

Molecular Biocoding of Insulin – Amino Acid Gly

Lutvo Kurić

Abstract - The modern science mainly treats the biochemical basis of sequencing in bio-macromolecules and processes in medicine and biochemistry. One can ask whether the language of biochemistry is the adequate scientific language to explain the phenomenon in that science. Is there maybe some other language, out of biochemistry, that determines how the biochemical processes will function and what the structure and organization of life systems will be? The research results provide some answers to these questions. They reveal to us that the process of sequencing in bio-macromolecules is conditioned and determined not only through biochemical, but also through cybernetic and information principles. Many studies have indicated that analysis of protein sequence codes and various sequence-based prediction approaches, such as predicting drug-target interaction networks (He et al., 2010), predicting functions of proteins (Hu et al., 2011; Kannan et al., 2008), analysis and prediction of the metabolic stability of proteins (Huang et al., 2010), predicting the network of substrate-enzyme-product triads (Chen et al., 2010), membrane protein type prediction (Cai and Chou, 2006; Cai et al., 2003; Cai et al., 2004), protein structural class prediction (Cai et al., 2006; Ding et al., 2007), protein secondary structure prediction (Chen et al., 2009; Ding et al., 2009b), enzyme family class prediction (Cai et al., 2005; Ding et al., 2009a; Wang et al., 2010), identifying cyclin proteins (Mohabatkar, 2010), protein subcellular location prediction (Chou and Shen, 2010a; Chou and Shen, 2010b; Kandaswamy et al., 2010; Liu et al., 2010), among many others as summarized in a recent review (Chou, 2011), can timely provide very useful information and insights for both basic research and drug design and hence are widely welcome by science community. The present study is attempted to develop a novel sequence-based method for studying insulin in hopes that it may become a useful tool in the relevant areas.

Index Terms-Amino Acid Gly, Human Insulin, Insulin Model, Insulin Code.

1 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 distinct 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 then assigned to these factors, enabling them to be measured. In this way it is possible to determine if 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 large number of these parameters, and each of them 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.

2 METHODS

Insulin can be represented by two different forms, ie, a discrete form and a sequential form. In the discrete form, a molecule of insulin is represented by a set of discrete codes or a multiple dimension vector. In the sequential form, an insulin molecule is represented by a series of amino acids according to the order of their position in the chains 1A10.

Therefore, the sequential form can naturally reflect all the information about the sequence order and length of an insulin molecule. The key issue is whether we can develop a different discrete method of representing an insulin molecule that will allow accommodation of partial,

if not all sequence order information? Because a protein sequence is usually represented by a series of amino acids 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?

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.

3 Expression of Insulin Code Matrix- 1AI0

3.1 Model

The structure 1AI0 has in total 12 chains: A,B,C,D,E,F,G,H,I,J,K,L.

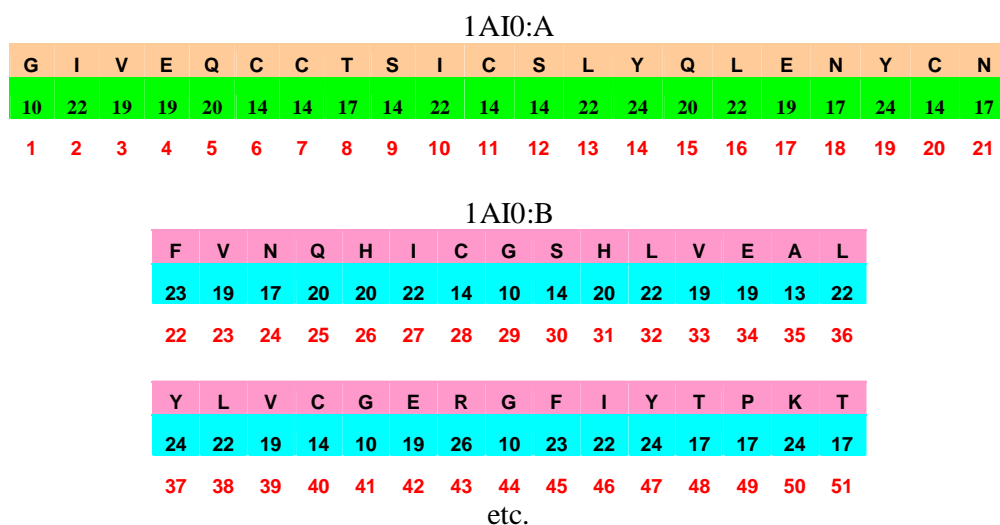


Fig. 1. Group of chains A,B,C,D,E,F,G,H,I,J,K,L.

