

The Management Ability for Head Nurses Assessed by Using Kendall's Coefficient of Concordance

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Abstract

Objectives: To assess the head nurses using Kendall's coefficient of Concordance.

Methods: An approach to assess the management ability of nurse heads was to collect data evaluated by nursing supervisors on 5-point ten management core axes in hospital. We analyze (1) the relation of Cronbach's α and Kendall's coefficient of concordance; (2) performance presentations on Google maps with visual displays; (3) the association between summation scores and the age of nurse heads.

Results: We found that (1) the relation between Cronbach's α and Kendall's coefficient is significantly

strong(r=0.61, t=5.87) in concordance(Kendall =0.85, χ^2 =100.38,df=59, p< .0001); (2) the dashboard shown on the Google map with multidisciplinary functionalities is merit; (3)the association between summation

scores and the age is trivial (r=0.18, t=1.5) without concordance(Kendall =0.59(χ^2 =76.33, df=65, p=0.16).

Conclusion: The demonstration of performance sheet is shown on Google maps with a dashboard. The Kendall's coefficient of concordance can be used for an example in assessing the selection of the nurse head in the hospital as well as in evaluating many kinds of competition judged by referees in the future.

Keywords: nurse head, Google maps, Kendall's coefficient, dashboard

1. BACKGROUND

A head nurse (as known as a chief nurse) in charge of nursing in a hospital is the head of the nursing staff and a registered nurse who supervises the care of all the patients, usually in a ward, or a health care facility [1]. Traditionally, head nurses wore a dark-blue dress that was usually darker than that of her subordinates in addition to a white-starched hat [2] and usually provides strong leadership and acts as a link between board-level nurses and clinical practice [3]. In the US military hospitals, head nurses are charged with the responsibility of making twicedaily rounds to supervise the nurses' duty performance [4].

Many large healthcare organizations such as medical center also have this kind of head nurses in healthcare service. They supervise a particular service and have insights into the facility or system such as surgical services, women's services, emergency services, critical care services, etc.

As of May 10 in 2018, 415 papers were found in Medline by searching keywords of (head nurse[Title]) OR chief nurse[Title] . However, only 18 exist with keywords of ((head nurse[Title]) OR chief nurse[Title]) and performance. Englebright and Perlin [5] explored characteristics of head nurse required for success in those increasingly responsible and visible roles. Melnyk and her colleagues [6] addressed that head nurses and hospital administrators need to invest in providing resources and an evidencebased culture so that clinicians can routinely implement evidence-based practice as the foundation of care to achieve higher healthcare quality and safety along with decreased costs. Babaeipour-Divshali et al. [7] suggested providing Head Nurse Empowerment Program

(HNEP) that can be used as a promoting tool in the nursing profession to increase head nurses' management skills and enhance job satisfaction among the staff nurses. However, none in those 18 papers regarding head nurse's performance has proposed an assessment system that can guide head nurses to improve their ability and skill in healthcare settings.

In hospital setting, we have observed many kinds of assessment or questionnaire to evaluate examinees' performance such as annual personal performance evaluation, nursing professional ability advanced evaluation, or such an Objective Structured Clinical Examination(OSCE)[8, 9], etc.. The most frequent type of assessment for head nurse is the annual personal performance evaluation.

The competence has frequently been discussed and applied to the field of human resources [10-12]. For instance, the competence can refer to nursing recruitment and the selection and promotion of personnel [13]. The competence can aid workforce allocation regarding staffing and development [14] and assist the selection of personnel for rewards and promotions based on performance evaluation criteria [15]. All of these factors require objective measurements in practice. We are thus interested in applying Kendall's coefficient of concordance (KCC for short in this study) [16] to objectively assess the performance of head nurses because KCC makes no assumptions regarding the nature of the probability distribution and can handle any number of distinct outcomes [17]. A novel way to visualize the results on the Google maps is worthily illustrated and shared with readers because Google maps have provided users to gain an overall geospatial visualization [18,19].

A study regarding the association between physician age and performance on the American Board of Emergency Medicine (ABEM) ConCert examination revealed that there was an association between advancing age and declining performance [20]. Another study investigated whether aging is associated with baseline crisis resource management (CRM) skills of acute care physicians (ACPs) and whether aging influences are learning from simulation-based education [21]. We are again interested in analyzing the relation between the performance and the age of the head nurses.

