IMPLEMENTATION OF EPS_T2DM Implementation of Early Prediction System for Type 2 Diabetes Mellitus

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Abstract: This paper describes the implementation of an early prediction system for Type 2 diabetes mellitus. Type 2 diabetes mellitus is a multifactorial disease. It is not only associated with an unhealthy lifestyle but also has a strong genetic component. Accordingly, in order to decrease an incidence rate of T2DM, it is important to predict T2DM risk with using multifactors which are supposed to affect T2DM. We have implemented a prediction system for T2DM, and it employs several statistical prediction models. These models are produced by statistical analysis about cohort data of Korean Genome and Epidemiology Study (KoGES), and include risk factors which are adequate for preventing T2DM in Korean populations. The prediction system is written in JSF and Java, and developed into web application which is designed through object oriented modeling. Web application of this system offers user interfaces in order to input data which is needed for predicting risk group, select predefined prediction models, and so on. The system provides the results which are predicted by selected models using inputted information.

1 INTRODUCTION

Type 2 diabetes mellitus (T2DM) is a multifactorial disease in which environmental triggers interact with genetic variants in the predisposition to the disease. In the last thirty years, due to rapid industrialization and the lifestyle change, the number of Korean patients with T2DM is on an increasing trend. In addition, more serious problem is that some significant number of T2DM patients do not aware of their disease before onset it. Therefore, it is not only important to treat T2DM, but also necessary to predict its risk and prevent it in advance.

To prevent and control a certain disease for specified population effectively, it is necessary to understand epidemiological features of patients. So, we have analyzed the data collected by cohort study from Korea National Institute of Health (KNIH). The primary prevention method of a disease is prediction of its risk group or patients unaware of their disease, and decreasing the incidence of the disease or complications by reduction of risk factors for that group. For that reason, we have implemented the early prediction system for T2DM which is named EPS T2DM.

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2 MATERIALS AND METHODS

At first, we have analyzed KoGES data to find risk factors for T2DM, and built prediction models by statistical analysis. Then, for constructing database schema, input values including risk factors are defined. Finally, the entire system is implemented as input modules, a select model module, model processing modules and displaying module for results.

2.1 Building Statistical Prediction Models

This study has been supported by community based cohort study Ansung and Ansan, Korean Genome and Epidemiology Study (KoGES) from KNIH. The data for this study consists of epidemiological information and genetic information (SNP values) for 212 T2DM public and 472 general (non-T2DM) public (Table 1).

Bi Kim D., Lee J., Jang E., Lee J. and Kim T. (2008).

Characteristic	Diabetes	Non Diabetes
Numbers	212	472
Sex(F/M)	133/79	264 / 208
Age(Years)	64.50±2.82	64.03±2.86
BMI(kg/m ²)	25.07±3.44	23.31±3.11
Triglyceride (mg/dl)	196.60±131.69	149.52±70.97
Fasting glucose(mg/dl)	111.20 ± 28.96	74.54±3.55
Total cholesterol(mg/dl)	202.78±44.79	180.66±31.64
HDL cholesterol(mg/dl)	44.56±10.59	44.34±9.91

Table 1: Basic characteristics. (Means±SD).

The Ansung and Ansan cohort data have been processed using the Statistical Analysis System for Windows (Ver. 9.1, SAS institute Inc., Cary, NC, U.S.A), and mined by several algorithms such as QUEST, C4.5, logistic regression, SVM, and KNN algorithm. Among data mining results, the results of QUEST, a binary-split decision tree algorithm for classification and data mining, is applied to EPS_T2DM. From this result, risk factors for T2DM are defined and prediction models are produced.

2.2 Design of Database for Input Values

According to data mining results, risk factors for T2DM are as follows:

- Clinical information Height and weight for BMI, Waist circumference for abdominal fatness, Blood pressure, Total cholesterol, High-density cholesterol, Triglyceride, Fasting glucose, and etc;
- History of diseases information Diabetes Mellitus, Cerebrovascular disease and other vascular diseases, Hypertension;
- Family history of diseases information Diabetes Mellitus;
- Genetic Information Selected single nucleotide polymorphisms (SNPs) in 15 genes in Insulin pathway, 8 genes in fatty acid binding/translocation, and 13 genes in GLUT4 translocation and 51 more genes related to T2DM.

Each category is represented in a table, and each factor is corresponds to each field (Figure 1).

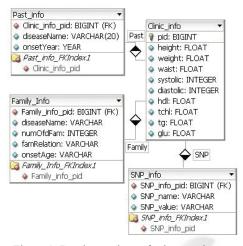


Figure 1: Database schema for input values.

