Patient-perceived hospital service quality: an empirical assessment

Author 1: Yogesh Pai P, B.E., M.B.A., M.Phil., Ph.D., Associate Professor, School of Management, Manipal, India, Telephone: +91-9880530633, e-mail: yogesh.pai@manipal.edu, yogeshpai.p@gmail.com

Author 2: Satyanarayana T Chary, M. Com, M.F.M., M.B.A., Ph.D., Professor, Department of Business Management, Telangana University, Nizamabad, India, Telephone: +91-9704599931, e-mail: tsnchary@gmail.com

Author 3: Rashmi Yogesh Pai, B.Sc., M.B.A., Ph.D., Assistant Professor, Poornaprajna Institute of Management, Udupi, India, Telephone: +91-9880530632, e-mail: rashyogipai@gmail.com

Corresponding author: Yogesh Pai P
Corresponding Author’s E-mail: yogesh.pai@manipal.edu, yogeshpai.p@gmail.com

Structured Abstract:

Purpose: Our purpose was to appraise Pai and Chary’s (2016) conceptual framework for measuring patient-perceived hospital service quality.

Design/Methodology/Approach: A structured questionnaire was used to obtain data from teaching, public and corporate hospital patients. Several tests were conducted to assess the instrument’s reliability and validity. Pai and Chary’s (2016) nine dimensions for measuring hospital service quality (HSQ) were examined.

Findings: Tests confirm that Pai and Chary’s (2016) conceptual framework is reliable and valid. Our study also establishes that nine dimensions measure hospital service quality (HSQ).

Practical implications: The framework empowers managers to assess service quality in any hospital setting: corporate, public and teaching, using an approach that is superior to existing hospital service quality scales.

Originality/Value: This article helps researchers and practitioners to assess hospital service quality from patient perspectives in any hospital setting.

Keywords: Reliability; Scale assessment; Hospital service quality; Validity; India.

Article Classification: Research

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Introduction

Healthcare is considered a ‘credence’ good – an offering that consumers will never be able to evaluate owing to deficient medical knowledge (Bloom and Reeve, 1990). Additionally, conceptualising and measuring service quality in a healthcare setting is more important and simultaneously more complex (Taner and Antony, 2006). However, researchers need to come up with ways to measure healthcare service quality, because unless we measure we cannot manage and improve healthcare services (Lohr, 2015). The literature indicates that there is a variability and confusion in how quality is conceptualized and operationalized (Sower et al., 2001). Many researchers attempt to define and conceptualize hospital service quality. Pai and Chary (2013),
for example, show that the Parasuraman et al., (1985, 1988) SERVQUAL/modified SERVQUAL questionnaire is conventionally practiced in healthcare. They also pointed out studies where SERVQUAL items have not loaded onto respective dimensions highlighting a five-component structure that is lacking, suggesting a new questionnaire needs developing. New instruments, designed for healthcare setting: PRIVHEALTHQUAL (Ramsaran-Fowdar, 2008) for private and PubHosQual (Aagja and Garg, 2010) for public settings, have emerged. Studies adopting these instruments are scarce because these scales are hospital specific and not a general scale that measures hospital service quality in any hospital context. An instrument that could measure hospital service quality (HSQ) in any hospital settings has gained importance. Pai and Chary (2016) proposed their conceptual framework for measuring hospital service quality using nine dimensions. Their framework, unlike other instruments such as SERVQUAL (Parasuraman et al., 1988), PRIVHEALTHQUAL (Ramsaran-Fowdar, 2008) and PubHosQual (Aagja and Garg, 2010) has an additional dimension (relationship) in line with researchers such as Carman (1990) and Reynoso and Moore (1995), who suggested adding service specific dimensions to SERVQUAL, there exists no empirical studies to support or refute their conceptual framework. As there are no research studies that used Pai and Chary’s framework (2016), our purpose is to empirically appraise this framework and addresses literature gaps.

Objectives
We conducted our study with the objective to test the Pai and Chary (2016) conceptual framework’s validity and reliability. We also aim to assess HSQ through nine dimensions: (i) healthscape; (ii) personnel; (iii) hospital image; (iv) trustworthiness; (v) clinical process; (vi) communication; (vii) relationship; (viii) personalization and (ix) administrative procedures.

Structured questionnaire
The questionnaire included:

1. Respondent demographics – gender, age, occupation, educational level, marital status and income.
2. Sixty-six items covering nine dimensions (Pai and Chary, 2016).
3. Four items measuring respondents’ hospital service quality perception.
4. Twelve items that measured respondent satisfaction and behaviour.

A Likert (1932) scale, having a balanced rating with equal categories above and below the midpoint, was used with each item; anchored with verbal statements like ‘strongly disagree’, ‘disagree’, ‘neither agree or disagree’, ‘agree’, ‘strongly agree’. All items were phrased positively (Easterby-Smith et al., 2002; Pai and Chary, 2013; Parasuraman et al., 1994). Sixty-six hospital service quality items were categorized into nine dimensions: (i) healthscape (15 items); (ii) personnel (11); (iii) hospital image (5); (iv) trustworthiness (5); (v) clinical process (6); (vi) communication (9); (vii) relationship (3); (viii) personalization (3); and (ix) administrative procedures (9 items). Four perceived service quality, twelve satisfaction and behavioural intention items were included (Appendix 1).

Piloting
Piloting a new instrument is imperative (DeVellis, 1991; Kirchhoff, 1999). Before the pilot test is performed, it is advisable to obtain the original evaluation (DeVellis, 1991). Andres (2012) recommends that conducting a pilot study by involving colleagues, friends and family members.
who assume an audience role. Our questionnaire was subjected to the pre-testing process by circulating it among ten colleagues, friends and family members. In any research, respondents often misunderstand words or concepts. Although communication difficulties exist, respondents still provide legitimate answers to survey questions (Clark and Schober, 1992; Tourangeau et al., 2000). To overcome these problems, Collins (2003, p.230) suggested cognitively testing survey questions that helps ‘to identify how and where the question fails to achieve its measurement purpose’. Although there are various cognitive methods that have been developed and applied to instrument testing, such as cognitive interviewing, paraphrasing, card sorts, vignettes, confidence ratings and response latency timing (Czaja, 1998; Forsyth and Lessler, 1991; Jobe and Mingay, 1991); cognitive interviewing is becoming widespread (Schwarz, 1997). Cognitive interviewing (involving two methods: think aloud interviewing and probing) and paraphrasing were used in testing. On completion, the instrument was tried in a hospital with patients as respondents. According to Lackey and Wingate (1998), pilot testing a newly developed instrument should be undertaken with respondents selected from the same population from which the subjects in the major study will be selected. Consequently, the questionnaire was tested with respondents from teaching, corporate and public hospitals. Pre-testing ensured correct phrasing, format, length and question sequence. Pretesting was done with 30 respondents using Hertzog’s (2008) guidance. The questionnaire was corrected after feedback.

Data collection
Data were collected in three hospitals – teaching, corporate and public in Karnataka. India has 29 states and seven union territories (Ashok, 2014), and every state has a specific language. Although India is an English-speaking country, Kannada is Karnataka’s local language. Hence, the questionnaires were in Kannada and translated into the Malayalam language as few hospitals had Malayalam speaking patients in significant numbers owing to hospital’s proximity to another state (Kerala), where the official language is Malayalam. Consequently, questionnaires were administered in three languages: English; Kannada and Malayalam. There are several studies in which the questionnaire was administered in different languages specific to the country studied; e.g., English and Maltese (Camilleri and O’Callaghan, 1998); English and Arabic (Jabnoun and Chakar, 2003); English and Turkish (Kara et al., 2005); English and Gujrati (Aagja and Garg, 2010); and English and Bengali (Akter et al., 2008). The Kannada/Malayalam versions were created through careful translation and back-translation techniques (Candell and Hulin, 1987; McGorr, 2000). First, the questionnaire was translated into Kannada/Malayalam and then the Kannada/Malayalam items were back-translated into English by a bilingual expert to ensure that the original content was kept intact. In translating the scale items into Kannada/Malayalam, Malinowski’s (1935) four-step translation technique was implemented:

1. An interlinear, or word-by-word, translation.
2. A ‘free’ translation in which clarifying terms, conjunctions, etc., are added and words reinterpreted.
3. Analysing and collating two translations leading to:
4. A contextual specification of meaning.

An attempt was made to remove discrepancies between English and Kannada/Malayalam questionnaire versions. No individual items were found to be problematic in translation. Previous studies reported backward translation; e.g., English to Bengali (Akter et al., 2008; Andaleeb,
Accordingly, our instrument was translated from English to Kannada and Malayalam languages.

Sample
The study population was defined as all patients 18 years or older with a stable mental and clinical condition during the data collection procedure. A stratified random sampling was adopted for the study. As the name implies, there is a stratification or segregation process followed by randomly selecting subjects from each stratum. The population is first divided into mutually exclusive groups which are relevant, appropriate and meaningful in the study context. The teaching, corporate and public hospitals constitute three strata. Hospitals were selected using systematic sampling and proportionate stratified sampling. The sampling frame was prepared for three strata and ten per cent from each stratum: three hospitals each from teaching and public hospitals and four corporate hospitals, totalling ten hospitals. Respondent from ten hospitals were chosen randomly (every fifth element in the population), both inpatients and outpatients yielding 602 respondents. The respondent’s demographic profile is depicted in the Table I, comprising males (44%). Their sample’s age ranged from 18 to 67+ years.

