Safety Assessment of an Aging Offshore Jacket Platform by Integrating Analytic Hierarchy Process and Grey Clustering Method

Xingxian Bao¹*, Aixia Cao² and Fuzhen Qin²

¹ School of Petroleum Engineering, China University of Petroleum (East China), Qingdao 266580, Shandong, China
² School of Ship Engineering, Qingdao Huanghai University, Qingdao, 266427, Shandong, China
* Correspondence: baoxingxian@upc.edu.cn

Abstract: It is a significant task to assess the safety of the aging offshore jacket platform in order to extend the service life. This paper analyzes the multiple risk factors of an aging jacket platform in Bohai Bay, China and builds the safety evaluation index system, which includes three levels, namely, the target layer, the first-grade indicators layer and the second-grade indicators layer. The target layer consists of three first-grade indicators: ocean environments, structure status, and human and organizational factors. Each first-grade indicator is divided into three second-grade indicators. The weight of each indicator is calculated by analytic hierarchy process to weaken subjective effect. Grey clustering method is applied to estimate the failure risk of the platform in Bohai Bay qualitatively and quantitatively. The assessment standard is divided into five grades and the whitening function of each grey cluster is determined by the assessment scheme. The grey evaluation weight vector of each second-grade indicator is calculated by the table dispatching method. Through layer by layer calculation, the grey assessment value of the target layer is finally estimated by making the grey assessment weight vector single-value and the grey grade is determined according to the maximum principle. The evaluation results show quantitatively that the failure risk grade of the jacket platform in Bohai Bay is medium and the safety assessment method is reasonable and feasible.

Keywords: aging offshore jacket platform; safety assessment; analytic hierarchy process; grey clustering method

1. Introduction

Offshore jacket platforms have been widely used in offshore oil and gas exploitation. There are more than 5,000 different types of jacket platforms in different seas in the world. Safety is a significant challenge for offshore platforms due to the demanding ocean environment, structure damage and the fire and explosion risk associated with hydrocarbons [1-3]. Adequate performance
of the platforms is ensured by designing for a service life of 20 years or so. However, up to date, most of the jacket platforms have already attained or exceeded their original design life. The use of the existing structures will in many cases, even with major upgrading of the present condition, be preferable from an economic point of view compared to installing new buildings. Accordingly, there is a need to assess existing offshore structures during operation, for instance, because of planned change of function or occurrence of damage or need to extend the service life.

A safety assessment of existing jacket platform should prove that the risk of structural failure is acceptable [4]. In order to achieve this, several assessment procedures are proposed in specifications and standards. The most general accepted standard for offshore structures is the ISO 13819-1 “Petroleum and natural gas industries - Offshore Structures – Part 1: General Requirements” [5]. This standard gives general design rules and general rules for assessment of existing structures. The NORSOK N-001 [6], which is a Norwegian regulation for structural design, also refers to ISO 13819-1 for assessment of existing structures. However, ISO 13819-1 is a rather general standard, not very specific on how to perform assessment. The standard gives some indications, such as design code format, reserve strength ratio format and a probabilistic format. More details are given in other standards, like API RP2A-WSD [7], ISO/DIS 13822 [8] and ISO/CD 19902 [9]. However, there are some arguments against all these methods [4]. Design codes are claimed to be very conservative, as after some years of experience the knowledge of the structure is increased compared with design. The increased information should not be disregarded since they may make it possible to perform more accurate analysis. Methods based on reserve strength ratio do formally only check the reserve strength of a structure [10] and will not cover possible failure modes as fatigue and wave-in-deck. Structural reliability methods typically account for capacity versus loading, particularly deal with the uncertainties of structural loads and their effects as well as resistance [1, 11, 12]. However, the structural reliability methods are not sufficiently mature at present time, as the reliability approaches are not consistent with the derivation of the reliability target levels [4].

Accidents of offshore structures are not usually caused by a single failure or mistake, but by the confluence of a whole series, or chain, of errors. To reduce the risk, some risk analysis techniques have been used in offshore industry for almost 20 years, and have contributed to the reduction of the incidence rate of severe accidents in many circumstances [13]. For instance, a
systemic approach considering all the aspects that may lead to hazardous events is adopted in offshore platforms safety assessment. A systemic approach means considering all functional entities that constitute the offshore system as a whole, exploring patterns and inter-relationships within subsystems and seeing undesired events as the products of the working of the system [14]. However, the systemic approach mainly uses descriptive, not predictive models, and is thus often not very effective in determining ways to prevent accidents. In order to explore causal relationships in offshore structures safety assessment, some conventional tools have been widely used, including Fault Tree Analysis (FTA), Event Tree Analysis (ETA), Failure Mode and Effects Analysis (FMEA), and Hazard and Operability Studies (HAZOP). Currently, the statistic analysis method, which is both qualitative and quantitative, is drawing more and more attention. For instance, grey correlation analysis [15], Monte-Carlo method [16], BP neural network method [17], fuzzy comprehensive evaluation [18] and support vector machines method [19] have been applied to the risk assessment in engineering field.