Notes: Aforementioned aminoacids are positioned from number 1 to 306. 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 bio chemical processes is the same for insulin and as well as for the other proteins and other 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 first amino acid has its own biochemistry, as does the second and the third, 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.

3.2 Algorithm

We shall now give some mathematical evidences that will prove that in the biochemistry of hemoglobin in 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.

3.2.1 Atomic progression

Step 1 (Amino acids from 1 to 306)

$$AC_1 = 10 \text{ atoms}; AC_2 = 22 \text{ atoms}; AC_3 = 19 \text{ atoms}; \dots AC_{306} = 17 \text{ atoms};$$

$$[AC_1 + (AC_1 + AC_2) + (AC_1 + AC_2 + AC_3) \dots + (AC_1 + AC_2 + AC_3 \dots + AC_{147})] = S_1;$$

$$AC_1 = APa_1 = 10;$$

$$(AC_1 + AC_2) = (10 + 22) = APa_2 = 32;$$

$$(AC_1 + AC_2 + AC_3) = (10 + 22 + 19) = APa_3 = 51;$$

$$(AC_1 + AC_2 + AC_3 \dots + AC_{306}) = APa_{306} = 5640 \text{ atoms};$$

APa_{1,2,3,n} = Atomic progression of amino acids 1,2,3,n

$$[APa_1 + APa_2 + APa_3 \dots + APa_{306}] = (10 + 32 + 51 \dots + 5640) = S_1;$$

$$S_1 = 863\,208;$$

Example :

Atomic progression 1 (APa)

G	I	V	E	Q	C	.	.	P	K	T	Sum
10	22	19	19	20	14	.	.	17	24	17	5640
1	2	3	4	5	6	.	.	304	305	306	46971

↓

G	I	V	E	Q	C	.	.	P	K	T	Sum
10	32	51	70	90	104	.	.	5599	5623	5640	863 208
1	2	3	4	5	6	.	.	304	305	306	46971

$$(0+10) = 10; (10+22)=32; (10+11+19) = 51; \text{ etc.}$$

Fig. 2. Atomic progression 1 (APa) of amino acids from 1 to 306.

Notes: By using chemical-information procedures, we calculated the arithmetic progression for the information content of aforementioned aminoacids.

Step 2 (Amino acids from 306 to 1)

$$AC_{306} = 17 \text{ atoms}; AC_{305} = 24 \text{ atoms}; AC_{304} = 17 \text{ atoms}; \dots AC_1 = 10 \text{ atoms};$$

$$[AC_{306} + (AC_{306} + AC_{305}) + (AC_{306} + AC_{305} + AC_{304}) \dots + (AC_{306} + AC_{305} + AC_{304} \dots + AC_1)] = S_2;$$

$$AC_{306} = APb_{306} = 17;$$

$$(AC_{306} + AC_{305}) = (17 + 24) = APb_{306} = 41;$$

$$\begin{aligned}
 (AC_{306} + AC_{305} + AC_{304}) &= (17+24+17) = APb_{304} = 58; \\
 (AC_{306} + AC_{305} + AC_{304} + \dots + AC_1) &= APb_1 = 5640 \text{ atoms;} \\
 APb_{306,305,304, \dots, 1} &= \text{Atomic progression of amino acids } 306,305,304, \dots, 1; \\
 [APb_{306} + APb_{305} + APb_{1304} + \dots + APb_1] &= (17+41+58 + \dots + 5640) = 868\,272; \\
 S_2 &= 868\,272;
 \end{aligned}$$

Example:

Atomic progression 2 (APb)

G	I	V	.	.	I	Y	T	P	K	T	Sum
10	22	19	.	.	22	24	17	17	24	17	5640
1	2	3	.	.	301	302	303	304	305	306	46971
↓											
Sum	G	I	V	.	.	I	Y	T	P	K	T
868 272	5640	5630	5608	.	.	121	99	75	58	41	17
46971	1	2	3	.	.	301	302	303	304	305	306

(0+17) = 17; (17+24)=41; (17+24+17)=58; etc.