Table I here

Reliability and validity
Instruments designed to measure a specific concept should measure what is set out to measure. Reliability and validity are the two main criteria for measuring how good measures are in any research instrument. Reliability indicates stability and consistency with which the instrument measures the concept, thus assessing a measure’s goodness (Sekaran and Bougie, 2010). It is important to calculate scale reliability, which refers to the extent to which a scale can reproduce the same results in repeated trials (Hair et al., 2003). Consistency can be examined through the inter-item consistency reliability and split-half-reliability tests (Sekaran and Bougie, 2010). The internal consistency reliability test is acceptable when the reliability coefficient exceeds Nunnally’s (1978) 0.7 reliability criterion.

The questionnaire included 82 items (Appendix 1). For internal consistency, the most popular test is coefficient alpha (Cronbach). For Pai and Chary’s (2016) 66 items, the value is 0.965 and 0.972 for the 82-item instrument. The higher the coefficients, the better the measuring instrument (Sekaran and Bougie, 2010). Another technique for testing internal consistency is the split-half technique, where items constituting the instrument are divided into two halves and the resulting half scores are correlated (Malhotra and Dash, 2012). For the 66-item version, Cronbach’s Alpha was 0.93 and 0.947, correlation between forms was 0.793, Spearman–Brown Coefficient for equal length and unequal length is 0.884, Guttman Split-Half Coefficient is 0.878. For the 82-item questionnaire, Cronbach’s Alpha was 0.944 and 0.956, correlation between forms was 0.821, Spearman–Brown Coefficient for equal length and unequal length is 0.902, Guttman Split-Half Coefficient is 0.897. We can conclude, therefore, that the instrument has adequate reliability. Cronbach’s alpha is a widely-used reliability coefficient that assesses the entire scale’s internal consistency and averages all possible split-half coefficients resulting from different ways of splitting the scale items ranging from 0 to 1. According to Malhotra and Dash (2012), it is appropriate to use internal consistency reliability for each dimension, if several items are used to measure each dimension. In our study, patient-perceived hospital service quality is
measured using different dimensions, each is measured by several items, hence we used Cronbach’s alpha to measure internal consistency. Cronbach’s alpha for the twelve constructs considered in the study was measured and values for each construct are 0.70 and above, indicating a strong reliability (Table II).

**Table II here**

Although reliability is a necessary contributor to validity, it is not a sufficient condition for validity (Sekaran and Bougie, 2010). Therefore, validity becomes equally important. Validity is ‘the extent to which a rating scale truly reflects the underlying variable that it attempts to measure’ (Parasuraman et al., 2004, p. 294). Content validity, criterion-related validity and construct validity are generally discussed in research articles.

**Content validity**

Content validity is the degree to which items adequately represents all relevant items under study (Cooper et al., 2012), determined using judgment and panel evaluation. For an attitudinal scale, content validity is an overall criterion that can be assessed only though a researcher’s subjective judgment (Parasuraman et al., 2004). We exercised judgment through carefully defining and analysing conceptual and empirical frameworks through an extensive literature review. Judges can attest to content validity (Sekaran and Bougie, 2010), so the questionnaire was subjected to expert review by practitioners and academics. Content/face validity was confirmed; i.e., the proposed measurement instrument measures the intended construct. The items for the current study were chosen from the literature (Sekaran and Bougie, 2010). Scales were refined using a pilot study. Patients revealed they had no difficulty understanding the questionnaire items indicating and confirming face validity (Arasli et al., 2008). All steps ensured that the instrument possesses face validity. Although implementing a scale based on only face validity claims has a far greater impact than item’s random deletion through mere formal scale refinement methods (Finn and Kayande, 2004), nevertheless, other validity tests are also conducted.

Criterion related validity (Ping, 2004) emphasizes that, for new measures, criterion validity should be assessed, which is established when the measure differentiates individuals on a criterion it is expected to predict. Criterion related validity can be established by concurrent or predictive validity (Sekaran and Bougie, 2010). Predictive validity indicates the instrument’s ability to differentiate among individuals with reference to a future criterion (Sekaran and Bougie, 2010). The researcher needs to ensure that the validity criterion used is valid and the intended measure is judged on four qualities: relevance; freedom from bias; reliability; and availability (Cooper et al., 2012). Criterion-related validity was established using predictive validity, adopting Jabnoun and Chakar’s (2003) technique, who correlated their service quality dimensions with overall service quality. Individual item scores on nine dimensions (listed earlier) were summed to obtain overall scores for each respondent. Scores were then correlated with a summated perceived service quality, satisfaction and behavioural intentions scale. All items were captured and measured using Likert scales with a five-point response format. A higher score indicated a more favourable response (Table III). All correlations were positive and statistically significant at 0.001 level, which establishes predictive validity.

**Table III here**
Concurrent validity was assessed by examining the association between perceived service quality and a related variable, patient satisfaction (Hansen et al., 2008). Patient satisfaction is a distant but related construct to perceived service quality (Choi et al., 2004; Rao et al., 2006). Although there are differences between patient satisfaction and perceived service quality, they are related and a moderate correlation is expected (Al-Qatari and Haran, 1999). Crompton and Love (1995) stressed that the two constructs are likely to be positively correlated, but unlikely to be linear. The correlation between perceived service quality and patient satisfaction is 0.633 (Table III), which indicates concurrent validity. Concurrent validity assessment is conducted in line with Duggirala et al., (2008) such that in the present context, perceived service quality (PSQ) is chosen as the nine-dimension’s outcome and a bivariate correlation analysis was carried out among all constructs, which have significant positive correlations with the criterion chosen, namely, PSQ (range 0.414 to 0.587, p < 0.001). According to Sekaran and Bougie (2010), construct validity tests how well the results obtained from a measure fit the theories around which the test is designed and assessed through convergent and discriminant validity. Convergent validity is the degree to which scores on one scale correlate with scores on other scales designed to assess the same construct (Cooper et al., 2012). Convergent validity is also assessed by calculating one-way analysis of variance (ANOVA) (Aagia and Garg, 2010; Bahia and Nantel, 2000; Parasuraman et al., 1988; Wong et al., 2001). The association between healthscape, personnel, hospital image, trustworthiness, clinical care process, communication, relationship, personalization and administrative procedures and PSQ was significant at less than 1% (Table IV), which means that the group differences are significant; i.e., good convergent validity.

Table IV here

Alternate methods for establishing criterion and construct validity
Campbell and Fiske’s (1959) procedure has been used as an alternate method for establishing criterion and construct validity, adopted in similar studies such as Boshoff and Terblanche, 1997; Kaul 2007. Convergent validity (Table V) is confirmed as there is a high correlation between scale items and PSQ (correlation coefficient, r = 0.665, p<0.001). To assess predictive validity, respondents were asked whether they intended to return to the same hospital (repurchase intentions). Results confirm predictive validity (r = 0.621, p< 0.001).

Table V here

Empirically validating nine patient-perceived hospital service quality dimensions
We propose the following hypothesis:

H₀: The heathscape, personnel, hospital image, trustworthiness, clinical process, communication, relationship, personalization and administrative procedure domains do not predict hospital service quality.

To test our hypothesis, a multiple regression approach is adopted with the nine dimensions as independent variables and PSQ as the dependent variable. Variables were examined for multicollinearity and autocorrelation before the regression analysis was run. Multicollinearity exists when there is a strong correlation between two or more predictors in a regression model.
leading to a regression model that fits the data, but no variable had a significant impact in predicting the dependent variable. In SPSS, there are collinearity diagnostics that are variance inflation factor (VIF) and tolerance statistics. The VIF indicates whether the predictor has a strong linear relationship with other predictor(s), while the tolerance statistic reciprocates the variance inflation factor (VIF). We used the Durbin-Watson test to identify any serial correlations between errors (Table VI and VII).

**Table VI here**

The tolerance values need to be greater than 0.1 and the VIF values should be less than 10 (Ho, 2006). From the tolerance and VIF values shown in Table VI, multicollinearity does not exist in the predictor values. Hence, the nine independent variables influence on dependent variables is analysed. The regression model is formed using the statistical (stepwise) regression method taking all nine independent variables together and examining their combined effect on the dependent variable. In the statistical (stepwise) regression model, the predictor variables’ order is solely based on statistical criteria and variables that correlate most strongly with the dependent variable will be given entry priority. Our regression models require no references to theoretical considerations, and are primarily used when the researcher is unsure about the independent variable’s predictive power (Ho, 2006). This method is apt for the present study and therefore stepwise regression is adopted, which includes forward inclusion, backward elimination and stepwise solution (Ho, 2006; Malhotra and Dash, 2012). To identify an optimal regression equation, one needs to compute combinational solutions in which all possible combinations are examined (Malhotra and Dash, 2012). The optimal regression equation in our study is computed using all possible combinations. However, the stepwise regression method is preferable to the forward method, as they reduce Type II errors (i.e., missing a predictor that does in fact predict the outcome) (Field, 2013). Regression analysis results are shown in Table VII.

