# THE MULTIFACTOR MATHEMATICAL MODEL FOR CONSTRUCTING A MULTITUDE OF CONSISTENT EDUCATIONAL PATHS FOR TRAINING FULL STACK SPECIALISTS

### **Vyacheslav Kozlov**

Samara State Technical University, Russian Federation

## **Alexander Guryanov**

Samara State Technical University, Russian Federation

#### Elena Alontseva

Samara State Technical University, Russian Federation

Abstract. The modern labor market, especially in the field of information technology startups, requires the training of a sufficiently large number of specialists with competencies in creating the minimum viable product (MVP). The educational organization should be able to quickly form the trajectory of specialist training based on the challenges of the labor market, taking into account the dynamics of its change, while maintaining the integrity and consistency of the educational program. The mathematical model of the formation of an educational program is proposed taking into account a large number of variable parameters, which allows one to construct many possible training paths for specialists and select the optimal ones according to the criteria of cost, efficiency or laboriousness.

**Keywords:** educational program, full stack specialist, information technology, minimum viable product, multifactor mathematical model, startup.

#### Introduction

In the present work, the term "full-stack" of a specialist will mean a person who is able, alone, without involving other specialists, to solve the problem of a complex production task. For example, in the context of web development (hereinafter, examples of specific competencies will be given from the field of web development), this is a person who is able to develop a small website application completely on his own - from a technical task to a result worthy of an act of work (Gurcan & Köse, 2017).

The modern labor market, especially in the field of information technology startups, requires the training of a sufficiently large number of specialists with competencies in creating the minimum viable product (MVP) (Münch et al., 2013; Bosch, Olsson, Björk, & Ljungblad, 2013). The demand for such specialists has

been continuously growing since 2013, as evidenced by the growth of search queries (https://trends.google.com, 100 points - maximum popularity) in Fig. 1, while the need for highly specialized specialists is not only not growing, and has a downtrend.

This is due to the fact that full development is rapidly gaining momentum, and full-stack developers are becoming very popular in some companies (Guryanov, Kozlov, & Zhuravliova, 2019; Savoskina, Domnina, & Kozlov, 2017). A study by Stack Overflow 2017 showed that this type of developer is not only the most popular, but also the most popular. The demand, especially in small companies, for full-stack developers is related to the specifics of these companies, for example, the "start-up" movement (Dennehy et al., 2019), and the fact that having one person with several skills rather than several people with specific skills is of real value to many organizations. In addition, there is a big time savings, as a developer who can switch between levels and understand the whole process.

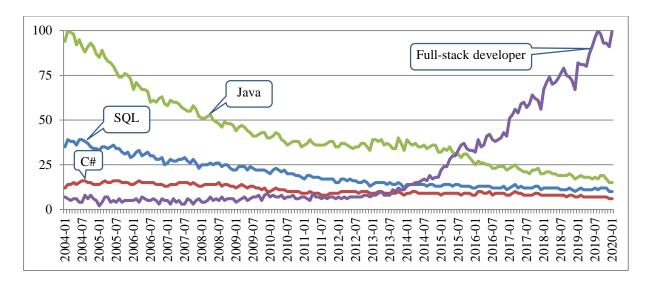


Figure 1 Search query dynamics

General developers work in all layers of software. They understand the principles and can work on both sides, although they do not always master all the subtleties as their highly specialized colleagues (Kozlov & Nasyrov, 2014).

#### **Model**

The development of a full-stack specialist training plan requires the creation of a specialized mathematical model that should take into account a lot of factors and form a coordinated stack of the technologies studied, covering all the necessary competencies for the chosen training area and should have developed

monitoring of the quality of training (Kozlov, Alontseva, & Guryanov, 2019; Yusupova, Alontseva, Kozlov, & Kulakova, 2017).

