System Dynamics and Artificial Neural Network Integration: A Tool to Evaluate the Level of Job Satisfaction in Services

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Job Satisfaction,
System dynamics,
Artificial Neural Network (ANN),
Healthcare field,

ABSTRACT
This study presents an integrated intelligent algorithm to evaluate the level of job satisfaction (JS) in services. Job Satisfaction plays an important role as a competitive advantage in organizations especially in health industry. Recruitment and retention of human resources are persistent problems associated with this field. Most of the researches have focused on job satisfaction factors and few of researches have noticed about its effects on productivity. However, little researches have focused on the factors and effects of job satisfaction simultaneously by system dynamics approaches. In this paper, firstly, analyses the literature relating to system dynamics and job satisfaction in services specially at a hospital clinic and reports the related factors of employee job satisfaction and its effects on productivity. The conflicts and similarities of the researches are discussed and argued. Then a novel procedure for job satisfaction evaluation using (Artificial Neural Networks) ANNs and system dynamics is presented. The proposed procedure is implemented for a large hospital in Iran. The most influential factors on job satisfaction are chosen by using ANN and three different dynamics scenarios are built based on ANN’s result. The modeling effort has focused on evaluating the job satisfaction level in terms of key factors which obtain from ANN result such as Pay, Work and Co-Workers at all three scenarios. The study concludes with the analysis of the obtained results. The results show that this model is significantly useful for job satisfaction evaluation and provide policy makers with an appropriate tool to make more accurate decision on kind of reward. This is because the proposed approach is capable of handling non-linearity, complexity as well as uncertainty that may exist in actual data and situation.

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1. Introduction
Job satisfaction (JS) is the most widely-studied topic in organizational psychology. It has a long history in industry [1], where it is determined as a potential cause, correlate, and consequence of both work-related and non-work variables [2]. The traditional model of job satisfaction does focus on all the feelings that an individual has about his/her job. However, what makes a job satisfying or dissatisfying does not depend only upon the nature of the job but also on the expectations that individuals have of what their job
should provide. The work of Maslow(1954)[3] is seminal and do suggest that human needs form a five-
level hierarchy ranging from physiological needs, safety, belongingness, love, and esteem to self-
actualization. Based on Maslow’s theory, job satisfaction has been approached by a number of researchers from the perspective of the need fulfillment [4].

Job satisfaction at a hospital has been of interest to many researchers. Three main approaches have been used in these studies. One approach involves determining the level of satisfaction with very specific facets of the job as such as working conditions, salary, and supervision.

A second approach examines the differences in overall job satisfaction by personal demographic or workplace characteristics (i.e., age, gender, practice setting, and position). Third, researchers examine possible antecedents and consequences of job satisfaction such as role stress, skill utilization, commitment, and intention to leave [5]. Although, several different measures have been developed to assess job satisfaction, there are a few studies noticing system dynamics as a tool in their field. Many factors do influence on job satisfaction but their interactions are of very important to us. Therefore, using the system dynamics model for these topics would be very useful. On the other hand because of large number of factors for investigating job satisfaction field, the use of system dynamics can help to decrease the complexity of this study.

Using the other techniques to reduce the number of factors can help the researchers to design the accurate causal loop. So in this paper we use the ANN model in order to determine the most important factors. Then by the ANN result, the system dynamic model can be design correctly.

The rest of the paper is organized as follows. Section 1.1 briefly presents the definition of job satisfaction, and section 1.2 is devoted to introduce Artificial Neural Networks and in section 1.3 present the System dynamics. In Section 2, the related factors of job satisfaction are reviewed and a background discussion is presented. The research methodology is illustrated in Section 3. In Section 4, The NN modeling of job satisfaction is presented. According to literature review and ANN result, three scenarios of System dynamics related to job satisfaction field are presented in section 5. Section 6 presents the concluding remarks of this study. It contains the unique features of the study.

1-1. The Definition of Job Satisfaction
One of the most used definitions of job satisfaction is an emotional state resulting from appraising one’s job [6]. There are many definitions given by researchers but four of the most important one of them are listed in Tab. 1.