**Table VII here**

The F value tests how well the regression model fits the data (Ho, 2006). In case the probability associated with the F statistics is small, the hypothesis that R-squared is equal to zero is rejected. In our case, F is 71.65, p < 0.001 (Table VII); i.e., the hypothesis that there is no linear relationship between the predictor and the dependent variables is rejected. The R statistic indicates that the correlation coefficient as 0.677 and the R-squared value is 0.458. For our sample, the predictor variables explained 45.8 per cent of the variance in the dependent variable. Hence, the null hypothesis H₀ is rejected. In Table VII, Durbin-Watson’s d = 1.987. Durbin-Watson can vary between 0 and 4; i.e., 2 implies that the residuals are uncorrelated. If the value is greater than 2, then it indicates a negative correlation between adjacent residuals; whereas a value below 2 indicates a positive correlation (Field, 2013). Thus, Durbin-Watson statistic tests serial correlations between errors, a metric that assesses another linear model assumption. Field (2013) suggests that the stepwise method should be avoided except for exploratory model building. He recommends cross-validating the model if the stepwise method is used, suggesting two cross-validation methods: adjusted R² and data splitting.

*Cross-validating the model using adjusted R square (R²)*
The adjusted $R^2$ shows how well our model generalizes. Although the adjusted $R^2$ should be the same or very close to $R^2$, it differs generally. This difference is known as shrinkage, which equals $R^2$ - Adjusted $R^2$. As per model summary (Table VII) shrinkage is equal to $0.458 - 0.451 = 0.007$ (about 0.7 per cent); i.e., the model’s shrinkage is small. Although there are no shrinkage guidelines in the literature, it is valuable cross-validation such that one can be confident in the equation’s generalizability if they are similar (Pedhazur, 1997). According to Field (2013), shrinkage means that if the model were derived from the population rather than a sample, then it would account for approximately 0.7 per cent less variance in the outcome. Stein’s formula helps us to calculate adjusted $R^2$ and tells us how well the model cross-validates (Stevens, 2002), which is computed as: 

$$\text{Adjusted } R^2 = 1 - \frac{(n-1/n-k-1) (n-2/n-k-2) (n+1/n)}{(1 - R^2)}$$

where $R^2$ is the unadjusted value, $n$ is total participants and $k$ is total predictors in the model. Substituting $R^2$ as $0.458$ (Table VII), $n = 602$ and $k=9$ in the above formula, we get $0.441$, which is computed as $1 - \frac{(602-1/602-9-1) (602-2/602-9-2) (602+1/602)}{(1 - 0.458)}$, which equals $0.441$. This value is like the observed $0.458$, indicating that cross-validity is acceptable (Field, 2013, p.312).

Cross-validating the model using data splitting

The other cross-validation method is the data splitting approach. Cross-validation is a validity test that examines whether the regression model continues to hold on comparable data not used in the original estimation (Malhotra and Dash, 2012). In a cross-validation procedure, we estimate the regression model for the entire data, which are split into two parts: estimation and validation samples. The estimation sample incudes 50-90% of the total sample. The regression model is estimated using the estimation sample only. This model is compared with the entire sample. The estimated model is applied to the data in the validation sample to predict the dependent variable. In case there is a large discrepancy between $R^2$ for the smaller and larger samples, this indicate over fitting and weak generalizability (Tabachnick and Fidell, 2014). In other words, if there is a significant difference between $R^2$ for the larger sample (estimation sample) and smaller sample (validation sample) then there is weak generalizability. Using SPSS, the step-wise regression on a random selection is adopted as suggested by Field (2013). Cases are randomly divided into two parts for cross-validation. Creating a variable that contains information about the estimation sample and validation samples using 0 and 1 to indicate the estimation sample and validation sample. We split the data into two samples: estimation and validation samples with 325 (53.9%) and 277 (46.1%) respondents/samples using Malhotra and Dash, (2012) guidance that estimation sample includes 50-90% of the total sample. Stepwise regression is conducted on the split data; i.e., estimation and validation sample, and the $R^2$ values are compared with the original model (Malhotra and Dash, 2012; Tabachnick and Fidell, 2014). The overall relationship between the dependent variable (PSQ) and the independent variable (nine dimensions) must be statistically significant for both validation analyses. Table VIII and IX indicate that significance is achieved. The $R^2$ for each validation must be within $\pm 5\%$ (plus or minus 5%) of the model’s $R^2$ using the full sample. The entire sample’s $R^2$ is $45.8\%$; thus, $R^2$ and validation models can vary by $5\%$ ($40.8\%$ - $50.8\%$). Table X shows the split sample $R^2$ as $48.4\%$ and $44.1\%$, whereas the full sample is $45.8\%$. The split sample validation’s value is confirmed; i.e., cross-validation is verified and supports our finding’s generalizability.

Table VIII, IX and X here

Discussion
Appraising reliability and validity tests

Internal consistency is the degree to which various multidimensional construct measures correlate with the scale (Hair et al., 2003). They indicate the item’s homogeneity in the measure that captures the construct after administering the instrument. All approaches adopted for testing reliability and validity are as shown in Table XI.

**Table XI here**

Adopting Cronbach’s alpha and split half technique show that the scale demonstrates reliability. The instrument was piloted and subjected to expert evaluation; i.e., demonstrating content validity. Concurrent and predictive validity have demonstrated criterion related validity. Discriminant validity was not assessed, convergent validity demonstrated construct validity. Aagja and Garg (2010); Kaul (2007) did not successfully demonstrate discriminant validity. Although there are different studies that measured healthcare service quality, only Duggirala et al., (2008) tested and supported all three: content, criterion and construct validity (Table XII). Pai and Chary (2013) observed that hospital service quality studies have neither evaluated all three validity methods nor discussed them. All three validity methods applied to Pai and Chary’s (2016) framework demonstrate instrument validity. The instrument, therefore, is deemed both reliable and valid for studying hospital service quality.

**Table XII here**

Pai and Chary’s (2016) conceptual framework, comprising nine dimensions, measures hospital service quality using multiple regression. Results indicate that the null hypothesis can be rejected. According to Salkind (2000), an R (correlation coefficient) between 0.4 to 0.6 is moderate; 0.6 to 0.8 is strong and 0.8 to 1 is a strong to perfect association. The R in our study is 0.677 (Table VII) indicating a strong association between dependent and independent variables (p < 0.001). The model is also cross validated using adjusted R² and data splitting (Field, 2013).

**Conclusion**

Our objective was twofold, first to evaluate the Pai and Chary (2016) conceptual framework’s reliability and validity and second to evaluate their proposed nine-dimension structure for measuring patient-perceived hospital service quality. Our study provides empirical evidence that the scale developed to measure hospital service quality is reliable and valid. We show that the nine dimensions predict hospital service quality.

**Theoretical contribution and managerial implications**

The study has strong implications for hospital quality managers seeking to measure hospital service quality who can implement the structured questionnaire in any hospital setting. Managers can also use the instrument to compare service quality among different combinations such as teaching, public and corporate hospitals.

**Scope for future studies**

Our study is cross sectional and limited to Karnataka state hospitals. Future studies should use a longitudinal study to provide greater diagnostic value. The instrument’s validity and reliability can also be tested in other developing countries.
References


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### Table I: Demographic profile

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<th>Demographic Factors</th>
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<td></td>
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### Table II: Cronbach alpha values

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<th>Constructs</th>
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Healthscape 0.875  
Personnel 0.869  
Hospital Image 0.739  
Trustworthiness 0.805  
Clinical Care Process 0.823  
Communication 0.869  
Relationship 0.845  
Personalization 0.730  
Administrative Procedures 0.896  
Perceived Service Quality 0.873  
Customer Satisfaction 0.811  
Behavioural Intentions 0.882

**Table III:** Bivariate correlations among the dimensions

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All are positive and significant at 0.001.

