

Human Capital Model

February 9, 2015

Contents

1	Introduction	2
1.1	Restatement of the Problem	2
2	Human Capital Model	3
2.1	Assumptions and Justifications	3
2.2	Model Overview	3
2.3	Human Performance Model	3
2.4	Social Network Model	5
2.5	Promote and Churn Model	6
3	The Improved Model	7
3.1	Assumptions and Justifications	7
3.2	Organizational Churn	7
3.3	Productivity	7
4	Performance and Analysis	8
4.1	Analysis for Task 3	8
4.2	Analysis for Task 4	9
4.3	Analysis for Task 5	10
5	Advice for HR	11
5.1	Incentive Mechanism	11
5.2	Matching Employees to the Right Position	12
6	Team Science	13
7	Sensitivity Analysis	14
8	Strengths and Weaknesses	14
9	Conclusions	14

1 Introduction

Considering the shortage of the talent, it's essential for companies to retain good people and make them well-trained. However, current situation is not satisfactory while many talents always tend to get a good job via job-hopping, causing organizational churn in employees who are closely connected to them. In order to simulate this process and improve it, we build a human capital model based on Social Network Analysis and Markov process.

1.1 Restatement of the Problem

Problem C describe ICM as an organization of 370 people on human capital. The company is facing some issues. ICM tend to make a positive work atmosphere to gain the loyalty of an employee in their early time. The churn of an some employee makes effect on productivity and emotion on other employee. Lower quality employees often stay with the company for a full career. The churn rate is 18% per year and mid-level positions suffer high churn rate, even twice the average rate of the rest of the company. HR prefer matching employees to the right position. It only provides promotion opportunity for those who have several years of work experience. ICM usually has only 85% of its 370 positions filled at any time and HR office is actively hiring about 8%-10% of the ICM positions. The CEO salary is approximately 10 times the median of the salaries of all the employees, compared to hundreds of times the median as found in many companies.

There are several tasks we need to finish according to our model:

- Build a Human Capital network model of ICM organizations personnel situation.
- Identify dynamic processes within the Human Capital network, which includes organizational churn and direct and indirect effects on the organizations productivity.
- Analyze your organizations budget requirements over the next 2 years.
- Judge if ICM could sustain its 80% full status for positions if the annual churn rate for all positions goes to 25% and 35% respectively. Calculate the costs of these higher turnover rates and describe the indirect effects of these high churn rates.
- Simulate the impact of 30% churn in both junior managers and experienced supervisors, if other churn values remain at 18%. It assumes that there is no external recruiting and ICM promote only qualified employees for the next two years.
- Connect our Human Capital network to other organizational network layers such as information flow, trust, influence, and friendship that the other offices of ICM are considering building.

2 Human Capital Model

2.1 Assumptions and Justifications

- **assumption 1** If an employee has probability to promote, he won't churn.
The possibility of the unforeseen accidents, which could force an employee to leave his position, is neglected. Based on human nature, an employee will stay at his position to chase for higher level.
- **assumption 2** For a vacancy, if there exists an employee measures up to it already, ICM won't recruit for it.
Since recruiting good people is difficult, time consuming and expensive according to issue 5, it is wasteful to recruit for a position if an employee can promote to it.
- **assumption 3** Demotion won't occur.
- **assumption 4** Administrative clerk won't promote or be transferred.
- **assumption 5** Each division or office have at least one middle manager or senior manager.

2.2 Model Overview

Most research for human capital can be classified as either microscopic and macroscopic. Since either macroscopic or microscopic methods are difficult to solve our problem perfectly, we approach the problem with the combination of macroscopic and microscopic methods.

To measure the ability of each person, we use Quantitative Management Performance. Via this measurement, we can classify different kinds of employees which will influence the promotion process.

To build the employee network and analyze its properties, we employ the social network analysis (SNA) technique. In our case, the employees are viewed as nodes and relationships as links among them. So, We can simulate the complex relationship via this network.