Let a set of knowledge, skills and competencies related to the direction of training. Obviously, some competencies require preparatory steps in mastering the previous competencies, for example, it is obvious that the "access to databases" competency requires a preliminary study of the SQL data access language and some programming language for which the database is bound. So for a professional in the field of web development, the set will be as follows ("System Administrator" Magazine, 2016 / Issue No. 01-02 (p. 158-159), "Vacancy: Full Stack-Web Developer"), for understanding, the full minimum list of competencies of the developer of WEB applications (Park & Wiedenbeck, 2011; Northwood, 2018):

- Understanding the principles of networks and cloud technologies;
- Ability to design and develop APIs and integrate;
- Basic experience in administering operating systems;
- Knowledge and ability to work with relational DBMS;
- Knowledge of the working principles of the HTTP protocol and the network stack as a whole;
- Ability to work with WEB servers (understanding of their differences, advantages, features, etc.);
- Knowledge of server programming language (one or more);
- Knowledge of server frameworks (one or more);
- Experience working with modern IDEs;
- Knowledge of the client layer at an average level;
- Ability to work with version control system;
- Ability to work with caching systems;
- Ability to work with NoSQL database;
- Skills of development of highly loaded services. Understanding scalability;
- Experience in writing Unit tests.
- Knowledge of algorithms;
- Understanding User Experience;
- Knowledge of flexible development methods and ability to work in a team;
- Ability to work with customers and to formulate technical requirements from business requirements;
- The ability to design and develop data layers and business logic, arguing for architectural decisions.

Let  $B_j$  - a set of technologies, and the value  $C_{ij}$  determines the relationship i-th of that competency from the set  $A_i$  and j-th of that technology from the set.

$$C_{ij} = \begin{cases} 1, & \text{if technology j provides competency i} \\ 0, & \text{if technology j is not related to competency i} \end{cases}$$
 (1)

Due to the large size of the matrix  $C_{ij}$ , we present in table (Table 1) only a fragment of it (for clarity, zero elements are not indicated).

The components of the vector  $B_i$  are binary variables, defined as:

$$B_{j} = \begin{cases} 1, & \text{if technology j is being studied} \\ 0, & \text{if technology j is not studied} \end{cases}$$
 (2)

Obviously, each competency must be supported by at least one technology, therefore, the inequality that determines the formation of all competencies will be of the form:

$$\sum_{i} C_{ij} B_{j} \geq P_{i} for \, \forall \, i \tag{3}$$

Table 1 Technology and competency interconnection Cij

	Technologies $B_j$																
Competencies A <sub>i</sub>	Visual Studio	Zend Studio	Net Beans	Eclipse	ODBC	Ling	Hibernate	HTML	CSS	JavaScript	PHP	C#	Java	DOT NET	Spring	YII	Laravel
Knowledge of server														1	1	1	1
frameworks																	
Knowledge of server											1	1	1				
programming language																	
Mid-level knowledge of the								1	1	1							
client layer																	
Knowledge and ability to					1	1	1										
work with relational DBMS																	
Experience with modern IDE	1	1	1	1													

where  $P_i$  sets the minimum number of technologies that ensure the formation of the *i-th* competency. Note that an additional matrix  $V_{ij}$  can be introduced whose elements will determine the value (weight) of the technology for competency *i*. Then the inequality that determines the formation of all competencies can be written as:

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$$\sum_{i} V_{ij} B_j \ge P_i for \, \forall \, i \tag{4}$$

In this case  $P_i$  sets the minimum level of formation of the *i-th* competence.

Obviously, the technologies are interdependent, that is, individual technologies cannot be studied without a preliminary study of the supporting technologies. For example, learning the YII or Laravel framework involves learning the PHP language, but the reverse is not true. Thus, the interconnection of technologies can be represented as a directed graph or as a square asymmetric adjacency matrix. Let  $D_{jk}$  (the index k varies within the same limits as the index k) characterizes the dependence of k that technology on k and is given by the rule:

$$D_{jk} = \begin{cases} 1, & \text{if technology } j \text{ requires knowledge of technology } k \\ 0, & \text{if technology } j \text{ is independent of technology } k \end{cases}$$
 (5)

Table 2 shows an example of a matrix  $D_{jk}$  as applied to the technologies described in table 1.