Fig. 3. Schematic representation of the atomic progression 2 from 306 to 1.

Within the digital pictures in biochemistry, the physical and chemical parameters are in a strict compliance with programmatic, cybernetic and information principles. Each bar in the protein chain attracts only the corresponding aminoacid, and only the relevant aminoacid can be positioned at certain place in the chain. Each peptide chain can have the exact number of aminoacids necessary to meet the strictly determined mathematical conditioning. It can have as many atoms as necessary to meet the mathematical balance of the biochemical phenomenon at certain mathematical level, etc. The digital language of biochemistry has a countless number of codes and analogue codes, as well as other information content. These pictures enable us to realize the very essence of functioning of biochemical processes. There are some examples:

Table 1. Atomic progression APa and APb (Amino acid Gly – position from 1 to 306 AA)

The structure 1AI0 – Amino acid Gly								
	G	G	G	G	G	G	G	G
Number of atoms	10	10	10	10	10	10	10	10
Rank	1	29	41	44	52	80	92	95
APa	10	523	741	796	950	1463	1681	1736
APb	5640	5127	4909	4854	4700	4187	3969	3914
AP(a,b)	5650	5650	5650	5650	5650	5650	5650	5650
	G	G	G	G	G	G	G	G
Number of atoms	10	10	10	10	10	10	10	10
Rank	103	131	143	146	154	182	194	197
APa	1890	2403	2621	2676	2830	3343	3561	3616
APb	3760	3247	3029	2974	2820	2307	2089	2034
AP(a,b)	5650	5650	5650	5650	5650	5650	5650	5650

	G	G	G	G	G	G	G	G
Number of atoms	10	10	10	10	10	10	10	10
Rank	205	233	245	248	256	284	296	299
APa	3770	4283	4501	4556	4710	5223	5441	5496
APb	1880	1367	1149	1094	940	427	209	154
AP(a,b)	5650	5650	5650	5650	5650	5650	5650	5650

Table 1. Schematic representation of the atomic progression APa and APb (Amino acid Gly – position from 1 to 306 AA).

Notes: Namely, having mathematically analyzed the atomic pregression model of Insulin Model (Table 1) we have found out that the protein code is based on a periodic law. This being the only to „read“ the picture, the solution of the main problem (concerning an arrangement where each amino acid takes only one, precisely determined position in the code), is quite manifest:

Atomic progression model of insulin should, in fact, be „remodelled“ into a periodic system. Examples:

Atomic progression APa and APb

	G	G	G	G	G	G	G	G	
Rank	299	1	29	296	41	284	44	256	> 1250
APa	5496	10	523	5441	741	5223	796	4710	> 22940
R	-144	-144	314	314	314	314	-144	-144	> 680
APb	5640	154	209	5127	427	4909	940	4854	> 22260
Rank	1	299	296	29	284	41	256	44	> 1250
$R > (5496-5640) = (-)144; (10-154) = (-)144; (523-209) = 314; \text{etc.}$									
Rank	52	248	80	245	92	233	95	205	> 1250
APa	950	4556	1463	4501	1681	4283	1736	3770	> 22940
R	-144	-144	314	314	314	314	-144	-144	> 680
APb	1094	4700	1149	4187	1367	3969	1880	3914	> 22260
Rank	248	52	245	80	233	92	205	95	> 1250
Rank	103	197	131	194	143	182	146	154	> 1250