Definitions of symbols employed in this section are listed in Table 1:

2.3 Human Performance Model

In this part, we build a people model to evaluate an employee in four aspects, in terms of **Quantitative Management Performance** — work achievement, work ability, work attitude and potential. These four aspects are supposed to be quantized according to annual evaluation based on performance as judged by the supervisor and we take these independent variables as A_{ac_i} , A_{ab_i} , A_{at_i} and A_{po_i} for each employee i . For each of the four parameters, it goes from 0 to 1. The statues are used for calculating the probability that employee can promote. Meanwhile, they influence leaving probability and team cohesiveness as well.

Variable	Description
i	Index of an employee
$L_{i,t}$	The level of i
$s_{ij,t}$	The relation strength from i to j
$a_{ij,t}$	Represent the influence caused by relationship between superior and subordinate
$f_{ij,t}$	Represent the influence caused by relationship between person with many friends and person with few friends
$c_{i,t}$	Clustering Coefficient of i

Table 1: Variables and Descriptions

It is obvious that some of the parameters are somehow more important than others. So in an effort to make our model more accurate and reliable, we introduce a weighted index of deviation AD_i , with

$$A_i = w_{ac} \cdot A_{ac_i} + w_{ab} \cdot A_{ab_i} + w_{at} \cdot A_{at_i} + w_{po} \cdot A_{po_i} \quad (1)$$

We determine weights via the Analytical Hierarchy Process(AHP) [Saaty 1982]. We build a 4×4 reciprocal matrix by pair comparison:

	A_{ac}	A_{ab}	A_{at}	A_{po}
A_{ac}	1	5	2	1
A_{ab}	$\frac{1}{5}$	1	$\frac{1}{3}$	$\frac{1}{4}$
A_{at}	$\frac{1}{2}$	3	1	1
A_{po}	1	4	1	1

The meaning of the number in each cell is explained in Figure 2. The numbers themselves are based on our own subjective decisions.

Intensity of Importance	Definition
1	Equal Importance
2	Weak or slight
3	Moderate importance
4	Moderate plus
5	Strong importance
6	Strong plus
7	Very strong or demonstrated importance
8	Very, Very strong
9	Extreme importance

Table 2: The fundamental scale of absolute numbers

We then get the weight of each factor by calculating the biggest eigenvalue and its corresponding eigenvector, as given in Table 3.

We test the consistency of the preferences for this instance of the AHP. For good consistency [Alonso and Lamata 2006, 446 - 447]:

Factor	A_{ac}	A_{ab}	A_{at}	A_{po}
Weight	0.3805	0.0709	0.2371	0.3030

Table 3: Weight for factors

- The principal eigenvalue λ_{max} of the matrix should be close to the number n of alternatives, here 4; we get $\lambda_{max} = 4.047$.
- The consistency index $CI = (\lambda_{max} - n)/(n - 1)$ should be close to 0; we get $CI = 0.0157$.
- The consistency ratio $CR = CI/RI$ (where RI is the average value of CI for random matrices) should be less than 0.1; we get $CR = 0.0182$.

Hence, our decision method displays perfectly acceptable consistency and the weights are reasonable.

2.4 Social Network Model

The social network model contains a directed weighted graph $G(V, E)$ in which V denote the employees and E denote the connection between employees. Since there are personnel changes, $G(V, E)$ will change with time goes by. In order to simulate this situation, we use $G_t(V_t, E_t)$ instead of $G(V, E)$ where t is a discrete variable. So $G_t(V_t, E_t)$ denote the social network in the t -th month.

First, we explain the way we build edges of $G_t(V_t, E_t)$.