	Technology B <sub>j</sub>																	
Competencies $A_i$	Visual	Zend	Studio	Net	Eclipse	ODBC	Ling	Hibernat	HTML	CSS	JavaScri	PHP	<b>#</b> 3	Java	DOT	Spring	ΙΙΑ	Laravel
Visual Studio																		
Zend Studio																		
Net Beans																		
Eclipse																		
ODBC																		
Linq															1			
Hibernate														1				
HTML																		
CSS									1									
JavaScript									1	1								
PHP																		
C#	1																	
Java																		
DOT NET													1					
Spring											1			1				
YII												1						
Laravel					•							1						

Table 2 Technology adjacency Matrix Dik

The compatibility condition of the studied technologies can be written as a system of inequalities:

$$D_{jk}B_j \ge B_j for \, \forall j,k \tag{6}$$

However, not so unambiguous connections of technologies are possible, but the requirements of studying one or more technologies from a certain family, for example, Java technology requires knowledge of one of the modern IDEs - Net Beans or Eclipse. We introduce the concept of a family of technologies and number them. Each technology can be assigned to one or more families, through matrix  $M_{nj}$  (Table 3).

$$M_{nj} = \begin{cases} 1, & \text{if technology j belongs to the family n} \\ 0, & \text{if technology j is not part of the family n} \end{cases}$$
 (7)

Technology  $B_i$ Visual Studio Zend Studio Technology family Net Beans JavaScript DOT NET Hibernate number Eclipse Laravel ODBC HTML Spring Ling PHP 1 (IDE for Java) 2 (IDE for PHP) 1 3 (Programming 1 1 1 languages)

Table 3 Technology joining families M<sub>ni</sub>

Then we introduce  $R_{jn}$  as a matrix whose elements determine the dependence of the *j-th* technology on the *n-th* family of technologies (Table 4).

$$R_{nj} = \begin{cases} 1, & \text{if technology j requires at least one technology of the family n} \\ 0, & \text{if technology j does not depend on technology n} \end{cases}$$
(8)

		Technology $B_j$															
Technology family number	Visual Studio	Zend Studio	Net Beans	Eclipse	ODBC	Ling	Hibernate	HTML	CSS	JavaScript	PHP	C#	Java	DOT NET	Spring	YII	Laravel
1 (IDE for Java)													1				
2 (IDE for PHP)											1						

3 (Programming languages)

Table 4 Technology dependence on technology families  $R_{nj}$ 

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Then the value  $\max_{j}(M_{nj}B_{j})$  will be equal to 1 if at least one of the technologies of the family n and 0 is studied otherwise. To ensure resolution of the dependence of the technologies under study on technology families, the following system of inequalities must be implemented:

$$B_{j} \max_{j} (M_{nj}B_{j}) \ge B_{j} \text{ for } \forall j, n \text{ or } B_{j}R_{nj} \sum_{j} M_{nj} B_{j} \ge B_{j} \text{ for } \forall j, n$$
 (9)

For clarity, you can display information on technology relationships from tables 2-4 in graphical form Fig. 2.

The solid lines in the Fig. 2 show the relationships from table 2 (unambiguous dependence), and the dashed lines show the data from tables 3 and 4 (requirements of at least one technology from the family). The arrows indicate the dependent technology to the supporting one. You may notice that in Fig. 2 some technologies or technology blocks are not connected to each other, which corresponds to independent technologies or technology blocks.

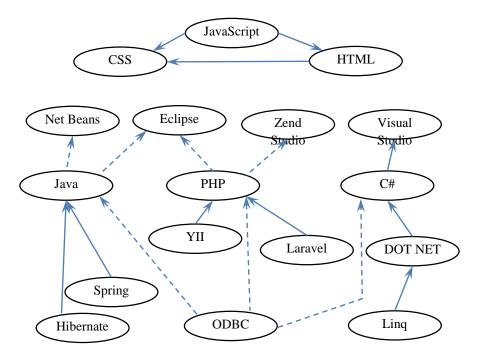


Figure 2 Technology interconnection graph

So, it can be argued that any set of technologies studied, defined by a vector  $B_j$  and satisfying conditions (6 and 9), is joint, and if conditions (3) are satisfied, it is complete, that is, it provides the formation of all the required competencies for a given training profile and represents the set of possible training paths for specialists.