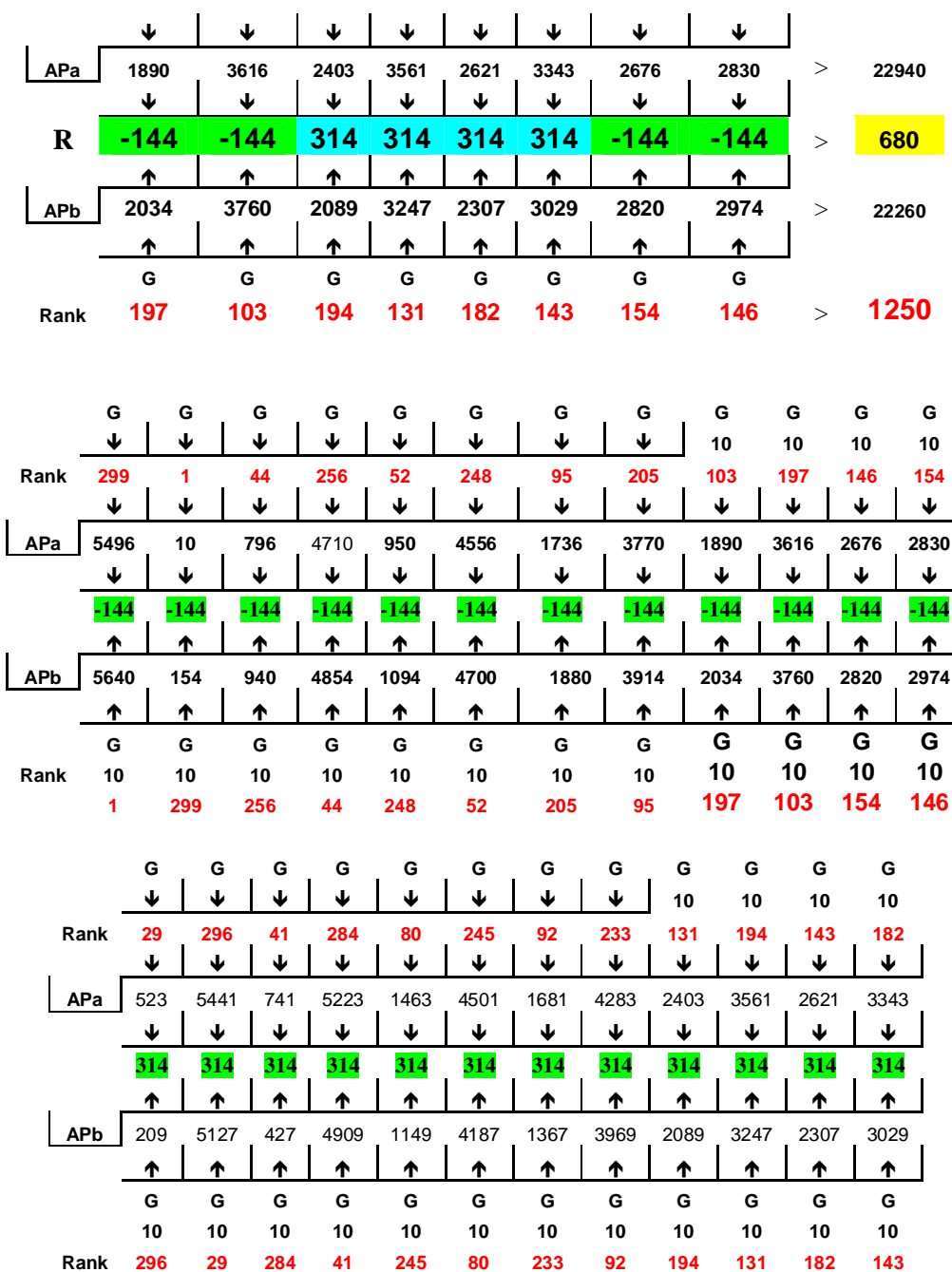


Fig. 4. Atomic progression APa and APb (Amino acid Gly – position from 1 to 306 AA).

In this example, the amino acids Gly atomic progression APa and APb as a result was given the codes 144 and 314th.

As we see, the insulin code is itself a unique structure of program, cybernetic and informational system and law.

The research we carried out have shown that atomic progression are one of quantitative characteristics in biochemistry. Atomic progression 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.