When $t = 0$, there are about $370 \times 85\%$ nodes (employees) in G_t . We build edges between employees in the same division or office, since employees in the same division or office certainly know each other. So each division or office form a complete graph and employees in the different division or office don't know others, which is impossible. To solve this problem, we build 10 edges for each employee with employees in other divisions with equal probability. Then we build the other edges with probability $p = \frac{|N_{i,t} \cap N_{j,t}|}{|N_{i,t} \cup N_{j,t}|}$ which is called Jaccard similarity coefficient[Jaccard 1901].

When $t > 0$, there will be employees leaving or joining the company. If an employee leave, all his edges with other employees will be deleted. If an employee newly join the company, he will follow steps which employees at $t = 0$ take. This is the dynamic process of graph $G_t(V_t, E_t)$.

Let $s_{ij,t}$ denote the weight from i to j at time t . We have these properties of $G_t(V_t, E_t)$:

- $s_{ij,t} \neq s_{ji,t}$

We made this graph directed and weighted because one person may consider another person his best friend while that person doesn't consider him a good friend. This situation may appear because of the relationship between superior and subordinate and the relationship between person with more friend and person with less friend. In general, $s_{ij,t} \neq s_{ji,t}$ for the reason above.

- $s_{ij,t} = \frac{a_{ij,t} + f_{ij,t}}{2}$

$a_{ij,t}$ denote the influence caused by the relationship between superior and subordinate, $f_{ij,t}$ denote the influence caused by the amount of friends for i and j . We define

$$a_{ij,t} = \begin{cases} \frac{1}{2+|L_{i,t}-L_{j,t}|}, & L_{i,t} \geq L_{j,t} \\ 1 - \frac{1}{2+|L_{i,t}-L_{j,t}|}, & L_{i,t} < L_{j,t} \end{cases}$$

, where $L_{i,t}$ denote the level of i at time t , and

$$f_{ij,t} = \frac{|N_{i,t} \cap N_{j,t}|}{|N_{i,t}|}$$

2.5 Promote and Churn Model

First, we explain the model of promotion aimed to predict the promotion probability. We define the promotion rate of i as p_i to evaluate the probability to promote. As a matter of fact, if there is a vacancy, judging if an employee suits the site involves work experience and ability. It is essential that he is supposed to have several years of work experience according to issue 6. If an employee satisfies the experience condition, it turns out to think about his ability. Since in Human Performance Model, each employee's ability is evaluated by a parameter A_{D_i} . For each level of position, it has an ability standard, as shows in Table[]]. The ability of an employee are supposed to reach the four standard parameters respectively, otherwise its p_i is 0. For those who reach the standard, the promotion probability can be calculated by the equation: $p_i = \frac{A_{D_i}}{\sum_{\alpha} A_{D_{\alpha}}}$ where α represent employee who have probability to promote.

According to assumption 1 and assumption 2, from t to $t+1$, the model update once according to the following rules:

- step 1* Promotion: If there's a vacancy in superior and i has the ability to occupy this position, let $L_{i+1,t} = L_{i,t} + 1$ with probability mentioned above and build edges between i and his new colleagues.
- step 2* Churn: If i isn't promoted, he will have the probability to churn with a churn rate l .
- step 3* Stay: If i hasn't been promoted or churned, he will keep his job and do nothing.
- step 4* Recruit: Recruit for vacancies when all the employees have done these 3 steps above.

3 The Improved Model

3.1 Assumptions and Justifications

- **assumption 1** For the promotion probability and organization change, the other factors effects the churn probability is invariable.

Though churn derives from varieties of reasons and they are actually lacking of known conditions and data to estimate them, we have to regard it as stable in our model.

3.2 Organizational Churn

The network of the company we've built changes in terms of orgnizational churn and promotion. Considering various of factors in reality, we build an orgnizational churn and promotion model to predict the dynamic process.