2.3 Implementing EPS_T2DM

EPS_T2DM is based on Object Oriented Modeling. It is developed on Fedora core 6, written in HTML, JSF, JavaScript, Java, and etc. MySQL is used as a database to store and access data. The entire system is implemented with Spring Framework which is a layered Java/J2EE application framework, and then Model-View-Controller (MVC) design pattern is applied. MVC is an architectural pattern which encapsulates some data together with its processing (the model) and isolates it from the manipulation (the controller) and presentation (the view) part. Figure 2 illustrates the system architecture which is composed of five layers with UI layer, 3 layers above, and persistence layer.

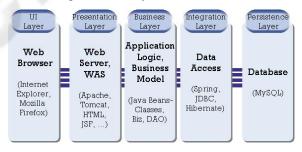


Figure 2: System architecture of EPS_T2DM.

The function of EPS_T2DM consists of 4 parts: (1) user interfaces to get input data and user's selection, (2) management modules of inputted data, (3) processing and management modules of statistical prediction models, and (4) interfaces to offer prediction results. JSF is used to implement user interfaces, and XML is employed to manage input data, SNP information, and prediction models.

3 RESULTS

EPS_T2DM is implemented as a web application. It offers interfaces in order to input SNP values and epidemiology data which include clinical information, history of diseases information and family history of diseases information. After getting input data, the system shows prediction models to users, and users can select them. The system applies selected models to user's input values, and displays prediction results that are represented whether or not a person corresponding to the inputted data belongs to risk group of T2DM. Figure 3 gives a flow chart of this system.

The web interfaces consist of epidemiology information input page with tabbed pane (Figure 4), SNP values input page, selecting models page, displaying results page, and so on.

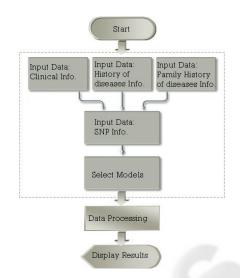


Figure 3: Web application's flow chart of EPS_T2DM.

General Info.	Hist	ory of Diseases Info.	F	Family History of Diseases Info.			Clinical Info.		
Residential distri	ct.*	Si/Do: Select! 💌 Si/Gu	ı/Gun:	: Select!	≚ Eup/My	eon/Dong: Selec			
Sex*		🔿 Male 💿 Female						11	
Birth*		MM/DD/YYYY							
The Number of Fa	amily*	4 v persons							
Marriage Status*		Select! 👻 Etc.:							
Religion*		Select! Y Etc.:	ilect	Genera	l Info.	History of Dise	ases Info.	Famil	y History of Diseases Int
Job*		Select!				,			
Academic Career Family's Monthly		Select!		Diseas	e Name	The Number of Diseased	Family Relati Incidence 4		
General Info.	History	y of Diseases Info.	Fa	Hyperto	ension*	Family 2persons 💌	Select! 👻	years	
)isease Name	Existar Dise	ase Diagnosis Year		Diabete Mellitus		Select!			
lypertension*	O No (O Yes		Hyperlipidaemia* Cerebrovascular Disease* Other Vascular Diseases*		Select! 💌			
iabetes Iellitus*	O No (Select! 💌			
lyperlipidaemia* erebrovascular						Select! 💌			
visease* Other Vascular	O No		ry of I	Diseases Info. Family Hist			ory of Diseases Info. Clinical Info.		
)iseases*	O NO								
		Health Condition*	Se	lect!	~				
		Height*	0.0	cm					
		Weight*	0.0	kg					
		Waist girth*	0.0	c	m				
		Blood Pressure *		1	mml	Hg			
		High-density Cholesterol*	0.0	n	ng/dl				
		Total Cholesterol*	0.0	n	ng/dl				
		Triglyceride*	0.0	n	ng/dl				
		Fasting Blood Sugar *	0.0	mg/dl					

Figure 4: The snapshots of EPS_T2DM's web application. The interface for inputting epidemiology data which is defined as risk factors of T2DM and other additive information such as sex, birth, etc.

4 DISCUSSION

In many cohort studies, various T2DM predicting models have been developed to guide intervention and inform health policy. Most of these studies tested models that only used personal information and clinical variables, not including genetic information. However, since development of T2DM is influenced by a complex interaction between genetic and environmental factors, genetic information is also needed to develop the prediction models for T2DM.

In this study, we implemented the system to predict the T2DM risk group with applying T2DM risk factors and statistical prediction models that include clinical and genetic information. The risk factors and prediction models are base on Ansung and Ansan, KoGES data and decision tree learning method, and these can be updated or changed by based data or analysis methods. EPS_T2DM is developed as object-oriented program, so it is easy to extended and enhance the system.

Our next step will be to expand and improve the system. This includes followings:

- The betterment of statistical models. This means improving accuracy or building new prediction models by analyzing other data or applying other data mining methods
- The enhancement of web application for uploading or processing massive data sets.

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