**Table IV:** One-way ANOVA for assessing convergent validity – independent variable service quality

<table>
<thead>
<tr>
<th>Descriptive Variables</th>
<th>Degrees of freedom</th>
<th>F</th>
<th>Sig.</th>
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### Table V: Validity tests

<table>
<thead>
<tr>
<th>Validity</th>
<th>Measures used</th>
<th>Correlation coefficient (r)</th>
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<tr>
<td>Convergent validity</td>
<td>Perceived service quality</td>
<td>0.665 (p&lt;0.001)</td>
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<tr>
<td>Predictive validity</td>
<td>Repurchase intentions</td>
<td>0.621 (p&lt;0.001)</td>
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### Table VI: Collinearity statistics

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<th>Independent Variable</th>
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<th>VIF*</th>
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<td>Personnel (PL)</td>
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<td>Hospital Image (HI)</td>
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<td>Trustworthiness (TW)</td>
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<tr>
<td>Clinical Care Process (CCP)</td>
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<td>Communication (COM)</td>
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<tr>
<td>Relationship (REL)</td>
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<td>Personalization (PER)</td>
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<td>Administrative Procedures (AP)</td>
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</table>

*assessed using Enter regression method
Table VII: Model summary

<table>
<thead>
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<th></th>
<th>R</th>
<th>R Square</th>
<th>Adjusted R Square</th>
<th>F</th>
<th>Sig.</th>
<th>Durbin-Watson</th>
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<tr>
<td></td>
<td>.677</td>
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<td>.451</td>
<td>71.65</td>
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Table VIII: Split data model summary - estimation sample – 0

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<th>Adjusted R Square</th>
<th>F</th>
<th>Sig.</th>
<th>Durbin-Watson</th>
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<td>split ~= 0.00 (Unselected)</td>
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<td>2.053</td>
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Table IX: Split data model summary - validation sample – 1

<table>
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<th>Adjusted R Square</th>
<th>F</th>
<th>Sig.</th>
<th>Durbin-Watson</th>
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</table>

Table X: Summary - Validation results: stepwise regression

<table>
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<tr>
<th></th>
<th>Entire data</th>
<th>Estimation sample</th>
<th>Validation sample</th>
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<td>ANOVA significance</td>
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<td>(sig&lt;=0.05)</td>
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<tr>
<td>R²</td>
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Table XI: Reliability and validity tests

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<th>Construct Validity</th>
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<td>split-half reliability tests</td>
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<td>Literature</td>
<td>Predictive validity (2 tests)</td>
<td>Convergent validity (2 tests)</td>
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<td>inter-item consistency reliability</td>
<td>inter-item consistency reliability</td>
<td>Pilot testing</td>
<td>Concurrent validity (2 tests)</td>
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</tr>
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</table>

Table XII: Reliability and validity a comparison with established scales

|---------|-------------|-----------------------------|-----------------------|--------------------------|----------------------|
Appendix 1: Questionnaire

1. Modern and up-to-date equipment (e.g., Computerized tomography (CT) and Magnetic Resonance Imaging (MRI), patient information and billing) to serve patients more effectively.
2. Physical facilities are visually appealing.
3. Facility adequacy (e.g., wards, beds, operation theatre, Intensive Care Unit (ICU)).
4. Cleanliness (toilets, rooms and wards).
6. Hospital staff follow hygienic care and procedures (e.g., wearing gloves).
7. Employees are dressed neatly.
8. Drugs are available.
9. Comfortable ambient conditions with proper lighting.
10. The hospital has an appealing atmosphere.
11. The hospital has clean rooms without foul smell.
12. There are sufficient waiting areas for patients and families.
13. It is easy to find my way in the hospital.
14. It is easy to find care facilities (laboratory, Doctor’s office).
15. It is easy to use the amenities (public telephone, cafeteria, etc.).
16. Courtesy is shown by hospital staff towards patient and patient party.
17. Doctors and nurses are available as and when required.
18. The doctors are competent and skilful.
20. Paramedical and support staff are competent.
21. The doctors are friendly, caring and understand patient’s feelings and needs.
22. Doctors talked to me frankly and politely.
23. Nurses give prompt and timely attention.
24. Nursing staff are polite and well-mannered.
25. Staff at the hospital are polite.
26. Doctors are professional.
27. Good doctors are available in the hospital.
28. The hospital has positive reputation.
29. Sincerity, honesty and ethics followed in providing medical services.
30. The hospital staff run various programs for patients to support different societal sections.
31. The hospital provides medical services with nominal cost to the needy patients.
32. Patient privacy and confidentiality are maintained by hospital staff.
33. Hospital staff provide services as promised and on time.
34. Equal treatment for all.
35. Confidence in the doctor who treated you in the hospital.
36. The hospital provides patients with services beyond medical treatment.
37. Faultless assessment of health conditions by doctors.
38. Explanations are provided by the doctor about health status, medical tests.
39. Treatment procedures and outcomes are explained.
40. Medical advice and instructions provided by doctor.
41. Diagnosis is only made after careful examination.
42. Doctors spent enough time examining the patient.
43. Doctors provide information quickly.
44. My family was told what they needed to know.
45. Hospital staff provided adequate information about my illness/treatment(s).
46. Information can be easily obtained.
47. Obtaining information from hospital administrative personnel (e.g., admission, treatment, discharge) is easy.
48. Extent to which doctors’ answer patient’s questions and explain treatment that I could understand.
49. I feel good about the interaction I have with doctors at the hospital.
50. I feel good about the interaction I have with nurses at the hospital.
51. I feel good about the interaction I have with staff at the hospital.
52. I have built a close relationship with some staff at the hospital.
53. I have built a close relationship with the doctor at the hospital.
54. I have built a close relationship with nurses at the hospital.
55. I always get personalized attention from staff at the hospital.
56. Hospital staff treat me as a human being and not just a patient.
57. The doctor calls my name while addressing me.
58. Waiting time to consult the doctors is minimum.
59. Staff provide right patient services the first time, every time.
60. Reasonable waiting time spent for diagnostic test and treatments.
61. Time between the two successive processes is minimum.
62. The process for setting up the appointment was simple and easy.
63. Appointments at the hospital run on time.
64. The hospital’s records and documentation are error free.
65. The interaction among department staff is well managed.
66. I believe the hospital is well-managed.
Perceived Service Quality (Brady and Cronin, 2001; Dagger et al., 2007; Parasuraman et al., 1988)
67. The overall service quality provided by hospital staff is excellent.
68. The overall service quality provided by hospital staff is impressive.
69. The service provided by the hospital is a high standard.
70. I believe the hospital offers service that is superior in every way.

Satisfaction (Dagger et al., 2007; Greenfield and Attkisson 1989; Oliver 1997)
71. My feelings towards the hospital are very positive.
72. I feel good about coming to this hospital for my treatment.
73. Overall, I am satisfied with the hospital and the service it provides.
74. I feel satisfied that the treatment results are the best that can be achieved.
75. The extent to which my treatment has produced the best possible outcome is satisfying.

Behavioral Intention (Dagger et al., 2007; Headley and Miller 1993; Taylor and Baker 1994; Zeithaml et al., 1996)
76. If I had to start treatment again, then I would want to come to this hospital.
77. I would highly recommend the hospital to other patients.
78. I have said positive things about the hospital to my family and friends.
79. I intend to continue having treatment, or any follow-up care I need, at this hospital.
80. I have no desire to change hospital.
81. I intend to follow the medical advice given to me at the hospital.
82. I am glad to have my treatment at this hospital rather than somewhere else.
Patient-perceived hospital service quality: an empirical assessment

Introduction
Healthcare is considered a ‘credence’ good – an offering that consumers will never be able to evaluate owing to deficient medical knowledge (Bloom and Reeve, 1990). Additionally, conceptualising and measuring service quality in a healthcare setting is more important and simultaneously more complex (Taner and Antony, 2006). However, researchers need to come up with ways to measure healthcare service quality, because unless we measure we cannot manage and improve healthcare services (Lohr, 2015). The literature indicates that there is a variability and confusion in how quality is conceptualized and operationalized (Sower et al., 2001). Many researchers attempt to define and conceptualize hospital service quality. Pai and Chary (2013), for example, show that the Parasuraman et al., (1985, 1988) SERVQUAL/modified SERVQUAL questionnaire is conventionally practiced in healthcare. They also pointed out studies where SERVQUAL items have not loaded onto respective dimensions highlighting a five-component structure that is lacking, suggesting a new questionnaire needs developing. New instruments, designed for healthcare setting: PRIVHEALTHQUAL (Ramsaran-Fowdar, 2008) for private and PubHosQual (Aagja and Garg, 2010) for public settings, have emerged. Studies adopting these instruments are scarce because these scales are hospital specific and not a general scale that measures hospital service quality in any hospital context. An instrument that could measure hospital service quality (HSQ) in any hospital settings has gained importance. Pai and Chary (2016) proposed their conceptual framework for measuring hospital service quality using nine dimensions. Their framework, unlike other instruments such as SERVQUAL (Parasuraman et al., 1988), PRIVHEALTHQUAL (Ramsaran-Fowdar, 2008) and PubHosQual (Aagja and Garg, 2010) has an additional dimension (relationship) in line with researchers such as Carman (1990) and Reynoso and Moore (1995), who suggested adding service specific dimensions to SERVQUAL, there exists no empirical studies to support or refute their conceptual framework. As there are no research studies that used Pai and Chary’s framework (2016), our purpose is to empirically appraise this framework and addresses literature gaps.

Objectives
We conducted our study with the objective to test the Pai and Chary (2016) conceptual framework’s validity and reliability. We also aim to assess HSQ through nine dimensions: (i) healthscape; (ii) personnel; (iii) hospital image; (iv) trustworthiness; (v) clinical process; (vi) communication; (vii) relationship; (viii) personalization and (ix) administrative procedures.

Structured questionnaire
The questionnaire included:

1. Respondent demographics – gender, age, occupation, educational level, marital status and income.
2. Sixty-six items covering nine dimensions (Pai and Chary, 2016).
3. Four items measuring respondents’ hospital service quality perception.
4. Twelve items that measured respondent satisfaction and behaviour.