The first part of the model is churn model. We define the churn rate of an employee i as l_i to evaluate the probability to churn. l_i is usually controlled by sorts of factors to different degrees. We divides l_i into three parts: l_{i1} , l_{i2} and l_{i3} . l_{i1} represents the churn rate because of lacking of promotion opportunity. l_{i2} represents the churn rate because of the changes of other employees related to employee i . To simplify our model, we presume that l_{i1} , l_{i2} is linear correlated with $(1 - p_i)$ and s_i , which means

$$l_{i1} = \lambda_1(1 - p_i), l_{i2} = \lambda_2 s_i \quad (2)$$

λ_1 and λ_2 could be ensured in later calculation.

l_{i3} represents the other factors we can't get any information from the known conditons so that we regard it as stable. Thus

$$l_i = \lambda_1(1 - p_i) + \lambda_2 s_i + l_{i3} \quad (3)$$

After analyzing a great deal of churn rate reports, we get the composition of the three parts are 10.9%, 2.2% and 75.4% respectively. According to the percentage and the general churn rate 18%, we can calculate λ_1 , λ_2 and l_{i3} . As an orinigal condition, it satisfies

$$\begin{cases} \lambda_1 \sum_{i=1}^{370} (1 - p_i) &= 10.9\% \times 370 \times 1.5\% \\ \lambda_2 \sum_{i=1}^{370} s_i &= 2.2\% \times 370 \times 1.5\% \\ l_{i3} &= 1.5\% \times 75.4\% \end{cases} \quad (4)$$

Thus we can use Equation 4 to calculate the churn rate l_i .

Analyzing the consequence of this phenomenon, we use our program to simulate this process. As a result, we find a organizational churn in our simulation as shown in Figure 1

3.3 Productivity

First, we explain the concept of clustering coefficient [Duncan J. Watts and Steven Strogatz 1998] of i denoted as $c_{i,t}$, which is a measure of the degree to which nodes in a graph tend to cluster together. $c_{i,t}$ is defined as:

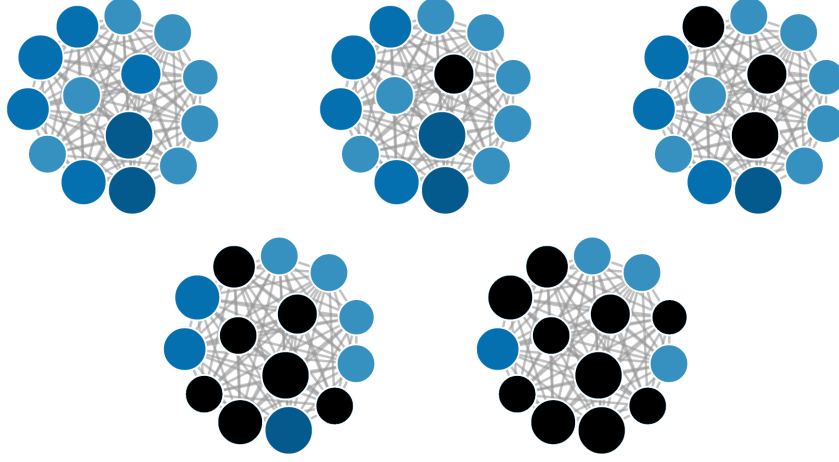


Figure 1: Organizational Churn

$$c_{i,t} = \frac{|\{s_{jk,t} : v_j, v_k \in M_i, s_{jk,t} \in E\}|}{k_{i,t}(k_{i,t} - 1)}$$

where $M_{i,t} = \{v_j : s_{ij,t} \in E, s_{ji,t} \in E\}$ and $k_{i,t} = |N_{i,t}|$.

4 Performance and Analysis

4.1 Analysis for Task 3

We assume that the company offers training programs for its employees monthly and newly hired employees start to get their salaries next month after they enter the company. With these two assumption, results can be drawn according to our model through simulation.

Budget can be divided into three parts: salary budget, training budget and recruiting budget. The budget requirement predicted for next two years is listed in the table below in terms of σ .