In a similar way we shall calculate bio codes of other unions of amino acids. Once we do this, we will find out that all these unions of amino acids are connected by various bio codes, analogue codes as well as other quantitative features. Examples:

Atomic progression APa and APb

	G	G	G	G	G	G	G	G
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙
APa	5496	10	523	5441	741	5223	796	4710
APa	950	4556	1463	4501	1681	4283	1736	3770
APa	1890	3616	2403	3561	2621	3343	2676	2830
APb	5640	154	209	5127	427	4909	940	4854
APb	1094	4700	1149	4187	1367	3969	1880	3914
APb	2034	3760	2089	3247	2307	3029	2820	2974
	⊙		⊙		⊙		⊙	⊙
		67800					67800	

	G	G	G	G	Sum
	⊙	⊙	⊙	⊙	
APa	5496	10	523	5441	> 11470
APa	950	4556	1463	4501	> 11470
APa	1890	3616	2403	3561	> 11470

	G	G	G	G	Sum
	⊙	⊙	⊙	⊙	
APa	741	5223	796	4710	> 11470
APa	1681	4283	1736	3770	> 11470
APa	2621	3343	2676	2830	> 11470

	G	G	G	G	Sum
	⊙	⊙	⊙	⊙	
APb	5640	154	209	5127	> 11130
APb	1094	4700	1149	4187	> 11130
APb	2034	3760	2089	3247	> 11130

	G	G	G	G	Sum
	⊙	⊙	⊙	⊙	
APb	427	4909	940	4854	> 11130
APb	1367	3969	1880	3914	> 11130
APb	2307	3029	2820	2974	> 11130

Fig. 5. Atomic progression APa and APb (Amino acid Gly – position from 1 to 306 AA).

Atomic progression presented in figure 2 are calculated using the relationship between corresponding groups of those regressions. These are groups with different progression. There are different ways and methods of selecting these groups of progressions, which method is most efficient some We hope that science will determine which method is most efficient for this selection.

3.2.2 Rank of atomic progression

	G	G	G	G	G	G	G	G	Sum
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
Rank	299	1	29	296	41	284	44	256	1250
Rank	1	299	296	29	284	41	256	44	1250
Rank	52	248	80	245	92	233	95	205	1250
Rank	248	52	245	80	233	92	205	95	1250
Rank	103	197	131	194	143	182	146	154	1250
Rank	197	103	194	131	182	143	154	146	1250
Sum	900	900	975	975	975	975	900	900	

	G	G	G	G		
	⊙	⊙	⊙	⊙		
Rank	299	1	29	296	>	625
Rank	1	299	296	29	>	625
Rank	52	248	80	245	>	625
Rank	248	52	245	80	>	625
Rank	103	197	131	194	>	625
Rank	197	103	194	131	>	625

	G	G	G	G		Sum
	⊙	⊙	⊙	⊙		
Rank	41	284	44	256	>	625
Rank	284	41	256	44	>	625
Rank	92	233	95	205	>	625
Rank	233	92	205	95	>	625
Rank	143	182	146	154	>	625
Rank	182	143	154	146	>	625

	G	G	G	G		Sum
	⊙	⊙	⊙	⊙		
Rank	299	1	44	256	>	600
Rank	1	299	256	44	>	600
Rank	52	248	95	205	>	600
Rank	248	52	205	95	>	600
Rank	103	197	146	154	>	600
Rank	197	103	154	146	>	600
Sum	900	900	900	900		

	G	G	G	G		Sum
	⊙	⊙	⊙	⊙		
Rank	29	296	41	284	>	650
Rank	296	29	284	41	>	650
Rank	80	245	92	233	>	650
Rank	245	80	233	92	>	650
Rank	131	194	143	182	>	650
Rank	194	131	182	143	>	650
Sum	975	975	975	975		

Fig. 6. Rank atomic progression APa and APb (Amino acid Gly – position from 1 to 306 AA).

In those examples, we have the mathematical balance in rows and columns in this figure.

In fact, we have discovered the mathematical balance in distribution of sequences in figure 6 is achieved.