Our aims are to inspect (i)whether both KCC and Cronbach's α are associated, and consistent with each other on subscales of the annual personal performance evaluation for head nurses

in hospital; (ii) how to build a visual display on Google maps to show the performance of head nurses; (iii) what is the relation or the KCC effect between the total score and the age for the study head nurses.

2. METHODS

2.1. Data Sources

According to literature reviews and brain storms in a nursing department of a medical center in Taiwan, 37-item 5-point Likert-type instrument for ten subscales (see Additional file 1) were constructed and applied to the annual personal performance evaluation for head nurses at the end in 2017. The ten subscales were responded (i.e., 5: Strongly disagree,4: Disagree,3: Neutral,2: Disagree, 1: Strongly disagree) by the six nurse supervisors to 66 head nurses. Finally, the composite scores for each domain(axis) constructed a 66(person)*10(domain) composite dataset.

2.2. Data Analysis

2.2.1. Kendall's coefficient of concordance (KCC)

For calculating KCC, we firstly ranked the responses across items in each domain and assigned a serial number from 1 to the number of examinees. The higher response was assigned, the greater ranking score. The tie on rankings was endorsed with the mean (e.g., the lower three responses with an equal score are given the mean $(=2=(1+2_3)/3)$.

When persons are on rows and items in columns, the summation of the deviation from the mean squared (i.e., using the MS Excel function DEVSQ) was applied to the row summation. Kendall's W from 0 to 1 is yielded by the function $(=12*DEVSQ/(k^2*(n^3-n)))$, where k=item length and n=person size. The Chi-square is calculated by the equation $(=k^{*}(n-1)^{*}W$. The pvalue equals CHIDIST(χ^2 , df), where df= degree of freedom=n-1). We will reject the null hypothesis (i.e., W=0, inconsistency amongst items or raters) if the p-value is less than the significant level of 0.05. The study flowchart is shown in Figure 1. If SPSS statistical software was applied, the steps are proceeded as following: Analysis \rightarrow Non-parameter test \rightarrow History dialog records \rightarrow k correlation sample \rightarrow Kendall's W test checked, and the statistical results including W and χ^2 test will appear on the canvas.



Figure 1. The study flowchart

2.2.2. Cronbach's α

We computed all those 60 Cronbach's α for those ten subscales responded by the six nurse supervisors. Accordingly, the relation between Cronbach's α and Kendall's W can be drawn. Kendall's W on the two variables of Cronbach's α and Kendall's W can also be computed and tested.

2.3. The Summation Score and the Visual Dashboard Used for Displaying the Performance of Head Nurses

To avoid the difference in lenient and severe judgment among nurse supervisors, we adjusted the subscale scores by using the equation (=each response – the subscale's mean) on each supervisor's subscale. That is, the means of all the supervisor's subscales are equal (=0). The Kendall's W can be yielded by those 66(persons) *10(items) dataset accordingly.

Rasch model for continuous responses [22, 23] was used for estimating the person (i.e., head nurse) performance, calibrating the item difficulties, and computing Rasch fit statistics. If Infit mean square errors(MNSQ) for each item are less than 1.5, the unidimensionality scale is formed [23]. The variable map (also known as Wright map [24]) will be shown on the Google maps. The animated dashboard for the performance of head nurses can be also displayed on the Google maps, on which logit(=log odds) scores on the vertical(the higher, the better), and Z-score(=observed score – expected score) /standard deviation) for subscales on the horizontal(the higher, the more beyond the expected score). A rectangle box to circle the range within 95% confidence interval is plotted on the map. Each kind of domain performance for all head nurses will be popped up if the very icon was clicked. The names of the head nurse mean the one performing best in the domain.

2.4. The Relation Between the Summation Score and the Age for Examinees_o

For displaying the relation between the summation score and the age for head nurses, we calculated Kendall's W and tested the significance in statistics.