Total Budget	Salary Budget	Training Budget	Recruiting Budget
1170.8 σ	951.387 σ	164.423 σ	55.08 σ

4.2 Analysis for Task 4

To analyze the status of positions under different churn rate, we use our model to simulate dynamic processes with these churn rate constraints. We execute our program 100 times for each churn rate and average the predicted values. Figure 2 shows the averaged results our model predicted. Under all of these three conditions, the number of employees in the company keeps rising. The higher the churn rate, the lower the final full rate the company reaches after two years. But ICM can sustain its 80% for positions even if the churn rate goes to 35% according to our model's prediction.

The churn rate effect the budget of the company as well. The three parts of the budget behave differently when churn rate increases. The calculated budget

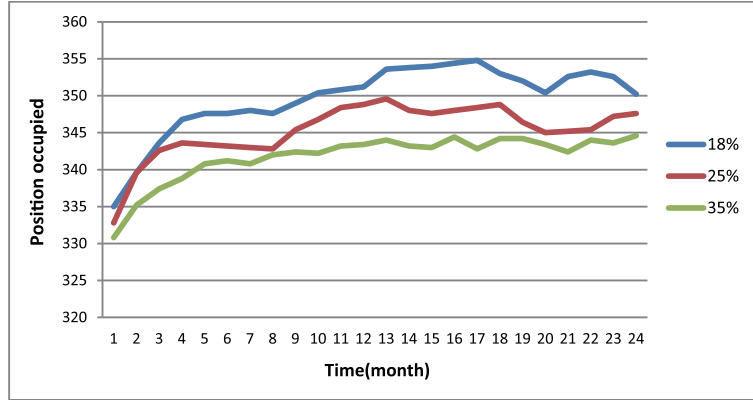


Figure 2: Status of positions

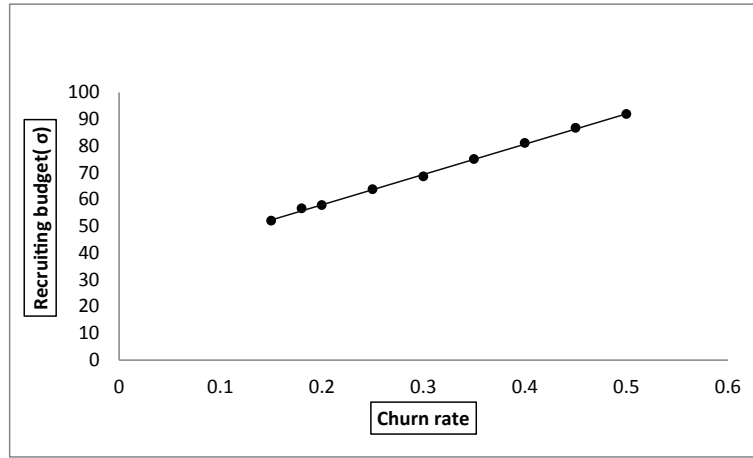


Figure 3: Recruiting budget

is shown in Figure 3 and Figure 4. Each data point in three charts is an averaged result of 10 predictions and a linear trendline is added to each chart. It is clear that recruiting budget showed in Figure 3 is likely to be proportional to the churn rate while salary budget and training budget showed in Figure 4 are likely to be inversely proportional to the churn rate.

To maintain enough employees, the company has to spend more on recruiting. So high turnover rate directly increase the recruiting budget. High turnover rate's effect on training budget and salary budget is more complex. On the one hand, when churn rate goes up, vacancies in the middle level keeps rising due to long recruiting time and low promote rate. On the other hand, the vacancies in lower level remains low because of the short recruiting time. So the full rate of the company decreases when churn rate rises. Since training budget and salary budget are closely related to full rate, both of them decrease when turnover rate goes up.