3.2.3 Correlation of atomic progression

Atomic progression of this amino acid to us as a result of its correlation give a variety of codes. Here are some examples:

	G	G				G	G				G	G		
	10	10				10	10				10	10		
	41	284	>	325		80	245	>	325		143	182	>	325
APa	741	5223	>	5964		1463	4501	>	5964		2621	3343	>	5964
APb	4909	427	>	5336		4187	1149	>	5336		3029	2307	>	5336
	⊙	⊙		⊙		⊙	⊙		⊙		⊙	⊙		⊙
R	314	314		11300		314	314		11300		314	314		11300

	G	G				G	G				G	G		
	10	10				10	10				10	10		
	44	256	>	300		52	248	>	300		146	154	>	300
APa	796	4710	>	5506		950	4556	>	5506		2676	2830	>	5506
APb	4854	940	>	5794		4700	1094	>	5794		2974	2820	>	5794
	⊙	⊙		⊙		⊙	⊙		⊙		⊙	⊙		⊙
R	314	314		11300		314	314		11300		314	314		11300

Fig. 7. Correlation of atomic progression and rank of APa and APb (Amino acid Gly – position from 1 to 306 AA).

In those examples we have the correlation of atomic progression and rank of APa and APb.

3.2.4 Odd and even progression

Progression of the APa and APb, in fact, odd and even numbers. These numbers are one of the keys to decoding and decoding molecules insulina. This decoding we can make this:

Steam progression

	G	G	G	G	G	G	G	G	G	G	G	G
	10	10	10	10	10	10	10	10	10	10	10	10
	1	44	52	95	103	146	154	197	205	248	256	299
APa	10	796	950	1736	1890	2676	2830	3616	3770	4556	4710	5496
APb	5640	4854	4700	3914	3760	2974	2820	2034	1880	1094	940	154
	5650	5650	5650	5650	5650	5650	5650	5650	5650	5650	5650	5650

etc.

Steam progression are:

APa = (10, 5640); (1736, 3914); (1890, 3760); etc.

Progression are the odd and even rank.

Odd rank

	G	G	G	G	G	G
	10	10	10	10	10	10
Rank	1	95	103	197	205	299
						>
						900

APa	10	1736	1890	3616	3770	5496		16518
APb	5640	3914	3760	2034	1880	154	>	17382
	5650	5650	5650	5650	5650	5650		

Even rank

	G	G	G	G	G	G		
	10	10	10	10	10	10		
Rank	44	52	146	154	248	256	>	900
APa	796	950	2676	2830	4556	4710		16518
APb	4854	4700	2974	2820	1094	940	>	17382
	5650	5650	5650	5650	5650	5650		

Odd rank = Even rank = 900;
 Odd APa = Even APa = 16518;
 Odd APb = Even APb = 17382;

Fig. 8. Odd and even progression of aminoacid Gly.

Notes: Within the digital pictures in biochemistry, the physical and chemical parameters are in a strict compliance with programmatic, cybernetic and information principles. Each bar in the protein chain attracts only the corresponding aminoacid, and only the relevant aminoacid can be positioned at certain place in the chain. Each peptide chain can have the exact number of aminoacids necessary to meet the strictly determined mathematical conditioning. It can have as many atoms as necessary to meet the mathematical balance of the biochemical phenomenon at certain mathematical level, etc. The digital language of biochemistry has a countless number of codes and analogue codes, as well as other information content. These pictures enable us to realize the very essence of functioning of biochemical processes.

Odd rank progression

	G	G	G	G	G	G		
	⊙	⊙	⊙	⊙	⊙	⊙		
Rank	1	299	95	205	103	197	>	900
	⊙	⊙	⊙	⊙	⊙	⊙		
APa	10	5496	1736	3770	1890	3616	>	16518
	⊙	⊙	⊙	⊙	⊙	⊙		
	-144	-144	-144	-144	-144	-144		
	⊙	⊙	⊙	⊙	⊙	⊙		
APb	154	5640	1880	3914	2034	3760	>	17382
	⊙	⊙	⊙	⊙	⊙	⊙		
	G	G	G	G	G	G		
Rank	299	1	205	95	197	103	>	900
	↘	↘	↘	↙	↙			
								34836