A Likert (1932) scale, having a balanced rating with equal categories above and below the midpoint, was used with each item; anchored with verbal statements like ‘strongly disagree’, ‘disagree’, ‘neither agree or disagree’, ‘agree’, ‘strongly agree’. All items were phrased
Sixty-six hospital service quality items were categorized into nine dimensions: (i) healthscape (15 items); (ii) personnel (11); (iii) hospital image (5); (iv) trustworthiness (5); (v) clinical process (6); (vi) communication (9); (vii) relationship (3); (viii) personalization (3); and (ix) administrative procedures (9 items). Four perceived service quality, twelve satisfaction and behavioural intention items were included (Appendix 1).

Piloting
Piloting a new instrument is imperative (DeVellis, 1991; Kirchhoff, 1999). Before the pilot test is performed, it is advisable to obtain the original evaluation (DeVellis, 1991). Andres (2012) recommends that conducting a pilot study by involving colleagues, friends and family members who assume an audience role. Our questionnaire was subjected to the pre-testing process by circulating it among ten colleagues, friends and family members. In any research, respondents often misunderstand words or concepts. Although communication difficulties exist, respondents still provide legitimate answers to survey questions (Clark and Schober, 1992; Tourangeau et al., 2000). To overcome these problems, Collins (2003, p.230) suggested cognitively testing survey questions that helps ‘to identify how and where the question fails to achieve its measurement purpose’. Although there are various cognitive methods that have been developed and applied to instrument testing, such as cognitive interviewing, paraphrasing, card sorts, vignettes, confidence ratings and response latency timing (Czaja, 1998; Forsyth and Lessler, 1991; Jobe and Mingay, 1991); cognitive interviewing is becoming widespread (Schwarz, 1997). Cognitive interviewing (involving two methods: think aloud interviewing and probing) and paraphrasing were used in testing. On completion, the instrument was tried in a hospital with patients as respondents. According to Lackey and Wingate (1998), pilot testing a newly developed instrument should be undertaken with respondents selected from the same population from which the subjects in the major study will be selected. Consequently, the questionnaire was tested with respondents from teaching, corporate and public hospitals. Pre-testing ensured correct phrasing, format, length and question sequence. Pretesting was done with 30 respondents using Hertzog’s (2008) guidance. The questionnaire was corrected after feedback.

Data collection
Data were collected in three hospitals – teaching, corporate and public in Karnataka. India has 29 states and seven union territories (Ashok, 2014), and every state has a specific language. Although India is an English-speaking country, Kannada is Karnataka’s local language. Hence, the questionnaires were in Kannada and translated into the Malayalam language as few hospitals had Malayalam speaking patients in significant numbers owing to hospital’s proximity to another state (Kerala), where the official language is Malayalam. Consequently, questionnaires were administered in three languages: English; Kannada and Malayalam. There are several studies in which the questionnaire was administered in different languages specific to the country studied; e.g., English and Maltese (Camilleri and O’Callaghan, 1998); English and Arabic (Jabnoun and Chakar, 2003); English and Turkish (Kara et al., 2005); English and Gujarati (Aagja and Garg, 2010); and English and Bengali (Akter et al., 2008). The Kannada/Malayalam versions were created through careful translation and back-translation techniques (Candell and Hulin, 1987; McGorr, 2000). First, the questionnaire was translated into Kannada/Malayalam and then the Kannada/Malayalam items were back-translated into English by a bilingual expert to ensure that
the original content was kept intact. In translating the scale items into Kannada/Malayalam, Malinowski’s (1935) four-step translation technique was implemented:

1. An interlinear, or word-by-word, translation.
2. A ‘free’ translation in which clarifying terms, conjunctions, etc., are added and words reinterpreted.
3. Analysing and collating two translations leading to:
4. A contextual specification of meaning.

An attempt was made to remove discrepancies between English and Kannada/Malayalam questionnaire versions. No individual items were found to be problematic in translation. Previous studies reported backward translation; e.g., English to Bengali (Akter et al., 2008; Andaleeb, 2000); English to Japanese (Amira, 2008); English to Arabic (Mostafa, 2005); English to Hindi (Rao et al., 2006). Accordingly, our instrument was translated from English to Kannada and Malayalam languages.

Sample
The study population was defined as all patients 18 years or older with a stable mental and clinical condition during the data collection procedure. A stratified random sampling was adopted for the study. As the name implies, there is a stratification or segregation process followed by randomly selecting subjects from each stratum. The population is first divided into mutually exclusive groups which are relevant, appropriate and meaningful in the study context. The teaching, corporate and public hospitals constitute three strata. Hospitals were selected using systematic sampling and proportionate stratified sampling. The sampling frame was prepared for three strata and ten per cent from each stratum: three hospitals each from teaching and public hospitals and four corporate hospitals, totalling ten hospitals. Respondent from ten hospitals were chosen randomly (every fifth element in the population), both inpatients and outpatients yielding 602 respondents. The respondent’s demographic profile is depicted in the Table I, comprising males (44%). Their sample’s age ranged from 18 to 67+ years.

Table I here

Reliability and validity
Instruments designed to measure a specific concept should measure what is set out to measure. Reliability and validity are the two main criteria for measuring how good measures are in any research instrument. Reliability indicates stability and consistency with which the instrument measures the concept, thus assessing a measure’s goodness (Sekaran and Bougie, 2010). It is important to calculate scale reliability, which refers to the extent to which a scale can reproduce the same results in repeated trials (Hair et al., 2003). Consistency can be examined through the inter-item consistency reliability and split-half-reliability tests (Sekaran and Bougie, 2010). The internal consistency reliability test is acceptable when the reliability coefficient exceeds Nunnally’s (1978) 0.7 reliability criterion.

The questionnaire included 82 items (Appendix 1). For internal consistency, the most popular test is coefficient alpha (Cronbach). For Pai and Chary’s (2016) 66 items, the value is 0.965 and 0.972 for the 82-item instrument. The higher the coefficients, the better the measuring instrument (Sekaran and Bougie, 2010). Another technique for testing internal consistency is the
split-half technique, where items constituting the instrument are divided into two halves and the resulting half scores are correlated (Malhotra and Dash, 2012). For the 66-item version, Cronbach’s Alpha was 0.93 and 0.947, correlation between forms was 0.793, Spearman–Brown Coefficient for equal length and unequal length is 0.884, Guttman Split-Half Coefficient is 0.878. For the 82-item questionnaire, Cronbach’s Alpha was 0.944 and 0.956, correlation between forms was 0.821, Spearman–Brown Coefficient for equal length and unequal length is 0.902, Guttman Split-Half Coefficient is 0.897. We can conclude, therefore, that the instrument has adequate reliability. Cronbach’s alpha is a widely-used reliability coefficient that assesses the entire scale’s internal consistency and averages all possible split-half coefficients resulting from different ways of splitting the scale items ranging from 0 to 1. According to Malhotra and Dash (2012), it is appropriate to use internal consistency reliability for each dimension, if several items are used to measure each dimension. In our study, patient-perceived hospital service quality is measured using different dimensions, each is measured by several items, hence we used Cronbach’s alpha to measure internal consistency. Cronbach’s alpha for the twelve constructs considered in the study was measured and values for each construct are 0.70 and above, indicating a strong reliability (Table II).

**Table II here**

Although reliability is a necessary contributor to validity, it is not a sufficient condition for validity (Sekaran and Bougie, 2010). Therefore, validity becomes equally important. Validity is ‘the extent to which a rating scale truly reflects the underlying variable that it attempts to measure’ (Parasuraman et al., 2004, p. 294). Content validity, criterion-related validity and construct validity are generally discussed in research articles.

**Content validity**

Content validity is the degree to which items adequately represents all relevant items under study (Cooper et al., 2012), determined using judgment and panel evaluation. For an attitudinal scale, content validity is an overall criterion that can be assessed only though a researcher’s subjective judgment (Parasuraman et al., 2004). We exercised judgment through carefully defining and analysing conceptual and empirical frameworks through an extensive literature review. Judges can attest to content validity (Sekaran and Bougie, 2010), so the questionnaire was subjected to expert review by practitioners and academics. Content/face validity was confirmed; i.e., the proposed measurement instrument measures the intended construct. The items for the current study were chosen from the literature (Sekaran and Bougie, 2010). Scales were refined using a pilot study. Patients revealed they had no difficulty understanding the questionnaire items indicating and confirming face validity (Arasli et al., 2008). All steps ensured that the instrument possesses face validity. Although implementing a scale based on only face validity claims has a far greater impact than item’s random deletion through mere formal scale refinement methods (Finn and Kayande, 2004), nevertheless, other validity tests are also conducted.

Criterion related validity (Ping, 2004) emphasizes that, for new measures, criterion validity should be assessed, which is established when the measure differentiates individuals on a criterion it is expected to predict. Criterion related validity can be established by concurrent or predictive validity (Sekaran and Bougie, 2010). Predictive validity indicates the instrument’s ability to differentiate among individuals with reference to a future criterion (Sekaran and Bougie, 2010). The researcher needs to ensure that the validity criterion used is valid and the
intended measure is judged on four qualities: relevance; freedom from bias; reliability; and availability (Cooper et al., 2012). Criterion-related validity was established using predictive validity, adopting Jabnoun and Chakar’s (2003) technique, who correlated their service quality dimensions with overall service quality. Individual item scores on nine dimensions (listed earlier) were summed to obtain overall scores for each respondent. Scores were then correlated with a summated perceived service quality, satisfaction and behavioural intentions scale. All items were captured and measured using Likert scales with a five-point response format. A higher score indicated a more favourable response (Table III). All correlations were positive and statistically significant at 0.001 level, which establishes predictive validity.