4.3 Analysis for Task 5

We apply following changes to our model to simulate the required process:

- Change the churn rate of junior managers and experienced supervisors to 30%

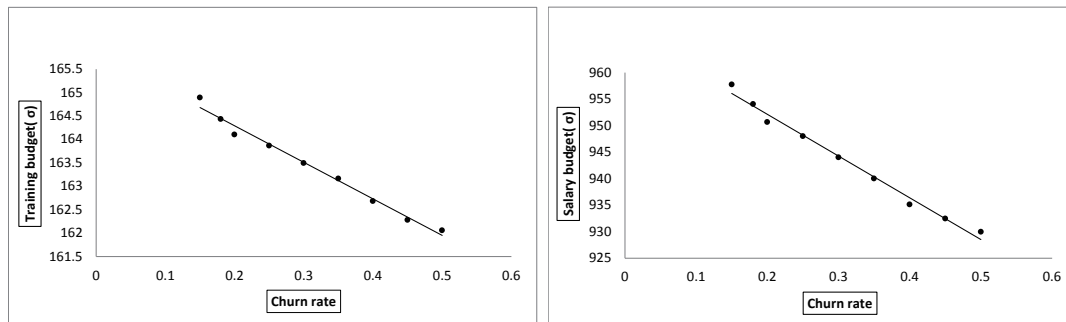


Figure 4: Training budget and salary budget

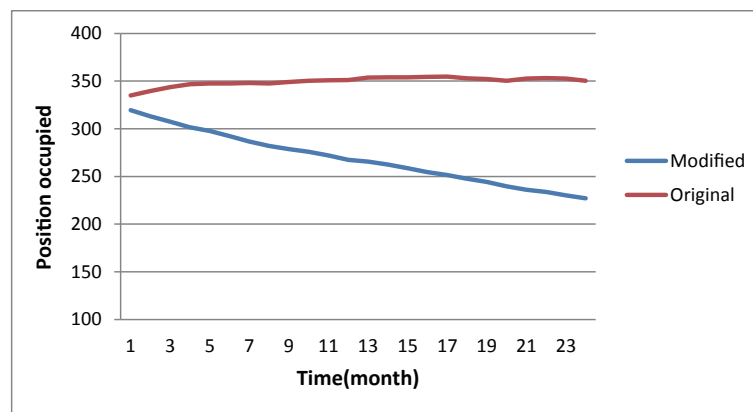


Figure 5: Status of position

- Prohit external recruiting
- Promoting only qualified employees

The result of simulation is shown in Figure 5 and Table 4. All the data shown is an average of ten predictions. While the number of positions occupied remains stable with original conditions, it drops remarkably with modified conditions.

We list specific data of each level in Table 4. In the modified case, the numbers of employees are lower than original case especially the those of middle levels. Since there is no external recruiting in modified case, it is obvious that the full rate will decrease due to employees' leave. Although some qualified employees can be

Level of Position	Modified	Original
Senior manager/Executive	5.6	8.4
Junior manager/Executive	9.0	18.4
Experienced supervisor	7.4	23.0
Inexperienced supervisor	9.2	23.2
Experienced employee	72.6	107.4
Inexperienced employee	99.6	149.6
Administrative clerk	24.0	24.0

Table 4: Status of position

promoted into higher level, the high churn rate of the middle level and difficulty of satisfying the promotion conditions make the numbers of middle level employees relatively low. The situation given in that task 5 will cause unrecoverable damage to ICM's HR health. With the full rate of middle level employees lower than 50%, the HR structure is broken into fragments and the company won't be able to function normally.

5 Advice for HR

As the HR manager faces many problems such as identifying those that are likely to churn and maximizing employees' knowledge and abilities, we explain our further improved model.

5.1 Incentive Mechanism

A worker is more likely to churn if he or she was connected to other former employees who have churned. So there are two kinds of people who should be highly paid attention to and incented.

First, we should pay attention to those who have more friends than others. Because of their wide connection with other employees, the consequence of their churn may be destructive, which cause many friends' churn.

