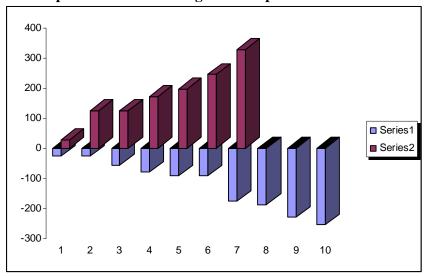
Schematic representation of the negative and positive standard deviations 1

"Fig 2." Schematic representation of the negative and positive standard deviations 1

Schematic representation of the negative and positive standard deviations 3



"Fig 3." Schematic representation of the negative and positive standard deviations 3.



Schematic representation of the negative and positive standard deviations 4

"Fig 4." Schematic representation of the negative and positive standard deviations 4.

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