<table>
<thead>
<tr>
<th>Definition</th>
<th>researcher</th>
<th>year</th>
</tr>
</thead>
<tbody>
<tr>
<td>A function of satisfaction with the various elements of the job</td>
<td>Herzberg and Mausner</td>
<td>1959</td>
</tr>
<tr>
<td>The individual matching of personal needs to the perceived potential of the occupation for satisfying those needs</td>
<td>Kuhlen</td>
<td>1963</td>
</tr>
<tr>
<td>All the feelings that an individual has about his job</td>
<td>Grunenberg</td>
<td>1976</td>
</tr>
<tr>
<td>The affective orientation that an employee has towards his or her work</td>
<td>Price</td>
<td>2001</td>
</tr>
</tbody>
</table>

1-2. Artificial Neural Networks (ANN)
ANNs mimic the ability of the biological neural systems in a computerized way by resorting to the learning mechanism as the basis of human behavior [7]. ANNs are information-processing systems that generate output values based on specific logic. The ANNs ‘learns’ the governing relationships in the input and output data by modifying the weights between its nodes.

In essence, a trained Neural Network (NN) model can be viewed as a function that maps the input vectors onto the output vectors. ANNs can be applied to problems with non-linear nature or with too complex algorithmic solution to be found. Their ability to perform complex decision-making tasks without prior programming tasks makes ANNs more attractive and powerful than parametric approaches especially for nonlinear problems. ANNs have at least two potential strengths over the more traditional model fitting technique such as parametric methods and multiple regression analysis. First, ANNs can detect and extract nonlinear relationships and the interactions among inputs and outputs. Second, they are capable of estimating a function without requiring a mathematical description of how the output functionally depends on the inputs and other irritating assumptions of parametric methods. The architecture of a NN model usually consists of three parts: an input layer, hidden layers and an output layer. The information contained in the input layer is mapped into the output layer through the hidden layers.

Each neuron can receive its input only from the lower layer and send its output to the neurons on the higher layer only. Fig. 1 illustrates the network architecture of a NN model consisting of one hidden layer with S neurons along with the input and output layers with R and one neuron. Here the input vector p with R input elements is represented by the solid dark vertical bar at the left. Thus, these inputs post multiply S row, R column matrix W. One as a constant value enters the neuron as an input and is multiplied by a vector bias b with S row and single column. The net input to the
transfer function $f$ is $n$, the sum of the bias $b$ and the product $Wp$. This summation is passed to the transfer function $f$ to get the neuron’s output $a$, which in this case is a $S$ length vector [8].

$$
\begin{array}{c}
\text{R} \\
\times S
\end{array}
\begin{array}{c}
W
\end{array}
\begin{array}{c}
S
\end{array}
\begin{array}{c}
S
\end{array}
\begin{array}{c}
1
\end{array}
\begin{array}{c}
\times 1
\end{array}
\begin{array}{c}
b
\end{array}
\begin{array}{c}
S
\end{array}
\begin{array}{c}
1
\end{array}
\begin{array}{c}
\times S
\end{array}
\begin{array}{c}
W
\end{array}
\begin{array}{c}
S
\end{array}
\begin{array}{c}
1
\end{array}
\begin{array}{c}
\times 1
\end{array}
\begin{array}{c}
S
\end{array}
\end{array}

Fig. 1. The typical network architecture of a NN model consisting of one hidden layer

with $S$ neurons along with input and output layers with $R$ and one neuron.

1-3. System dynamics

System Dynamics was developed by Forrester [9] and it is designed to analyze the dynamic behavior of complex feedback systems. It was used as a preferred analysis methodology due to the fact that it explicitly can present the relationship between the variables in a non-linear, dynamic feedback system [20]. It has been used extensively in the modeling of various systems. It is a methodology based upon the systems’ theory. Systems theory has been widely used to study various aspects of an organization such as individual and group behavior [10] but there is few studies that relate system dynamics to job satisfaction. The steps of system dynamics model can be explained as a follow diagram according to Fig. 2:

Survey the related references and interview with the experts for identifying influential variables

1: Influence diagramming

2: Flow Diagram

3: Systems Equations

4: Model Validation

Fig. 2. The procedure for system dynamics

(i) Influence Diagramming. This involves identification of a sequence of pair wise causal hypotheses between variables or components of the system as well as their direction of influence. These variables are connected by arrows and plus or minus signs. A positive relationship (an arrow with a plus sign) occurs when, other things remaining the same, a change in one variable results in a change in another variable in the same direction, whereas a negative relationship occurs when the change is in the opposite direction. The systematic procedure for constructing an influence diagram has been provided by Wolstenholme and Coyle [11] and Roos and Hall [12].

(ii) Flow Diagram. This consists of symbols borrowed from control engineering to represent levels, rates and flows. The levels are the accumulators or state variables, the rates are the decision points that control the flow into and out of the levels, and information flows represent the decision structure surrounding each rate system regulator.