Second, we focus on those having more friends who have churned during the past year. This kind of people have a great probability to churn because of the influence of others.

We choose one sample in our simulations and 50 people in this sample to explain our incentive mechanism.

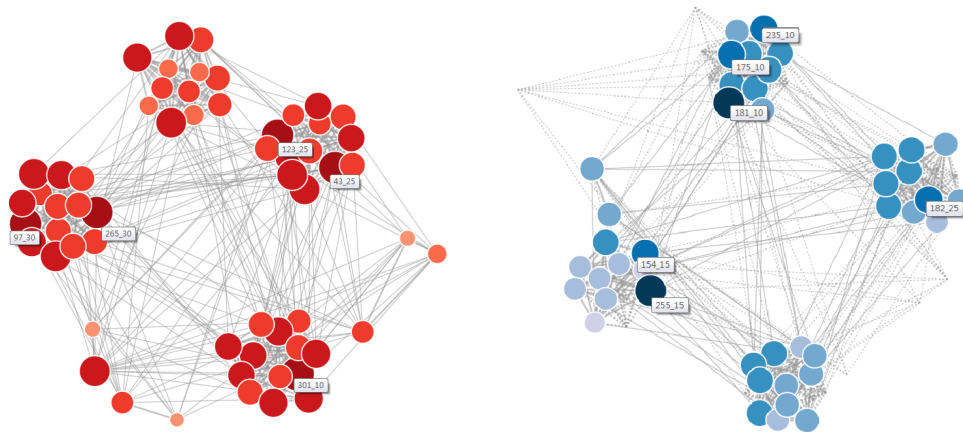


Figure 6: employees should be pay attention to

In Figure6, as we can see, each vertex represents an employee, each full line represents a relation between 2 employees, each imaginary line represent an relation between person who is on the job and person who churn. There are 5 employees in left graph we should pay attention to which are 43, 97, 123, 265 and 301 because they have more full lines than others. There are 6 employees in the right graph we should pay attention to which are 154, 175, 181, 182, 235 and 255 because they have more imaginary lines than others.

We provide incentives to these employees which will decrease their probabilities to churn.

At time t , we choose top 10% of the employees which have the properties above and decrease their churn rate by 50%. As shown in Figure7, the churn rate decrease rapidly.

Figure 7: Churn rate-Time Curve

5.2 Matching Employees to the Right Position

We consider different abilities of different employees. As we have an annual evaluation based on performance for each employee and each division or office has it's necessarily needed abilities, we can match employees with their most suitable position.

Because the evaluation is given annually, we reassign employees annually according to their abilities and the formula below:

$$m_{i,j} = (A_{ac_i} - B_{ac_j})^2 + (A_{ab_i} - B_{ab_j})^2 + (A_{at_i} - B_{at_j})^2 + (A_{po_i} - B_{po_j})^2 \quad (5)$$

where B_{ac_j} , B_{ab_j} , B_{at_j} , B_{po_j} represent the abilities needed for position j .

Defining $x_{ij} = \begin{cases} 1, & \text{reassign employee } i \text{ to position } j \\ 0, & \text{otherwise} \end{cases}$, we can describe this

problem as a mathematical programming problem:

$$\begin{aligned} \min & \sum_{i=1}^n \sum_{j=1}^m m_{ij} x_{ij} \\ \text{s.t.} & \sum_{j=1}^m x_{ij} = 1, i = 1, 2, \dots, n \\ & \sum_{i=1}^n x_{ij} \leq 1, j = 1, 2, \dots, m \\ & x_{ij} = 0 \text{ or } 1, i = 1, 2, \dots, n, j = 1, 2, \dots, m \end{aligned} \quad (6)$$

We can solve this problem via KuhnMunkres algorithm [Harold Kuhn and James Munkres 1957] whose time complexity is $O(n^4)$.