Table III here

Concurrent validity was assessed by examining the association between perceived service quality and a related variable, patient satisfaction (Hansen et al., 2008). Patient satisfaction is a distant but related construct to perceived service quality (Choi et al., 2004; Rao et al., 2006). Although there are differences between patient satisfaction and perceived service quality, they are related and a moderate correlation is expected (Al-Qatari and Haran, 1999). Crompton and Love (1995) stressed that the two constructs are likely to be positively correlated, but unlikely to be linear. The correlation between perceived service quality and patient satisfaction is 0.633 (Table III), which indicates concurrent validity. Concurrent validity assessment is conducted in line with Duggirala et al., (2008) such that in the present context, perceived service quality (PSQ) is chosen as the nine-dimension’s outcome and a bivariate correlation analysis was carried out among all constructs, which have significant positive correlations with the criterion chosen, namely, PSQ (range 0.414 to 0.587, p < 0.001). According to Sekaran and Bougie (2010), construct validity testifies how well the results obtained from a measure fit the theories around which the test is designed and assessed through convergent and discriminant validity. Convergent validity is the degree to which scores on one scale correlate with scores on other scales designed to assess the same construct (Cooper et al., 2012). Convergent validity is also assessed by calculating one-way analysis of variance (ANOVA) (Aagia and Garg, 2010; Bahia and Nantel, 2000; Parasuraman et al., 1988; Wong et al., 2001). The association between healthscape, personnel, hospital image, trustworthiness, clinical care process, communication, relationship, personalization and administrative procedures and PSQ was significant at less than 1% (Table IV), which means that the group differences are significant; i.e., good convergent validity.

Table IV here

Alternate methods for establishing criterion and construct validity

Campbell and Fiske’s (1959) procedure has been used as an alternate method for establishing criterion and construct validity, adopted in similar studies such as Boshoff and Terblanche, 1997; Kaul 2007. Convergent validity (Table V) is confirmed as there is a high correlation between scale items and PSQ (correlation coefficient, \( r = 0.665, p<0.001 \)). To assess predictive validity, respondents were asked whether they intended to return to the same hospital (repurchase intentions). Results confirm predictive validity (\( r = 0.621, p<0.001 \)).

Table V here
Empirically validating nine patient-perceived hospital service quality dimensions

We propose the following hypothesis:

\[ H_0: \] The heathscape, personnel, hospital image, trustworthiness, clinical process, communication, relationship, personalization and administrative procedure domains do not predict hospital service quality.

To test our hypothesis, a multiple regression approach is adopted with the nine dimensions as independent variables and PSQ as the dependent variable. Variables were examined for multicollinearity and autocorrelation before the regression analysis was run. Multicollinearity exists when there is a strong correlation between two or more predictors in a regression model leading to a regression model that fits the data, but no variable had a significant impact in predicting the dependent variable. In SPSS, there are collinearity diagnostics that are variance inflation factor (VIF) and tolerance statistics. The VIF indicates whether the predictor has a strong linear relationship with other predictor(s), while the tolerance statistic reciprocates the variance inflation factor (VIF). We used the Durbin-Watson test to identify any serial correlations between errors (Table VI and VII).

Table VI here

The tolerance values need to be greater than 0.1 and the VIF values should be less than 10 (Ho, 2006). From the tolerance and VIF values shown in Table VI, multicollinearity does not exist in the predictor values. Hence, the nine independent variables influence on dependent variables is analysed. The regression model is formed using the statistical (stepwise) regression method taking all nine independent variables together and examining their combined effect on the dependent variable. In the statistical (stepwise) regression model, the predictor variables’ order is solely based on statistical criteria and variables that correlate most strongly with the dependent variable will be given entry priority. Our regression models require no references to theoretical considerations, and are primarily used when the researcher is unsure about the independent variable’s predictive power (Ho, 2006). This method is apt for the present study and therefore stepwise regression is adopted, which includes forward inclusion, backward elimination and stepwise solution (Ho, 2006; Malhotra and Dash, 2012). To identify an optimal regression equation, one needs to compute combinational solutions in which all possible combinations are examined (Malhotra and Dash, 2012). The optimal regression equation in our study is computed using all possible combinations. However, the stepwise regression method is preferable to the forward method, as they reduce Type II errors (i.e., missing a predictor that does in fact predict the outcome) (Field, 2013). Regression analysis results are shown in Table VII.

Table VII here

The F value tests how well the regression model fits the data (Ho, 2006). In case the probability associated with the F statistics is small, the hypothesis that R-square is equal to zero is rejected. In our case, F is 71.65, \( p < 0.001 \) (Table VII); i.e., the hypothesis that there is no linear relationship between the predictor and the dependent variables is rejected. The R statistic indicates that the correlation coefficient as 0.677 and the R-square value is 0.458. For our
sample, the predictor variables explained 45.8 per cent of the variance in the dependent variable. Hence, the null hypothesis $H_0$ is rejected. In Table VII, Durbin-Watson’s $d = 1.987$. Durbin-Watson can vary between 0 and 4; i.e., 2 implies that the residuals are uncorrelated. If the value is greater than 2, then it indicates a negative correlation between adjacent residuals; whereas a value below 2 indicates a positive correlation (Field, 2013). Thus, Durbin-Watson statistic tests serial correlations between errors, a metric that assesses another linear model assumption. Field (2013) suggests that the stepwise method should be avoided except for exploratory model building. He recommends cross-validating the model if the stepwise method is used, suggesting two cross-validation methods: adjusted $R^2$ and data splitting.

**Cross-validating the model using adjusted $R$ square ($R^2$)**

The adjusted $R^2$ shows how well our model generalizes. Although the adjusted $R^2$ should be the same or very close to $R^2$, it differs generally. This difference is known as shrinkage, which is $R^2 - \text{Adjusted } R^2$. As per model summary (Table VII) shrinkage is equal to $0.458 - 0.451 = 0.007$ (about 0.7 per cent); i.e., the model’s shrinkage is small. Although there are no shrinkage guidelines in the literature, it is valuable cross-validation such that one can be confident in the equation’s generalizability if they are similar (Pedhazur, 1997). According to Field (2013), shrinkage means that if the model were derived from the population rather than a sample, then it would account for approximately 0.7 per cent less variance in the outcome. Stein’s formula helps us to calculate adjusted $R^2$ and tells us how well the model cross-validates (Stevens, 2002), which is computed as: $\text{Adjusted } R^2 = 1-[(n-1/n-k-1) (n-2/n-k-2) (n+1/n)] (1- R^2)$ where $R^2$ is the unadjusted value, $n$ is total participants and $k$ is total predictors in the model. Substituting $R^2$ as 0.458 (Table VII), $n = 602$ and $k=9$ in the above formula, we get 0.441, which is computed as $1-[(602-1/602-9-1) (602-2/602-9-2) (602+1/602)] (1- 0.458)$, which = 0.441. This value is like the observed 0.458, indicating that cross-validity is acceptable (Field, 2013, p.312).

**Cross-validating the model using data splitting**

The other cross-validation method is the data splitting approach. Cross-validation is a validity test that examines whether the regression model continues to hold on comparable data not used in the original estimation (Malhotra and Dash, 2012). In a cross-validation procedure, we estimate the regression model for the entire data, which are split into two parts: estimation and validation samples. The estimation sample includes 50-90% of the total sample. The regression model is estimated using the estimation sample only. This model is compared with the entire sample. The estimated model is applied to the data in the validation sample to predict the dependent variable. In case there is a large discrepancy between $R^2$ for the smaller and larger samples, this indicate over fitting and weak generalizability (Tabachnick and Fidell, 2014). In other words, if there is a significant difference between $R^2$ for the larger sample (estimation sample) and smaller sample (validation sample) then there is weak generalizability. Using SPSS, the step-wise regression on a random selection is adopted as suggested by Field (2013). Cases are randomly divided into two parts for cross-validation. Creating a variable that contains information about the estimation sample and validation samples using 0 and 1 to indicate the estimation sample and validation sample. We split the data into two samples: estimation and validation samples with 325 (53.9%) and 277 (46.1%) respondents/samples using Malhotra and Dash, (2012) guidance that estimation sample includes 50-90% of the total sample. Stepwise regression is conducted on the split data; i.e., estimation and validation sample, and the $R^2$ values are compared with the original model (Malhotra and Dash, 2012; Tabachnick and Fidell, 2014).
The overall relationship between the dependent variable (PSQ) and the independent variable (nine dimensions) must be statistically significant for both validation analyses. Table VIII and IX indicate that significance is achieved. The $R^2$ for each validation must be within ±5% (plus or minus 5%) of the model’s $R^2$ using the full sample. The entire sample’s $R^2$ is 45.8%; thus, $R^2$ and validation models can vary by 5% (40.8% 50.8%). Table X shows the split sample $R^2$ as 48.4% and 44.1%, whereas the full sample is 45.8%. The split sample validation’s value is confirmed; i.e., cross-validation is verified and supports our finding’s generalizability.