(iii) Systems Equations. Time difference equations represent the rates and flows described in the flow diagram in mathematical form using simulation languages such as DYNAMO, DYSMAP and STELLA.

(iv) Model Validation. In model validation the model is tested for structure (structure verification, parametric verification, extreme conditions, boundary adequacy and dimensional consistency) [24]. The advantages of system dynamics approach include [13]:

(i) It is based on a complementary computer simulation language, such as DYNAMO, which is easy to learn and is also available on microcomputers, and provides an excellent diagnostic tool to the novice programmer.

(ii) It allows analysts to maintain a one-to-one correspondence between the verbal description of the real world system and the influence diagram representing the cause and effect chain, and between the influence diagram and the set of equations in the computer program which simulate the model.

2. Literature Review

2-1. Related factors of job satisfaction

In order to identify the literature relating to job satisfaction, nurses and factors, we search through electronic databases (www.elsevier.com). There are a lot of researches in this filed. For example the abstracts or full texts of the papers were based on job satisfaction are more than 1,189 published research papers. Each of these papers were examined a lot of related factors of job satisfaction. For example, Spector (1997)[14] summarized the following facets of job satisfaction: appreciation, communication, co-workers, fringe benefits, job conditions, nature of the work itself, the nature of the organization itself, an organization’s policies and procedures, pay, personal growth, promotion opportunities, recognition, security and supervision. Because of variety in factors, we summarize these factors in Tab. 2. For example, Brown and Peterson (1993) identify individual level demographic and dispositional variables, role perceptions, supervisory behaviors, and job...
characteristics as influences on employees’ job satisfaction [15].

### Tab. 2. Summary of included studies regarding Related factors of job satisfaction at a hospital clinic

<table>
<thead>
<tr>
<th>Factors</th>
<th>Researchers</th>
<th>Influence</th>
<th>D/ID</th>
<th>Direct influence</th>
</tr>
</thead>
<tbody>
<tr>
<td>job stress</td>
<td>Shader K, Broome M, Broome C, West M, Nash M (2001), Fletcher C (2001), Davies et al. (2006), Davies et al. (2006), Hood (1997), Statovoky, 1992; Sarmento et al., 2004; Davies et al., 2006</td>
<td>+</td>
<td>D</td>
<td></td>
</tr>
<tr>
<td>work schedule</td>
<td>Shader K, Broome M, Broome C, West M, Nash M (2001)</td>
<td>+</td>
<td>D</td>
<td></td>
</tr>
<tr>
<td>group cohesion</td>
<td>Shader K, Broome M, Broome C, West M, Nash M (2001)</td>
<td>+</td>
<td>ID</td>
<td>Work schedule</td>
</tr>
<tr>
<td>promotion</td>
<td>Brown and Peterson (1993)</td>
<td>+</td>
<td>ID</td>
<td>Task Conflict, Social Conflict, Cohesion</td>
</tr>
<tr>
<td>fringe benefits, Praise and recognition</td>
<td>Tzeng (2002a, b), Wang (2002)</td>
<td>+</td>
<td>ID</td>
<td>Turnover</td>
</tr>
<tr>
<td>professional status(experienced)</td>
<td>Cowin L (2002)</td>
<td>+</td>
<td>ID</td>
<td>Task Conflict, Social Conflict, Cohesion</td>
</tr>
<tr>
<td>job security</td>
<td>Bailey (1995)</td>
<td>+</td>
<td></td>
<td></td>
</tr>
<tr>
<td>hardness</td>
<td>Fletcher C (2001)</td>
<td>+</td>
<td>D</td>
<td></td>
</tr>
<tr>
<td>ambiguity</td>
<td>Nathan A. Bowling(2008)</td>
<td>-</td>
<td>D</td>
<td></td>
</tr>
</tbody>
</table>
Last three columns of Tab. 2 can help researchers in identifying possible future studies or their similarities and dissimilarities to others research works. By this table we can find a sequence of pair-wise causal hypotheses between factors of the system. The abbreviations used are described beneath table 2.

2-2. Effects of Job Satisfaction
According to current literature review some of the researches have focused on the effects of job satisfaction on the outcomes of organisations. In this section, we summarized the effects of job satisfaction to find the relationship between the effects and the factors of job satisfaction.

The result of this section presented by Tab. 3 can help researchers to understand the system dynamics models presented in the next section. For instance Boles’ research shows that supportive aspects of the work environment has a positive influence on the job satisfaction and it has a negative influence on the turnover [16]. On the other hand, work stress, conceptualized as work-role conflict, work-role overload, and work-role ambiguity, is related to the job satisfaction negatively and to intend of leave organization positively. According to prior researches a number of outcomes of job satisfaction have been investigated including turnover intentions, absenteeism, and motivation (for example Tett & Meyer, 1993)[17]. Tab. 3 shows the summarized of these results which it’s obvious that we can recognize the relation between job satisfaction and turnover.