We turn to the problem of choosing the optimal training path. To do this, we introduce a restriction on the maximum complexity of the preparation. Let the vector  $H_j$  determine the complexity in hours of studying j of that technology. When the task of choosing the optimal trajectory from the position of minimizing training costs is to solve the optimization problem (10) subject to conditions (3, 6 and 9):

$$\sum_{j} H_{j}B_{j} \to min \tag{10}$$

If it is necessary to choose the most useful trajectory for training specialists, equation 10 takes the form of a restriction (where  $H^{max}$  is the maximum available training time):

$$\sum_{j} H_{j} B_{j} \leq H^{max} \tag{11}$$

By the criterion of the usefulness of the educational trajectory, we understand the degree of demand for specialists trained on this trajectory by the labor market. To do this, using recruiting data (job site in the field of information systems and technologies https://www.dice.com, January 7, 2020), we will build table 5, taking into account the data from table 1. Table 5 in the last column gives the normalized frequency of references. Rationing of the frequency of technology references was carried out within the framework of competencies to which these technologies relate. This method allowed us to come to comparable values and get rid of the economies of scale of the demand for competencies, which is absolutely justified, since condition (3) guarantees training in all competencies of the specialty. Thus, the last column of table 5 is a vector of technology  $E_i$  demand.

Table 5 The usefulness of technology in the labor market

Technology	Number mention	Competence	Number of mention technologies within the competence	Normalized frequency of references,% (The vector of the utility of technology in the labor market E <sub>j</sub> )		
Visual Studio	269			67		
Zend Studio	20	experience with	404	5		
Net Beans	4	modern IDEs	404	1		
Eclipse	111			27		
ODBC	37	knowledge and ability		5		
Linq	71	to work with relational	685	10		
Hibernate	577	DBMS		84		

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HTML	2972	111		24	
CSS	1513	knowledge of the client layer on average level	12		
JavaScript	7909	layer on average level		64	
PHP	832	1		6	
C#	3017	knowledge of server programming language 14389 21			
Java	10540	programming ranguage		73	
DOT NET	2288			51	
Spring	2112	knowledge of server	4494	47	
YII	25	frameworks	4494	1	
Laravel	69			2	

Then the choice of the most useful training path is reduced to the search for maxim (12), subject to conditions (3, 6, 9 and 12):

$$\sum_{j} E_{j} B_{j} \to max \tag{12}$$

To set necessarily studied technologies, for example, for a specific employer, condition (13) can be introduced.

$$B_i \ge B_i' \text{ for } \forall j$$
 (13)

 $B'_i$  sets the required technologies to be studied by unit values.

#### **Results and Conclusions**

A multifactor mathematical model is developed for constructing a multitude of consistent educational paths for training full stack specialists, the formal formulation of which is presented below

#### Initial Data

 $A_i$  – a set of required competencies,

 $B_i$  – a set of available technologies,

 $C_{ij}$  – connection i of that competence and j of that technology,

 $V_{ij}$  – the significance (weight) of technology j for competence i,

 $D_{jk}$  – the dependence of technology j from k,

 $M_{nj}$  – the relationship of technology j and technology family n,

 $R_{jn}$  – he dependence of j hat technology on the nth family of technologies,

 $E_i$  – vector demand for technology by the labor market.

# **Optimization Parametrs**

 $P_i$  – minimum level of formation of *i-th* competency,

 $H_j$  – the complexity of j technology,

H<sup>max</sup> – the most labor intensive educational trajectory,

 $B_i'$  – compulsory technology.

Limitations

 $\sum_{i} V_{ij} B_{i} \geq P_{i} for \forall i$  – competency formation,

 $D_{jk}B_j \ge B_j$  for  $\forall j, k$  – compatibility (consistency) of technologies,

 $B_j R_{nj} \sum_j M_{nj} B_j \ge B_j$  for  $\forall j, n$  – compatibility (consistency) of technologies and technology families,

 $\sum_{j} H_{j}B_{j} \leq H^{max}$  – restriction on the complexity of training,

 $B_i \geq B_i'$  for  $\forall j$  – guaranteed study of required technologies.

Optimization Variables

 $B_i$  – binary vector trajectory preparation.

Objective Function

 $\sum_j E_j B_j \ \to \ max-efficiency$  from the standpoint of demand.

 $\sum_{j} H_{j} B_{j} \rightarrow min$  – minimization the complexity of training.

Solved tasks:

- The construction of many joint and complete trajectories of specialist training;
- Search for the optimal trajectory of the training of specialists;
- Searching for the most useful trajectory of specialist training with a limited labor intensity.

A mathematically sound technology has been obtained that allows us to develop training plans for specialists that cover all the necessary competencies for the chosen field of training on the basis of a specialized multifactor mathematical model for constructing a multitude of consistent educational ones. This allows us to satisfy the need for such specialists, which has been continuously growing since 2013, especially in small high-tech companies and startup projects.

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