Table VIII, IX and X here

Discussion
Appraising reliability and validity tests
Internal consistency is the degree to which various multidimensional construct measures correlate with the scale (Hair et al., 2003). They indicate the item’s homogeneity in the measure that captures the construct after administering the instrument. All approaches adopted for testing reliability and validity are as shown in Table XI.

Table XI here

Adopting Cronbach’s alpha and split half technique show that the scale demonstrates reliability. The instrument was piloted and subjected to expert evaluation; i.e., demonstrating content validity. Concurrent and predictive validity have demonstrated criterion related validity. Discriminant validity was not assessed, convergent validity demonstrated construct validity. Aagja and Garg (2010); Kaul (2007) did not successfully demonstrate discriminant validity. Although there are different studies that measured healthcare service quality, only Duggirala et al., (2008) tested and supported all three: content, criterion and construct validity (Table XII). Pai and Chary (2013) observed that hospital service quality studies have neither evaluated all three validity methods nor discussed them. All three validity methods applied to Pai and Chary’s (2016) framework demonstrate instrument validity. The instrument, therefore, is deemed both reliable and valid for studying hospital service quality.

Table XII here

Pai and Chary’s (2016) conceptual framework, comprising nine dimensions, measures hospital service quality using multiple regression. Results indicate that the null hypothesis can be rejected. According to Salkind (2000), an R (correlation coefficient) between 0.4 to 0.6 is moderate; 0.6 to 0.8 is strong and 0.8 to 1 is a strong to perfect association. The R in our study is 0.677 (Table VII) indicating a strong association between dependent and independent variables (p < 0.001). The model is also cross validated using adjusted $R^2$ and data splitting (Field, 2013).

Conclusion
Our objective was twofold, first to evaluate the Pai and Chary (2016) conceptual framework’s reliability and validity and second to evaluate their proposed nine-dimension structure for measuring patient-perceived hospital service quality. Our study provides empirical evidence that the scale developed to measure hospital service quality is reliable and valid. We show that the nine dimensions predict hospital service quality.
Theoretical contribution and managerial implications
The study has strong implications for hospital quality managers seeking to measure hospital service quality who can implement the structured questionnaire in any hospital setting. Managers can also use the instrument to compare service quality among different combinations such as teaching, public and corporate hospitals.

Scope for future studies
Our study is cross-sectional and limited to Karnataka state hospitals. Future studies should use a longitudinal study to provide greater diagnostic value. The instrument’s validity and reliability can also be tested in other developing countries.

References


Ho, R. (2006), Handbook of Univariate and Multivariate Data Analysis and Interpretation with SPSS, Taylor and Francis group, USA.


Table I: Demographic profile

<table>
<thead>
<tr>
<th>Demographic Factors</th>
<th>Valid Items</th>
<th>Frequency</th>
<th>Percentage</th>
</tr>
</thead>
</table>

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<table>
<thead>
<tr>
<th>Gender</th>
<th>Male</th>
<th>267</th>
<th>44.4</th>
</tr>
</thead>
<tbody>
<tr>
<td>Female</td>
<td>327</td>
<td>54.3</td>
<td></td>
</tr>
</tbody>
</table>

| Age (in Years) | | |
|----------------|---|---|---|
| 18-27          | 163 | 27.1 |
| 28-37          | 159 | 26.4 |
| 38-47          | 102 | 16.9 |
| 48-57          | 68  | 11.2 |
| 58-67          | 64  | 10.6 |
| 68 and above   | 37  | 6.1 |

**Table II: Cronbach alpha values**

<table>
<thead>
<tr>
<th>Constructs</th>
<th>Cronbach’s Alpha</th>
</tr>
</thead>
<tbody>
<tr>
<td>Healthscape</td>
<td>0.875</td>
</tr>
<tr>
<td>Personnel</td>
<td>0.869</td>
</tr>
<tr>
<td>Hospital Image</td>
<td>0.739</td>
</tr>
<tr>
<td>Trustworthiness</td>
<td>0.805</td>
</tr>
<tr>
<td>Clinical Care Process</td>
<td>0.823</td>
</tr>
<tr>
<td>Communication</td>
<td>0.869</td>
</tr>
<tr>
<td>Relationship</td>
<td>0.845</td>
</tr>
<tr>
<td>Personalization</td>
<td>0.730</td>
</tr>
<tr>
<td>Administrative Procedures</td>
<td>0.896</td>
</tr>
<tr>
<td>Perceived Service Quality</td>
<td>0.873</td>
</tr>
<tr>
<td>Customer Satisfaction</td>
<td>0.811</td>
</tr>
<tr>
<td>Behavioural Intentions</td>
<td>0.882</td>
</tr>
</tbody>
</table>

**Table III: Bivariate correlations among the dimensions**

<table>
<thead>
<tr>
<th></th>
<th>HS</th>
<th>PL</th>
<th>HI</th>
<th>TW</th>
<th>CCP</th>
<th>COM</th>
<th>REL</th>
<th>PER</th>
<th>AP</th>
<th>PSQ</th>
<th>SAT</th>
<th>BI</th>
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<tr>
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<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td></td>
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<td></td>
<td></td>
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<tr>
<td>PL</td>
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<td>.668</td>
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<td></td>
<td></td>
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<td></td>
<td></td>
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<tr>
<td>HI</td>
<td>.509</td>
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<td>.574</td>
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<td></td>
<td></td>
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<td></td>
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<td></td>
</tr>
<tr>
<td>TW</td>
<td>.603</td>
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<td>.508</td>
<td>1</td>
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<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>CCP</td>
<td>.600</td>
<td>.675</td>
<td>.604</td>
<td>.631</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>COM</td>
<td>.605</td>
<td>.673</td>
<td>.505</td>
<td>.646</td>
<td>.730</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>REL</td>
<td>.420</td>
<td>.462</td>
<td>.386</td>
<td>.454</td>
<td>.472</td>
<td>.567</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>PER</td>
<td>.446</td>
<td>.494</td>
<td>.423</td>
<td>.418</td>
<td>.523</td>
<td>.556</td>
<td>.582</td>
<td>1</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>AP</td>
<td>.526</td>
<td>.565</td>
<td>.450</td>
<td>.586</td>
<td>.581</td>
<td>.693</td>
<td>.520</td>
<td>.521</td>
<td>1</td>
<td></td>
<td></td>
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</tr>
<tr>
<td>PSQ</td>
<td>.55</td>
<td>.551</td>
<td>.483</td>
<td>.548</td>
<td>.564</td>
<td>.587</td>
<td>.445</td>
<td>.414</td>
<td>.531</td>
<td>1</td>
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<td></td>
</tr>
</tbody>
</table>
All are positive and significant at 0.001.

**Table IV:** One-way ANOVA for assessing convergent validity – independent variable service quality

<table>
<thead>
<tr>
<th>Descriptive Variables</th>
<th>Degrees of freedom</th>
<th>F</th>
<th>Sig.</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Between Groups</td>
<td>Within Groups</td>
<td></td>
</tr>
<tr>
<td>Healthscape</td>
<td>13</td>
<td>588</td>
<td>24.219</td>
</tr>
<tr>
<td>Personnel</td>
<td>13</td>
<td>588</td>
<td>28.381</td>
</tr>
<tr>
<td>Image</td>
<td>13</td>
<td>588</td>
<td>23.463</td>
</tr>
<tr>
<td>Trustworthiness</td>
<td>13</td>
<td>588</td>
<td>28.694</td>
</tr>
<tr>
<td>Clinical Care Process</td>
<td>13</td>
<td>588</td>
<td>29.622</td>
</tr>
<tr>
<td>Communication</td>
<td>13</td>
<td>588</td>
<td>31.435</td>
</tr>
<tr>
<td>Relationship</td>
<td>13</td>
<td>588</td>
<td>17.029</td>
</tr>
<tr>
<td>Personalization</td>
<td>13</td>
<td>588</td>
<td>13.205</td>
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<td>Administrative Procedures</td>
<td>13</td>
<td>588</td>
<td>25.967</td>
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</table>
### Table V: Validity tests

<table>
<thead>
<tr>
<th>Validity</th>
<th>Measures used</th>
<th>Correlation coefficient (r)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Convergent validity</td>
<td>Perceived service quality</td>
<td>0.665 (p&lt;0.001)</td>
</tr>
<tr>
<td>Predictive validity</td>
<td>Repurchase intentions</td>
<td>0.621 (p&lt;0.001)</td>
</tr>
</tbody>
</table>