<table>
<thead>
<tr>
<th>Effects</th>
<th>Researchers</th>
<th>influence</th>
</tr>
</thead>
<tbody>
<tr>
<td>Turnover</td>
<td>Christian (1986)</td>
<td></td>
</tr>
<tr>
<td>Intention to leave</td>
<td>Staurovsky (1992)</td>
<td></td>
</tr>
<tr>
<td>Burnout</td>
<td>Cavanagh and Coffin (1992)</td>
<td></td>
</tr>
<tr>
<td>Recruitment and retention</td>
<td>Marriner and Craigie (1977)</td>
<td></td>
</tr>
<tr>
<td>Productivity</td>
<td>Lenz and Waltz (1983)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Wakefield-Fisher (1987)</td>
<td>+</td>
</tr>
<tr>
<td>Absenteeism</td>
<td>Staurovsky (1992)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Nathan A. Bowling (2008)</td>
<td></td>
</tr>
</tbody>
</table>

In summary, the research findings are inconsistent regarding the relationship between job satisfaction turnover, intention to leave, burnout, and productivity.

2-3. System Dynamics and Job Satisfaction
The number of studies related to the job satisfaction and System dynamics is few. Gupta et al. (1990)[18], Mutuc.(1994)[19] and Holmström (2004) have conducted researches on job satisfaction using system dynamics concepts. These models are about the applications of system dynamics and information system employee [20,21], employee participation and job satisfaction in health care industry [22-23]. The first is about a process model for studying the impacts of role variables on turnover intentions of information systems personnel. This model demonstrates the usefulness of the system dynamics approach as a prototype methodology to deal with dynamic aspects of the organizational behaviors. The model provides an analysis of both the steady state behavior and the transient behavior of the turnover intentions model. This methodology studies dynamic consequence of
relationships among antecedents of job satisfaction and organizational commitment, and management policy variables. The second present an initial investigation of interacting factors in participation, and its construct, motivation. The model were developed using recognized relationships in social science literature, and the third is about job satisfaction at a maternity department and application of system dynamics to explore staff retention. Some of these researches are complicated due to many factors. Based on past research, there are many methods such as FA, PCA, and ANN for reducing the number of factors. Considering the main purpose of this paper, the ANN method are used to determine the most effective factors and the results of the ANN is used for modeling system dynamics.

3. The Research Methodology

In this section, we present the main structures of the proposed model according to Fig. 3. In the proposed evaluation model, the major factors of the model were identified with the help of ANN model (see Fig. 3). An ANN model was implemented and trained using collected data by Surveying Questionnaire in order to determine the most important related factors of job satisfaction. The details of ANN procedure have been shown in Fig. 3-b and it is explained briefly in section 3.1. By the result of ANN, we can achieve necessary input of system dynamics.

Developing the dynamic model based upon the ANN finding is the next stage of the proposed model. One of the most important things which the top management needs to select the best strategy for encourage of staff. Three scenarios is developed for analyses the influence of different strategy about the kind of reward. Financial reward, nonfinancial reward and both of them are the three elements of every scenarios of dynamic model which are explained in section 5.

3-1. The Procedure of NN Modelling

Selecting the most effective job satisfaction factor using artificial neural network is performed in two steps as described below (see Fig. 3-b): The first step is constructing various NN models of job satisfaction level as an output which vary in their inputs (the effective factors) then in the second step, is the selection of the best effective factors which are related to the most fitted model based on the performance indicators.

Phase 1. Selecting the neural network model topologies: There are 31 different NN topologies, which are being employed in researches at present time. The most common networks are: Multilayer Perceptrons (MLP) also called Multilayer Feed forward Networks, Adaptive Resonance Theory Models (ART),

![Diagram](image-url)
Recurrent Associative Networks (RAN), and Self-Organizing Maps (SOM) [23].
In this phase, models topology and their training algorithms considering the nature of modeling problem are chosen. The other important thing addresses the definition of input variables (e.g., effective factors) and output variables (e.g., Job satisfaction level) for assigning them to the network of each model as inputs and outputs, respectively. Also, the way of presenting data to the network have to be determined. One way to do this is to present all available process variables as network inputs, and then let the network modify itself during training so that the connection of any insignificant variables becomes weak. Another approach is to be more selective and introduce as inputs only those variables that are surely affecting on the process outputs. The first approach is called the “global network” while the second is termed the “focused network” [24].