$(1+10+(-)144+154+299...+ 103) = 34836;$

Even rank progression

	G	G	G	G	G	G		
	⊙	⊙	⊙	⊙	⊙	⊙		
Rank	44	256	52	248	146	154	>	900

	⊙	⊙	⊙	⊙	⊙	⊙		
APa	796	4710	950	4556	2676	2830	>	16518
	⊙	⊙	⊙	⊙	⊙	⊙		
	-144	-144	-144	-144	-144	-144		
	⊙	⊙	⊙	⊙	⊙	⊙		
APb	940	4854	1094	4700	2820	2974	>	17382
	⊙	⊙	⊙	⊙	⊙	⊙		
	G	G	G	G	G	G		
Rank	256	44	248	52	154	146	>	900
	↘	↘	↘		↙	↙		
								34836

$(44+796+ (-)144+940+256...+ 146) = 34836;$

Fig. 9. Odd and even rank progression of aminoacid Gly.

Odd AP-a progression

	G	G	G	G	G	G	G	G	G	G	G	G
	10	10	10	10	10	10	10	10	10	10	10	10
	29	41	80	92	131	143	182	194	233	245	284	296
APa	523	741	1463	1681	2403	2621	3343	3561	4283	4501	5223	5441
APb	5127	4909	4187	3969	3247	3029	2307	2089	1367	1149	427	209
	5650	5650	5650	5650	5650	5650	5650	5650	5650	5650	5650	5650

Odd APa progression – Even rank

	G	G	G	G	G	G
	⊙	⊙	⊙	⊙	⊙	⊙
Rank	80	296	92	284	182	194
	⊙	⊙	⊙	⊙	⊙	⊙
APa	1463	5441	1681	5223	3343	3561
	⊙	⊙	⊙	⊙	⊙	⊙
	1254	1254	1254	1254	1254	1254
	⊙	⊙	⊙	⊙	⊙	⊙
APb	209	4187	427	3969	2089	2307
	⊙	⊙	⊙	⊙	⊙	⊙
	G	G	G	G	G	G
Rank	296	80	284	92	194	182

Odd APa progression – Odd rank

	G	G	G	G	G	G
	⊙	⊙	⊙	⊙	⊙	⊙
Rank	29	245	41	233	131	143
	⊙	⊙	⊙	⊙	⊙	⊙

APa	523	4501	741	4283	2403	2621
	⊙	⊙	⊙	⊙	⊙	⊙
	-626	-626	-626	-626	-626	-626
	⊙	⊙	⊙	⊙	⊙	⊙
APb	1149	5127	1367	4909	3029	3247
	⊙	⊙	⊙	⊙	⊙	⊙
	G	G	G	G	G	G
Rank	245	29	233	41	143	131

Even rank and odd rank

	G	G	G	G	G	G
	⊙	⊙	⊙	⊙	⊙	⊙
Rank par	80	296	92	284	182	194
Rank nepar	29	245	41	233	131	143
	51	51	51	51	51	51
Rank par	296	80	284	92	194	182
Rank nepar	245	29	233	41	143	131
	51	51	51	51	51	51

$(80-29) = 51; (296-245) = 51; (92-41) = 51;$

Even APa and odd rank

	G	G	G	G	G	G
	⊙	⊙	⊙	⊙	⊙	⊙
APa rank even	1463	5441	1681	5223	3343	3561
APa rank odd	523	4501	741	4283	2403	2621
	940	940	940	940	940	940
APb odd	1149	5127	1367	4909	3029	3247
APb even	209	4187	427	3969	2089	2307
	940	940	940	940	940	940

$(80-29) = 51; (296-245) = 51; (92-41) = 51;$

Figure 10. Odd AP-a progression, odd APa progression–even rank, odd APa progression –odd rank, even rank and odd rank and even APa and odd rank

3.3 Bio frequency

Insulin is composed of aminoacids with various numerical values. This numerical values are in an irregular order. For example, the first one has 10 atoms, the second one 22. Their frequency is X. Second amino acid has 22 atoms, and the third one 19. Their frequency is Y; etc... Frequency is the measurement for establishment of intervals of numerical values of amino acids in proteins. This value can be positive, negative or a zero value. These frequencies are showing us one completely new dimension of protein sequencing. Through these frequencies we can establish which of aminoacids are of primary, and which are of secondary significance in biochemical processes of insulin. Here is a concrete example:

G	I	V	E	Q	C	C	T
10	22	19	19	20	14	14	17

1	2	3	4	5	6	7	8
10	12	-3	0	1	-6	0	-3

From 0 to 10 = 10; From 10 to 22 = 12; From 22 to 19 = (-) 3; From 19 to 19 = 0; etc

Schematic representation of the amino acid and frequency we will show in the fig.11.