3. RESULTS

3.1. Concordance Analysis of Subscales for Each Nurse Supervisor

The age stratification is shown in Figure 2, indicating the special (e.g., ICU) wards have a higher age than those in ordinary wards. Most of the head nurses whose ages are ranged between 35 and 40. What the correlation coefficient between Cronbach's α and Kendall's W for the total subscales of the ten domains assess by nurse supervisors is 0.61(p<0.05; t=5.87= r* $\sqrt{(n-2)/1-r^2}$). Kendall's W for concordance between these two variables(Cronbach's α and Kendall's

W) of all subscales is $0.85(\chi^2 = 100.38, df = 59.00, p < 0.0001)$, indicating Cronbach's α and Kendall's has a high concordance and internal consistency.

However, among the subscales assessed by nurse supervisors, we found only the supervisor D whose subscales present the most number of Kendall's W in concordance. That is, the examinees' scores across items have high correlations, followed by the supervisor C. In contrast, the supervisor A has no any subscale displaying concordant, see Table 1. It is worth noting that the Authenticity subscale assessed by the supervisor A shows Cronbach's $\alpha = 0.94$. It is because only one with a very low score causes a very high variance on persons and results in a high Cronbach's α . In contrast, Kendall's W is 0.34 to reflect the more fact in concordance.

On the other hand, Cronbach's α might be negative which is seen in Table 1 on the two subscales(i.e., Professionality subscale of supervisor E and Influence subscale of supervisor A). The Innovation subscale of supervisor E whose Cronbach's α equals zero.



Figure2. The sample stratification for the age

	Prof.	Flex.	Coh.	Dial.	Comm.	Exe.	Insight	Influ.	Innov.	Authe.
Supervisor A										
Kendall	0.33	0.04	0.23	0.10	0.33	0.12	0.23	0.08	0.56	0.34
Chi-square	9.25	0.78	4.89	3.67	9.31	4.27	4.89	1.78	7.83	12.02
d.f.	7.00	7.00	7.00	7.00	7.00	7.00	7.00	7.00	7.00	7.00
p-value	0.24	1.00	0.67	0.82	0.23	0.75	0.67	0.97	0.35	0.10
Cronbach	0.59	0	0.24	0.35	0.73	0.3	0.42	-0.5	0.64	0.94
Supervisor B										
Kendall	0.27	0.63	0.38	0.36	0.49	0.06	0.16	0.60	0.36	0.21
Chi-square	12.06	20.91	12.65	19.97	21.46	3.25	5.23	19.74	7.85	11.38
d.f.	11.00	11.00	11.00	11.00	11.00	11.00	11.00	11.00	11.00	11.00
p-value	0.36	0.03*	0.32	0.05*	0.03*	0.99	0.92	0.05	0.73	0.41
Cronbach	0.58	0.77	0.43	0.79	0.8	-0.75	0.31	0.84	0.35	0.66
Supervisor C										
Kendall	0.41	0.60	0.40	0.59	0.64	0.19	0.62	0.54	0.48	0.21
Chi-square	23.14	25.25	16.79	41.50	35.61	13.38	25.88	22.50	13.36	14.85
d.f.	14.00	14.00	14.00	14.00	14.00	14.00	14.00	14.00	14.00	14.00
p-value	0.06	0.03*	0.27	0.00*	0.00*	0.50	0.03*	0.07	0.50	0.39
Cronbach	0.69	0.79	0.66	0.88	0.88	0.73	0.81	0.71	0.22	0.72
Supervisor D										
Kendall	0.40	0.53	0.64	0.76	0.57	0.44	0.63	0.66	0.73	0.60
Chi-square	16.06	15.95	19.32	37.81	22.83	22.13	18.76	19.91	14.64	30.21
d.f.	10.00	10.00	10.00	10.00	10.00	10.00	10.00	10.00	10.00	10.00
p-value	0.10	0.10*	0.04*	0.00*	0.01*	0.01*	0.04*	0.03*	0.15	0.00*
Cronbach	0.70	0.73	0.74	0.92	0.81	0.81	0.74	0.82	0.62	0.86
Supervisor E										
Kendall	0.08	0.57	0.30	0.39	0.56	0.07	0.24	0.52	0.29	0.35
Chi-square	1.57	8.62	4.57	9.69	11.14	1.80	3.57	7.86	2.86	8.66
d.f.	5.00	5.00	5.00	5.00	5.00	5.00	5.00	5.00	5.00	5.00
p-value	0.90	0.13	0.47	0.08	0.05*	0.88	0.61	0.16	0.72	0.12
Cronbach	-1.38	0.69	0.67	0.78	0.87	0.46	0.52	0.75	0	0.81
Supervisor F										
Kendall	0.27	0.63	0.36	0.29	0.38	0.22	0.15	0.22	0.15	0.11
Chi-square	13.84	24.66	14.16	18.91	19.78	14.19	5.68	8.60	4.00	7.29
d.f.	13.00	13.00	13.00	13.00	13.00	13.00	13.00	13.00	13.00	13.00
p-value	0.39	0.03*	0.36	0.13	0.10	0.36	0.96	0.80	0.99	0.89
Cronbach	0.62	0.80	0.54	0.75	0.70	0.71	0.36	0.45	0	0.70