Phase 2. Collection and preparation of training data set: In the Data collection stage, it is necessary to ensure the sufficiency and integrity of the data used to train and test the network, whereas the network performance can be directly influenced by the presented performance. It is not simply possible to say how many data sets are required, because this depends on the nature of the process modeling problem and the cost of providing data. The prepared data set is categorized randomly into two sets: Training data set and test data set. Generally, the training data should cover all of the data variation range, unless there is a good reason to resort to stratification or data blocking. In order to enhance the model fitness, several tasks for preparing data may be performed such as: (1) data integrity check, (2) extreme data removal, (3) data scaling, and (4) data coding.

Phase 3. Constructing and fitting the NN models: Construction of NN model architecture includes determining few features:
- The number of layers;
- The number of neurons in each layer;
- Each layer’s transfer function;
- How the layers are connected to each other.

So the performance of a NN model depends on the fitness of the network features. There are various training algorithms to fit NN models. The most popular algorithm in optimization and estimation applications is the standard back propagation (BP) [25]. This algorithm is a widely used iterative optimization technique that locates the minimum of a function expressed as

$$E = \frac{1}{2} \sum_{m} (y_{dm} - y_{m})^2$$  \hspace{1cm} (1)

Where $y_{dm}$ is the target output value of the output layer, and $y_m$ is the value of the output layer. Based on BP algorithm, during the training process, the deviation between the network output and the desired output at each presentation is computed as an error. This error, in quadratic form, was then fed back (back propagated) to the network and used for modifying the weights by a gradient descent method.

The training process carries on while one of three user-specified conditions is met at least. These conditions consist of (1) exceeding the maximum number of epochs, (2) meeting the performance goal, and (3) decreasing of the gradient descent rate to the less than the allowable limit.

Phase 4. Model validation: In this phase, all of fitted models are validated. The validation data set is used to ensure that there is no overfitting in the final result. In order to validate the models, a data set is selected randomly from training data. When a significant overfitting has occurred, the error of validation data starts to increase and the training process is stopped.

Phase 5. Selection of performance indicators: To find out the most reliable model, several performance indicators can be used. The performance of the NN models based on their reliability is evaluated by a regression analysis between the predicted values by the NN models and the actual values. The indicators used with aim of performance evaluation for both training and test data sets are the root mean square error (RMSE) and correlation coefficient [26]. The root mean square error is calculated by:

$$RMSE = \sqrt{\frac{1}{m} \sum_{i=1}^{m} (\hat{y}_i - y_i)^2}$$  \hspace{1cm} (2)

Where $\hat{y}_i$ is the predicted value by NN model, $y_i$ is actual value and $m$ is the number of points in the data set. The absolute fraction of variance, a statistical criterion that can be applied to multiple regression analysis, is calculated by:

$$R^2 = 1 - \frac{\sum_{i=1}^{m} (y_i - \hat{y}_i)^2}{\sum_{i=1}^{m} (y_i - \bar{y})^2}$$  \hspace{1cm} (3)

The absolute fraction of variance ranges between zero and one. Ideally, R2 should be close to one, whereas a poor fit results in a value near zero.

Phase 6. Model performance evaluation: By termination of process training and validation, the networks are ready for prediction. So, the input vectors...
from the separate test data are introduced to the trained network and the responses of the network, i.e., the predicted output, are compared with the actual ones using the performance indicators. The most reliable model for the job satisfaction level is chosen in respect to the evaluation and training results. Hence, the factors related to the chosen model are the best effective factors.

4. The NN Modeling of Job Satisfaction

According to result of literature, there are a lot of effective factors which cause complexity of collecting data as an input of ANN model. For avoiding this problem we use the Job Descriptive Index (JDI) for measuring job satisfaction level. The JDI is a measure of job satisfaction that presents adjectives thought to relate to satisfaction with one of five aspects of jobs: Pay (PA-9 items), Work (WO-18 items), Opportunities for Promotion (OP-9 items), Supervision (SU-18 items), and Co-Workers (CW-18 items). Examiners respond to the adjective by endorsing either “Yes,” “No,” or “?"[27], scored 3, 0, and 1, respectively in addition to, the JDI is the most commonly used measure of job satisfaction in the world. Data were obtained from the JDI at large hospital in Iran.(JDI). In this study, a group of eligible members of nurses have completed the JDI.