G	G	G	G	G	G	G	G	G	G	G	G
10	10	10	10	10	10	10	10	10	10	10	10
1	29	41	44	52	80	92	95	103	131	143	146
12	4	9	13	12	4	9	13	12	4	9	13

> 114;

G	G	G	G	G	G	G	G	G	G	G	G
10	10	10	10	10	10	10	10	10	10	10	10
154	182	194	197	205	233	245	248	256	284	296	299
12	4	9	13	12	4	9	13	12	4	9	13

> 114;

⊙⊙

	G	G	G	G	G	G	G	G	G	
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
Rank	1	299	29	296	41	284	44	256		> 1250
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
Fr	12	13	4	9	9	4	13	12		> 76
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
	-1	1	-5	5	5	-5	1	-1		> 0
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
Fr	13	12	9	4	4	9	12	13		> 76
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
	G	G	G	G	G	G	G	G		
Rank	299	1	296	29	284	41	256	44		> 1250

	G	G	G	G	G	G	G	G		
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙		
Rank	52	248	80	245	92	233	95	205		> 1250
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
APa	12	13	4	9	9	4	13	12		> 76
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
	-1	1	-5	5	5	-5	1	-1		> 0
	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	
APb	13	12	9	4	4	9	12	13		> 76

In a similar way we can calculate the analogue expression for any numerical value. Our research has shown that analog codes are quantitative characteristics in biochemistry. Analogue biocode is a discrete code that protects and guards genetic information coded in biochemical processes.

This a recently discovered code, and more detailed knowledge about it is necessary.

Odd rank and analog frequency

		G	G	G	G	G	G	G	G	G	G	G			
		10	10	10	10	10	10	10	10	10	10	10			
		1	29	41	95	103	131	143	197	205	233	245	299		
Frequency	>	12	04	09	13	12	04	09	13	12	04	09	13	>	114;
		⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙		
Analogue															
frequency	>	21	40	90	31	21	40	90	31	21	40	90	31	>	546;

Even rank and analog frequency

		G	G	G	G	G	G	G	G	G	G	G			
		10	10	10	10	10	10	10	10	10	10	10			
		44	52	80	92	146	154	182	194	248	256	284	296		
Frequency	>	13	12	04	09	13	12	04	09	13	12	04	09	>	114;
		⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙	⊙		
Analogue															
Frequency	>	31	21	40	90	31	21	40	90	31	21	40	90	>	546;

Figure 13. Schematic representation of the amino acid Gly and their odd-even rank and analog frequency

Analogue code is , actually, a discrete code that protects and guards genetic information coded in biochemical processes.

In the previous examples we translated the physical and chemical parameters from the language of biochemistry into the digital language of programmatic, cybernetic and information principles. This we did by using the adequate mathematical algorithms. By using chemical-information procedures, we calculated the numerical value for the information content of molecules. What we got this way is the digital picture of the phenomenon of biochemistry. These digital pictures reveal to us a whole new dimension of this science. They reveal to us that the biochemical process is strictly conditioned and determined by programmatic, cybernetic and information principles.

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.

4 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.

The success of human genome project has generated deluge of sequence information. The explosion of biological data has challenged scientists to accelerate the speed for their analysis. Nowadays, protein sequences are generally stored in the computer database system in the form of long character strings. It would act like a snail's pace for human beings to read these sequences with the naked eyes (Xiao and Chou, 2007). Also, it is very hard to extract any key features by directly reading these long character strings. However, if they can be converted to some signal process, many important features can be automatically manifested and easily studied by means of the existing tools of information theory (Xiao and Chou, 2007). The novel approach as presented here may help improve this kind of situation.

5 CONCLUSIONS AND PERSPECTIVES

The process of sequencing in bio-macromolecules is conditioned and determined not only through biochemical, but also through cybernetic and information principles. 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.

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