Table1. The computation of Kendall's W for each subscales of the respective supervisor

3.2. The Performance was Shown on a Visual Dashboard

The variable map is shown with a visual dashboard [25] on the Google maps by the cloud computation technique [26], see Figure 3, indicating person logit score normally dispersed

on the left side and the most located near to -1 logits. The items are together located in a spot which is the reason that all items have a common total score because the ranking scores and the item difficulties (=0 logit) are all equal. The item Infit MNSQs are all less than 1,5, implying a unidimensional scale is constructed.



Figure3. The variable map showing on the Google maps

The visual dashboard is shown in Figure 4. Interested readers are recommended to click the link at the reference [27] to manipulate the

change for each domain in their way. The names on the map denote the one with the highest score on the domain, respectively.



Figure4. The performance sheet for head nurse on the Google maps

3.3. The Correlation Between Summation Score and the Age

The correlation between summation score and the age for head nurses is $0.18(t=1.5(=0.18 * \sqrt{(66-2)/(1-0.18^2)}))$, indicating the two variables have no any significant relation. That is to say that the performance of head nurses on those ten-axes competences will not be increased or decreased along with their ages. Kendall's W is $0.59(\chi^2 = 76.33)$. df=65, p=0.16), indicating the two variables as mentioned above are inconsistent.

4. DISCUSSION

The results show that (1) the relation between Cronbach's α and Kendall's coefficient is significantly strong(r=0.61, t=5.87) in concordance(Kendall =0.85, $\chi^2 = 100.38$, df=59, p < .0001); (2) the dashboard shown on the Google map with multidisciplinary functionalities is merit; (3) the association between summation scores and the age is trivial (r=0.18, t=1.5) without concordance (Kendall =0.59(χ^2 =76.33, df=65, p=0.16).

4.1. What this adds to what was Known and Changed

In tradition, the fairness of assessments are not allowed to doubt or to challenge the nurse supervisors. The unanimous responses endorsed by the judge will result in unfairness and bias occurrence. The judgment researches in fairness include two-way ANOVA or internal class coefficient(ICC) [28,29]. Rasch many facets used releasing much valuable information [30] is originated by item response theory which is hard to be applied and interpreted in practice [31].

In the current study, we applied Kendal's W to assess the competence of head nurses in a hospital which is similar to many annual employees' evaluation (i.e., supervisor assessing the subordinates) and different from other competitions or contests by multi-referees or those OSCE-like examinations. The features of this study are those (1) building the questionnaire illustrated in Additional file 1; (2) demonstrating the scoring and assessing method in Figure 1 and Table 1); (3) plotting visual representations in Figure 3 and Figure 4.

4.2. The Strength of this Study

Many journals [32-34] have used Google maps to show their research results. This study applied the layer design on Google maps to show the performance of head nurses on a dashboard which is a technique demonstration provided to hospital management for references in the future. After all, one picture is worth a thousand words. The animated dashboard is superior to the static image diagram.

Google maps provide an overall view of geospatial visualization with coordinates of latitude and longitude on a map [18,19]. However, none applies the Google maps to build report performance sheet as we did, not mention to create the animated dashboard. We are interested in this kind of performance report card online for the current study. Using the application programming interface (API) [35] technique with the Google maps to build the dashboard is the strength of the study, see the references [25-27].