The group consists of 250 nurses including staff nurses who worked on patient unit. Full-time and regular part-time employees on all shifts were included and managers are excluded. The 200 registered staff nurses at work in the hospital were reached by an online JDI which designed under excel program. The response rate, however, was %84 with a total number of 250 nurses completing useable questionnaires. Therefore, the collected data consisted of 212 data vectors. As, each vector consisted of five data elements related to the values of the effective factor (EF) of job satisfaction and one data element relevant to their corresponding job satisfaction level (JSL). The 212 data vectors were randomly split into 180 training data vectors and 30 test data vectors.

Seven models for the mentioned JSL with different inputs were proposed (Further information is presented in Tab. 4). The topologies of seven NN models was MLP trained by BP algorithm based on Levenberg-Marquart rule [28]. In order to choose the optimal configuration for each network, a number of different network configurations, consisting of one to three hidden layers and different number of neurons in hidden layers with various transfer functions were considered.

The training process was run under the MATLAB environment. A minimum of the user-specification error function i.e., RMSE of 1e-001 was reached while the number of epochs is less than 1000 and grad rate is more than 1e-010. The optimal configuration of each training network was selected based on the degree of model fitness which is demonstrated by the value of error function. As Fig. 4 indicates when the number of neurons in first and second hidden layers were equal six and nine respectively, it was the most of the training RMSE of network for model 5.

![Figure 4: Effect of number of neurons in first and second hidden layers on training RMSE for model 5](image)

| Tab. 4. The features of the configurations for networks of seven proposed models |
|-----------------|--------|-----------------|-----------------|
| Models  | Selected EFs  | MLP structure  | Transfer function |
| Model 1  | PA         | 1-4-6-1         | Tansig-logsig- purline |
| Model 2  | WO         | 1-4-1           | logsig- purline |
| Model 3  | CW         | 1-4-1           | logsig- purline |
| Model 4  | OP, SU     | 2-4-1           | logsig- purline |
| Model 5  | PA, WO     | 2-6-9-1         | purline-tansig- purline |
| Model 6  | SU, CO     | 2-8-8-1         | purline-logsig- purline |
| Model 7  | PA, WO, CO | 3-8-9-1         | purline-logsig- purline |

(Pay =PA, Work=WO, Opportunities for Promotion=OP, Supervision =SU, and Co-Workers=CW)
Tab. 4 shows the features of the configuration for networks of seven NN models and the result of training process relevant to models are demonstrated in Tab. 5. So, in order to find the most reliable model, all 30 test data vectors were used for model evaluations. The test data vectors were customized and presented to each model. The results of the model evaluation using performance indicators for seven models are provided in Tab. 5.

<table>
<thead>
<tr>
<th>Models</th>
<th>Training RMSE</th>
<th>Testing RMSE</th>
<th>R2</th>
<th>Priority</th>
</tr>
</thead>
<tbody>
<tr>
<td>Model 1</td>
<td>0.0021</td>
<td>0.0012</td>
<td>0.9524</td>
<td>4</td>
</tr>
<tr>
<td>Model 2</td>
<td>0.0789</td>
<td>0.0824</td>
<td>0.5068</td>
<td>3</td>
</tr>
<tr>
<td>Model 3</td>
<td>0.0812</td>
<td>0.0852</td>
<td>0.2713</td>
<td>5</td>
</tr>
<tr>
<td>Model 4</td>
<td>0.0897</td>
<td>0.0910</td>
<td>0.2512</td>
<td>7</td>
</tr>
<tr>
<td>Model 5</td>
<td>0.0423</td>
<td>0.0513</td>
<td>0.3825</td>
<td>1</td>
</tr>
<tr>
<td>Model 6</td>
<td>0.0845</td>
<td>0.0897</td>
<td>0.2632</td>
<td>6</td>
</tr>
<tr>
<td>Model 7</td>
<td>0.0493</td>
<td>0.0527</td>
<td>0.3427</td>
<td>2</td>
</tr>
</tbody>
</table>

With regard to the results of Tab. 5, “PA, WO”, “PA, WO, CO” , “WO”, “PA”, are four most effective factors. Because the relevant model (i.e., the model 5) is the most reliable one considering its minimum value for RMSE (both training and testing) and also its maximum value for R2. The results of proposed ANN model can help us in designing system dynamics models. The results of the ANN model shows that three main factors which influence on job satisfaction are PA, WO, and CW. Therefore, we can use these main factors in modeling our system dynamics.