We reassign them once a year according to their evaluation. Figure8 shows productivity with reassignment and without reassignment. We can find that the productivity increase rapidly after reassign employees, which means reassignment truly help with productivity of the company.

6 Team Science

This part shows how our model connects to other organizational network layers such as information flow, trust, influence, and friendship. Number the layers from

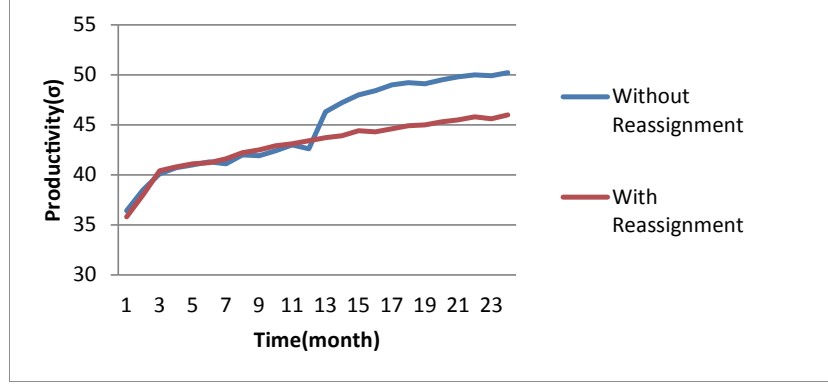


Figure 8: Productivity-Time Curve

0 to k .

$$G^\alpha = (V^\alpha, E^\alpha), \alpha = 1, 2, \dots, k$$

Each E^α is colored by a specific color. G^0 represents the network we built before, G^1, G^2, \dots, G^k represent the other layers added.

Still, in each network layer G_α , N_i represents employee i . The edge e_{ij} connects between two nodes N_i and N_j . Each edge has a value w_{ij}^α proving the weight of edge e_{ij}^α , positive correlated to the strength of connection between two employees in network layers α . Then we connect these network layers together to build a general network $G_e = (V, E, C)$, V is node set, C is color set, $E \subset V \times V \times C$ is edge set. In the general model, different layers effects each other. Take for example, if the friendship between employee i and employee j is deeper, with other conditions the same, their information flow is more fluent. w_{ij}^α is changing with time flows and G_e is a dynamic network.

However, human capital network make the maily effect on other network layers. In the general network, if an empolyee churns or is reassigned, the edge related to him will change. That is, w_{ij}^α get a new value for all α . Hence, human capital network takes the lead in this effort.

7 Sensitivity Analysis

a

8 Strengths and Weaknesses

Strengths

- Our model make fully use of the theory of multilayer networks so that it quantizes the relation accurately and reasonabllly.
- Our model exellently proves the interaction among these factors:leave probability, promotion probability and productivity.
- The network we built include both microcosmic part and macrocosmic part, and they react to each other.

- Our model proves the effect of time.

Weaknesses

- Limited by the time, we neglected sorts of factors which are not so significant. In fact, the model still has space to be perfected.
- The result has some randomness.

9 Conclusions

a

References

- [1] Huang Y. A Study of the Performance Assessment of State-owned Power Enterprises[D]. Tianjing University, 2010.
- [2] Journal T E D O T. The Investigation Report of the Condition of employee churn in 2011[J]. China New Time, 2012(2).
- [3] Zhang J. The Empirical Research of the Factors effecting employee churn in Resource-oriented Enterprise in Yuncheng, Shandong [D]. Southerneast University, 2014.
- [4] MEJNewman. The Structure and Function of Complex Networks[D]. University of Michigan, 2003.
- [5] Fang J, Zheng Z, Bi Q, et al. A Brand-new Interdisciplinary Science: Network Science (Part One)[J]. Progress in Physics, 2007, 27(3).
- [6] Fang J, Zheng Z, Bi Q, et al. A Brand-new Interdisciplinary Science: Network Science (Part Two)[J]. Progress in Physics, 2007, 27(4).