Researchers [36] addressed that training is so limited to help judges assess examinees. Even all judges perform excellently which cannot guarantee to avoid the unfairness occurrence [37,38]. The Kendall's coefficient of concordance is a way to assess the judge bias in occurrence.

The computation of Kendall's W is somewhat difficult, especially endorsing the mean to the tie in ranking. Even SPSS statistical software provides the tool to obtain the Kendall's W, many subscales all-in-one to gain the results like Table 1 is required to program a computer routine. We provide an mp4 video at the reference [39] or scan the QR-code in Figure 1 to introduce how to execute the routine in the calculation of the Kendall's W.

4.3. The Limitations and the Future Study

The forced ranking is suitable for using Kendall's W, but cannot discriminate the attributes of subordinates under each supervisor. For instance, the supervisor A assessed her head nurses with low variance and led a low to Kendall's W or Cronbach's α , which might be the performance equal among head nurses under the supervisor A. We cannot attribute the low Kendall's W to the supervisor's bias or misbehavior. As such, the adjustment for a score to be equal would lead the result to another unfair problem. This is the first limitation.

Only one judge assessing her subordinates is simple but subjective enough to produce personal bias which is unfair in annual employee evaluation. This is the second limitation. Future studies are suggested to adopt the 360-degree evaluation [40] or OSCE-like multi-judge assessment to let the evaluation more objective.

The sample came from one hospital. The results like no relation to the summation score and the age cannot generalize to other hospitals or institutes which is another limitation. Hopefully, more studies may be examined further to generalize the result in the future.

5. CONCLUSION

The demonstration of performance sheet is shown on Google maps with a dashboard. The Kendall's coefficient of concordance can be used for an example in assessing the selection of the nurse head in the hospital as well as in evaluating many kinds of competition judged by referees in the future.

Axis	Content	Item(5: Strongly disagree,4: Disagree,3: Neutral,2: Disagree, 1: Strongly disagree)				
1. Professionality	Knowledge	1 Her care expertise is eligible.				
	Skill	2 Her assistance in work rule or flow ruling and revising is eligible.				
		3 Her problem solving is eligible.				
	Academia	4 Her instruction in writing the special assignments like PICO, case report, and care project are eligible.				
2. Flexibility	Resistance	5 Her positivity in problems is eligible.				
	Resilience	6 Her resilience in frustration is eligible				
	Adaptation	7 Her adaptation in stress is eligible.				
3. Cohesiveness	Collaboration	8 Her collaboration is eligible.				
	Motivation	9 Her motivation is eligible.				
	Coaching	10 Her coaching is eligible.				
4. Dialectic	Logic	11. Her active learning is eligible.				
	Learning	12. Her answer acquiring is eligible.				
	Seeking in answer	13. Her complete insightfulness is eligible.				
	Insightfulness	14. Her criticalness is eligible.				
	Thinking	15. Her active learning is eligible.				
5. Communication	Organization	16.Her organization is eligible.				
	Convincement	17.Her convincement is eligible.				
	Priming	18. Her priming is eligible.				
	Compromise	19.Her compromise is eligible.				
6. Execution	Model	20. Her model is eligible.21.				
	Honesty	21. Her honesty and work ethic are eligible.				
	Equity	22. Her equity in affairs is eligible.				
	Fulfillment	23. Her execution and propelling are eligible.				
	Achievement	24. Her effective achievement before the deadline is eligible.				
7. Insight	Sensitiveness	25.Her sensitiveness in problem sensing is eligible.				
	Self-awareness	26. Her reflection is eligible.				
	Crisis avoiding and managing	27. Her crisis avoiding and managing are eligible.				
8. Influence	Belief	28.Her firm belief is eligible.				
	Model	29. Her model is eligible.				
	Charisma	30. Her charisma in leading is eligible.				
9. Innovation	Reform	31. Her acceptance in reforming is eligible.				
	Innovativeness	32. Her innovativeness in ideas is eligible.				
10. Authenticity	Liberalness	33.Her liberalness is eligible.				
	Empathy	34. Her empathy is eligible.				
	Caring	35. Her active caring is eligible.				
	Trustworthiness	36. Her trustworthiness is eligible.				
	Sincerity	37. Her sincerity is eligible.				

Additional File: Ten Axes of Competence for Head Nurses

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