5. The Proposed System Dynamics Model and its Interface

Causal loop (influence) diagrams exhibit the structure of a system in the system dynamics methodology. Causal loop diagrams play four important roles in SD. First, during model development, they can show the sketches of causal hypotheses, secondly, they can simplify the representation of a model, thirdly we can have different scenarios, and fourthly the major feedback mechanisms are captured by a causal loop diagram. These mechanisms are either positive feedback (reinforcing) or negative (balancing) loops. In a positive feedback loop an initial disturbance leads to further change, suggesting the presence of an unstable equilibrium. A negative feedback loop exhibits a goal-seeking behavior [29]. In this section we present job satisfaction Causal loop diagrams according to the various studied models based on the previous researches and the ANN result in three different scenarios. This model contains two important kinds of rewards which top management may use for increasing staff motivation. In the next sections we will analyze all these scenarios.

5-1. Scenario 1 – Financial & Non Financial Reward

In Fig. 5 we proposed casual loop according to previous finding for job satisfaction. The arrows represent the relations among variables. The direction of the influence lines displays the direction of the effect. Signs “+” or “−” at the upper end of the influence lines exhibit the sign of the effect. When the sign is “+”, the variables change in the same direction; otherwise they change in the opposite one. In this influence diagram (Fig. 5), several distinct loops can be observed, which control the job satisfaction. We can use this diagram for determining the rate of job satisfaction and also we can determine the kind of reward (financial and non financial rewards) in order to achieve the best rate of job satisfaction in healthcare organization. In addition we can make different scenarios for determining the best situation.

Fig. 5. Proposed influence diagram for healthcare's staff job satisfaction
Four main loops of this model are shown as loop#1, loop#2, loop#3, and loop#4. Two of them contain financial reward and two others have nonfinancial reward (Fig. 5).

The first loop (Fig. 6) shows that the job satisfaction has a negative influence on the job stress and vice versa, in the other word when job satisfaction increases, the job stress will decrease and when job stress increase the job satisfaction will decrease. Job stress has a positive influence on the work pressure which has a counteractive effect on the work itself (Workload; scheduling; challenging work; routinization; task requirements) and productivity. Therefore, when work pressure is high the work itself will be low and when the work itself is good the productivity will be high. The productivity has a positive influence on non-financial reward. When productivity increases so does the non-financial reward. The non-financial reward influences the expectation in the same direction (+). Therefore, as the non-financial reward increases the staff expectation will also increase. As staff expectation increases the influence on the perceived result is negative. Finally the perceived result has a counteractive (negative) effect on the job satisfaction. The above loop is positive as it involves a reinforcing influence. The overall influence of the loop can also be checked by multiplying the individual influences \((+)(+)(+)(-) = (+)\). Due to the fact that this is a positive type loop it will always try to increase the job satisfaction. Other factors which have positive influence on the job satisfaction, as shown in Fig. 5, are Culture, supervision co-workers and officialism.

The third and fourth loops represent the financial reward and job satisfactions. Loop#3 contains job satisfaction, quality of service, service cost, salary, financial reward, expectation, and perceived result. As Fig. 8 suggests, the overall influence of the loop can be determined by multiplying the individual influences \((+)(+)(+)(+)(-) = (-)\). Due to the fact that this is a negative type loop it will try to reduce the level of job satisfaction.

As can be seen from diagram in Fig. 9, job satisfaction is dependent upon the Perceived Result. It shows that the Perceived Result has a negative influence on the job satisfaction. The expectation has a negative influence on the perceived result which has affected by financial reward and income. In other words when productivity increases the income and thus the financial reward and expectation will also increase. When expectation and perceived results increase, job satisfaction would be decreased. Although in loop#4 (Fig. 9) there are some negative arrows the overall influence of the loop is positive.
The effects of any factors especially financial reward on job satisfaction is studied. In scenario 1, there are two distinct loops: financial reward and non-financial reward which is examined simultaneously. In scenarios 2 and 3, we study every kind of reward separately. In scenario 2, it is assumed that financial reward is that kind of reward which management wants to allocate. In this scenario the effect of financial reward on job satisfaction is studied. In Fig. 10, when income increases due to increasing productivity, the financial reward will increase. Two main effects of financial rewards are increasing expectation and increasing salary. Each of these effects can influence on the level of job satisfaction indirectly. Service cost increases when salary or quality service increases. The two most important loops in Fig. 10 are shown as loop#3 and loop#4 which were discussed in some more detail before. The overall influence of the loop#3 is negative and of loop#4 is positive. Therefore, in this model there are balancing and reinforcing loops. According to this model we can analyze all loops and study the effects of any factors especially financial rewards on job satisfaction.