### Table VI: Collinearity statistics

<table>
<thead>
<tr>
<th>Independent Variable</th>
<th>Tolerance*</th>
<th>VIF*</th>
</tr>
</thead>
<tbody>
<tr>
<td>Healthscape (HS)</td>
<td>.475</td>
<td>2.104</td>
</tr>
<tr>
<td>Personnel (PL)</td>
<td>.347</td>
<td>2.879</td>
</tr>
<tr>
<td>Hospital Image (HI)</td>
<td>.567</td>
<td>1.765</td>
</tr>
<tr>
<td>Trustworthiness (TW)</td>
<td>.413</td>
<td>2.421</td>
</tr>
<tr>
<td>Clinical Care Process (CCP)</td>
<td>.353</td>
<td>2.831</td>
</tr>
<tr>
<td>Communication (COM)</td>
<td>.308</td>
<td>3.248</td>
</tr>
<tr>
<td>Relationship (REL)</td>
<td>.559</td>
<td>1.787</td>
</tr>
<tr>
<td>Personalization (PER)</td>
<td>.544</td>
<td>1.838</td>
</tr>
<tr>
<td>Administrative Procedures (AP)</td>
<td>.455</td>
<td>2.198</td>
</tr>
</tbody>
</table>

*assessed using Enter regression method
Table VII: Model summary

<table>
<thead>
<tr>
<th>R</th>
<th>R Square</th>
<th>Adjusted R Square</th>
<th>F</th>
<th>Sig.</th>
<th>Durbin-Watson</th>
</tr>
</thead>
<tbody>
<tr>
<td>.677</td>
<td>.458</td>
<td>.451</td>
<td>71.65</td>
<td>.001</td>
<td>1.987</td>
</tr>
</tbody>
</table>

Table VIII: Split data model summary - estimation sample – 0

<table>
<thead>
<tr>
<th>R</th>
<th>R Square</th>
<th>Adjusted R Square</th>
<th>F</th>
<th>Sig.</th>
<th>Durbin-Watson</th>
</tr>
</thead>
<tbody>
<tr>
<td>Split = 0.00 (Selected)</td>
<td>.696</td>
<td>.623</td>
<td>.484</td>
<td>75.059</td>
<td>.001</td>
</tr>
<tr>
<td>split =~= 0.00 (Unselected)</td>
<td></td>
<td></td>
<td>.478</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Table IX: Split data model summary - validation sample – 1

<table>
<thead>
<tr>
<th>R</th>
<th>R Square</th>
<th>Adjusted R Square</th>
<th>F</th>
<th>Sig.</th>
<th>Durbin-Watson</th>
</tr>
</thead>
<tbody>
<tr>
<td>Split = 1.00 (Selected)</td>
<td>.664</td>
<td>.652</td>
<td>.441</td>
<td>53.618</td>
<td>.001</td>
</tr>
<tr>
<td>split =~= 1.00 (Unselected)</td>
<td></td>
<td></td>
<td>.433</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Table X: Summary - Validation results: stepwise regression

<table>
<thead>
<tr>
<th>ANOVA significance (sig&lt;=0.05)</th>
<th>Entire data</th>
<th>Estimation sample</th>
<th>Validation sample</th>
</tr>
</thead>
<tbody>
<tr>
<td>R²</td>
<td>0.458</td>
<td>0.484</td>
<td>0.441</td>
</tr>
</tbody>
</table>

Table XI: Reliability and validity tests

<table>
<thead>
<tr>
<th>Reliability and validity tests</th>
<th>Reliability</th>
<th>Content Validity</th>
<th>Criterion related Validity</th>
<th>Construct Validity</th>
</tr>
</thead>
<tbody>
<tr>
<td>split-half reliability tests</td>
<td>Literature</td>
<td>Predictive validity (2 tests)</td>
<td>Convergent validity (2 tests)</td>
<td></td>
</tr>
<tr>
<td>inter-item consistency reliability</td>
<td>Pilot testing</td>
<td>Concurrent validity (2 tests)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Table XII: Reliability and validity a comparison with established scales

|---------|-------------|---------------------------|----------------------|--------------------------|----------------------|
Appendix 1: Questionnaire

1. Modern and up-to-date equipment (e.g., Computerized tomography (CT) and Magnetic Resonance Imaging (MRI), patient information and billing) to serve patients more effectively.
2. Physical facilities are visually appealing.
3. Facility adequacy (e.g., wards, beds, operation theatre, Intensive Care Unit (ICU)).
4. Cleanliness (toilets, rooms and wards).
6. Hospital staff follow hygienic care and procedures (e.g., wearing gloves).
7. Employees are dressed neatly.
8. Drugs are available.
9. Comfortable ambient conditions with proper lighting.
10. The hospital has an appealing atmosphere.
11. The hospital has clean rooms without foul smell.
12. There are sufficient waiting areas for patients and families.
13. It is easy to find my way in the hospital.
14. It is easy to find care facilities (laboratory, Doctor’s office).
15. It is easy to use the amenities (public telephone, cafeteria, etc.).
16. Courtesy is shown by hospital staff towards patient and patient party.
17. Doctors and nurses are available as and when required.
18. The doctors are competent and skilful.
20. Paramedical and support staff are competent.
21. The doctors are friendly, caring and understand patient’s feelings and needs.
22. Doctors talked to me frankly and politely.

<table>
<thead>
<tr>
<th></th>
<th>Instrument</th>
<th>SERVQUAL</th>
<th>PubHosQual</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>Dimensional Structure</td>
<td>Five</td>
<td>Five</td>
</tr>
<tr>
<td>3</td>
<td>Application Domain</td>
<td>All services</td>
<td>Public Hospitals</td>
</tr>
<tr>
<td>4</td>
<td>Reliability</td>
<td>$0.72 &lt; \alpha &lt; 0.86$</td>
<td>$0.58 &lt; \alpha &lt; 0.89$</td>
</tr>
<tr>
<td>5</td>
<td>Content Validity</td>
<td>Measured</td>
<td>Measured</td>
</tr>
<tr>
<td>6</td>
<td>Construct Validity</td>
<td>Convergent Validity</td>
<td>Convergent and nomological validity are successful; discriminant validity had mixed results.</td>
</tr>
<tr>
<td>7</td>
<td>Criterion related validity</td>
<td></td>
<td>Concurrent validity</td>
</tr>
</tbody>
</table>
23. Nurses give prompt and timely attention.
24. Nursing staff are polite and well-mannered.
25. Staff at the hospital are polite.
26. Doctors are professional.
27. Good doctors are available in the hospital.
28. The hospital has positive reputation.
29. Sincerity, honesty and ethics followed in providing medical services.
30. The hospital staff run various programs for patients to support different societal sections.
31. The hospital provides medical services with nominal cost to the needy patients.
32. Patient privacy and confidentiality are maintained by hospital staff.
33. Hospital staff provide services as promised and on time.
34. Equal treatment for all.
35. Confidence in the doctor who treated you in the hospital.
36. The hospital provides patients with services beyond medical treatment.
37. Faultless assessment of health conditions by doctors.
38. Explanations are provided by the doctor about health status, medical tests.
39. Treatment procedures and outcomes are explained.
40. Medical advice and instructions provided by doctor.
41. Diagnosis is only made after careful examination.
42. Doctors spent enough time examining the patient.
43. Doctors provide information quickly.
44. My family was told what they needed to know.
45. Hospital staff provided adequate information about my illness/treatment(s).
46. Information can be easily obtained.
47. Obtaining information from hospital administrative personnel (e.g., admission, treatment, discharge) is easy.
48. Extent to which doctors’ answer patient’s questions and explain treatment that I could understand.
49. I feel good about the interaction I have with doctors at the hospital.
50. I feel good about the interaction I have with nurses at the hospital.
51. I feel good about the interaction I have with staff at the hospital.
52. I have built a close relationship with some staff at the hospital.
53. I have built a close relationship with the doctor at the hospital.
54. I have built a close relationship with nurses at the hospital.
55. I always get personalized attention from staff at the hospital.
56. Hospital staff treat me as a human being and not just a patient.
57. The doctor calls my name while addressing me.
58. Waiting time to consult the doctors is minimum.
59. Staff provide right patient services the first time, every time.
60. Reasonable waiting time spent for diagnostic test and treatments.
61. Time between the two successive processes is minimum.
62. The process for setting up the appointment was simple and easy.
63. Appointments at the hospital run on time.
64. The hospital’s records and documentation are error free.
65. The interaction among department staff is well managed.
66. I believe the hospital is well-managed.
Perceived Service Quality (Brady and Cronin, 2001; Dagger et al., 2007; Parasuraman et al., 1988)

67. The overall service quality provided by hospital staff is excellent.
68. The overall service quality provided by hospital staff is impressive.
69. The service provided by the hospital is a high standard.
70. I believe the hospital offers service that is superior in every way.
Satisfaction (Dagger et al., 2007; Greenfield and Attkisson 1989; Oliver 1997)

71. My feelings towards the hospital are very positive.
72. I feel good about coming to this hospital for my treatment.
73. Overall, I am satisfied with the hospital and the service it provides.
74. I feel satisfied that the treatment results are the best that can be achieved.
75. The extent to which my treatment has produced the best possible outcome is satisfying.
Behavioral Intention (Dagger et al., 2007; Headley and Miller 1993; Taylor and Baker 1994; Zeithaml et al., 1996)

76. If I had to start treatment again, then I would want to come to this hospital.
77. I would highly recommend the hospital to other patients.
78. I have said positive things about the hospital to my family and friends.
79. I intend to continue having treatment, or any follow-up care I need, at this hospital.
80. I have no desire to change hospital.
81. I intend to follow the medical advice given to me at the hospital.
82. I am glad to have my treatment at this hospital rather than somewhere else.