5.2. Scenario 2—Financial Reward
In scenario 1, there are two distinct loops: financial reward and non-financial reward which is examined simultaneously. In scenarios 2 and 3, we study every kind of reward separately. In scenario 2, it is assumed that financial reward is that kind of reward which management wants to allocate. In this scenario the effect of financial reward on job satisfaction is studied. In Fig. 10, when income increases due to increasing productivity, the financial reward will increase. Two main effects of financial rewards are increasing expectation and increasing salary. Each of these effects can influence on the level of job satisfaction indirectly. Service cost increases when salary or quality service increases. The two most important loops in Fig. 10 are shown as loop#3 and loop#4 which were discussed in some more detail before. The overall influence of the loop#3 is negative and of loop#4 is positive. Therefore, in this model there are balancing and reinforcing loops. According to this model we can analyze all loops and study the effects of any factors especially financial rewards on job satisfaction.

Fig. 9. Causal loop diagram of financial reward (loop#4)

Fig. 10. Influence diagram contain financial reward for scenario 2

Fig. 11. Influence diagram contain non-financial reward for scenario 3
In this scenario the effect of financial reward on job satisfaction is studied. In Fig. 10, when income increases due to increasing productivity, the financial reward will increase. Two main effects of financial rewards are increasing expectation and increasing salary. Each of these effects can influence on the level of job satisfaction indirectly. Service cost increases when salary or quality service increases. The two most important loops in Fig. 10 are shown as loop#3 and loop#4 which were discussed in some more detail before. The overall influence of the loop#3 is negative and of loop#4 is positive. Therefore, in this model there are balancing and reinforcing loops. According to this model we can analyze all loops and study the effects of any factors especially financial rewards on job satisfaction.

5.3. Scenario 3 – Non Financial Reward
Top managements believe that increasing non-financial reward has better impacts on job satisfaction in comparison with financial reward. Scenario3 is designed as Fig. 11 in order to survey the effect of nonfinancial reward on job satisfaction. In this scenario, when productivity increases the nonfinancial reward will increase as well. When non-financial reward increases, two factors (expectation and responsibility) also become high. In loop#1 increasing responsibility cause increasing task conflict and then it cause decreasing of job satisfaction. The overall influence of this loop is negative while the overall influence of loop#2 is positive. The detail of these two loops is shown in Fig. 6 and Fig. 7.

6. Summary and Conclusion
Determining the influenced factor on job satisfaction in service organization especially in health industry is so important for managers. Dissatisfaction at hospital has increased and it causes bad effect on organizations’ outputs such as productivity. Job satisfaction has been identified as a key factor in employees’ turnover with the empirical literature suggesting that it is related to a number of organizational, professional and personal variables which is summarized in Tab. 2 and Tab. 3.
In this study we can find some important factors which are influenced on job satisfaction (Tab. 2) and important effects of job satisfaction on organizations’ outputs (Tab. 3). The second important finding of this paper is, finding of confliction and similarity between factors and effects based on literature review. For example in Table 2 there are some studies which have different results. According to Lus’ study, educational level has positive influence on job satisfaction, while according to Ramburs’ study (2003), this factor has negative effect on job satisfaction. Result of Matrunolas’ study (1996) shows job satisfaction has no significant influence on absenteeism and Sius’ study (2002) shows that job satisfaction has negative influence on absenteeism. These are two important conflicts among JS’s studies. Therefore, although several models of job satisfaction have been postulated, these models require further testing especially regarding the relative contribution of different factors and effects by some tools such as system dynamics. The literature suggests that the current models of job satisfaction need to be modified as they omit some important factors or effects of job satisfaction such as role perception. The lack of a comprehensive model of job satisfaction and lack of using system dynamics in job satisfaction is a major shortcoming. Therefore the current worldwide dissatisfaction at hospitals highlights the importance of understanding the impact and interrelationships of the identified variables by system dynamics.

In this study, we presented a system dynamics based on ANN result for mapping and analyzing the job satisfaction evaluation and effect of kind of reward on it. The methodology constructs ANN for selecting best factors base on JDI. The evaluation weights provided by ANN can be applied as the criteria for selecting the important factors. The system dynamics designed based on the ANN result which it can be used to identify effective policies for selecting the kind of reward. The methodology has been implemented for the large hospital in Iran. Three main scenarios are developed to survey effect of different factors (i.e. financial and nonfinancial rewards and both of them) on job satisfaction. The developed model can be further used to simulate various scenarios. The model can further be used in a wide range of healthcare system.

References


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