The aim of the paper is to contribute to offshore platforms safety assessment by proposing a methodology to model evaluation system and determine the grade of the failure risk. We take an aging offshore jacket platform in Bohai Bay, China as a case. The analytic hierarchy process (AHP) is applied to establish the assessment index system with three layers, considering multiple risk factors in complicated ocean environments, structure status, and human and organizational factors (HOFs). The weight of each index is also calculated by AHP to weaken subjective effect. Through the layer by layer calculation, the failure risk of the jacket platform is finally evaluated qualitatively and quantitatively by the grey clustering method.

2. The Target Platform in Bohai Bay, China

The target platform studied in this paper is a steel jacket production platform that was built in 1998 in Bohai Bay, China. The original design life is 20 years. The jacket structure is 48m high and the water depth at the site is 19.5m. The production platform consists of two parts: the metering platform and the well group platform, which are connected by a trestle bridge. The metering platform and the well group platform are secured to the sea floor with 4 piles and 6 piles, respectively. The diameter of the pipes varies from 1.3m to 0.4m with depth, and the thickness
varies from 26mm to 14mm. An illustration of the platform frame is shown in Fig. 1. Although there are protection devices against corrosion in submerged elements of the platform, corrosion still happened, especially at the areas near the border of the submerged and atmospheric zone. The maximum corrosion amount is 2.6mm through the observation. After nearly 20 years in use, several small local cracks have been detected on the welding parts of tubular joints.

Although the platform is about to reach its original design life, there are still oil produced from the well heads. Accordingly, the platform needs to be extended its service life and the safety assessment of the jacket structure is necessary.

![Figure 1. The target jacket platform in Bohai Bay, China.](image)

3. Evaluation Index Systems

The construction and quantification of evaluation indicators are the basis and the key of safety assessment of the jacket platform, which directly affect the evaluation results. The determination of evaluation indicators must be justified scientifically and rationally. Multiple risk factors in ocean environments, structure status, and HOFs are considered in this section.

3.1. Ocean Environments
The platform is threatened by complicated ocean environments during its service life. The main ocean environment factors are considered as follows.

3.1.1. Environmental Loads

Environmental loads are loads imposed on the platform by natural phenomena including wind, wave, current and ice. Environmental loads also include the variation in hydrostatic pressure and buoyancy on members caused by changes in the water level due to waves and tides. Based on 50-year-return-period standard in the target sea, the average wind speed of 10 minutes is 28m/s, the average wind speed of one minute is 30.5m/s, and the gust wind speed of 3 seconds is 33.6m/s. The maximum wave height is 10.2m, and the corresponding wave period is 10.3s. The maximum current speed is 1.69m/s. The maximum ice thickness is 0.4m and the ultimate compressive strength is 2.24MPa.

3.1.2. Earthquakes

Seismic effects should be considered in platform assessment for areas that are determined to be seismically active. Earthquake activity may trigger soil liquefaction and submarine slides, which causes instability of the subsurface soils at the platform site and may result in collapse of the platform. The offshore platform in Bohai Bay is located in the intersection of Tan-Lu earthquake zone and Zhangjiakou-Bohai earthquake zone. There has been 4 recorded earthquakes of M >7.0 in 1548, 1597, 1888 and 1969 in Bohai Bay, respectively.

3.1.3. Scour

Scour is removal of seafloor soils caused by currents and waves. Such erosion can be a natural geologic process or can be caused by structural elements interrupting the natural flow regime near the seafloor. Scour can result in removal of vertical and lateral support for foundations, causing undesirable settlements of mat foundations and overstressing of foundation elements. The scour amount of the platform is about 1.2m through the observation.
3.2. Structure Status

Structure status of the platform reflects the circumstance of service and the condition of health, including the dead loads and live loads of the jacket platform, some types of damage during its service life.

3.2.1. Dead Loads and Live Loads

Dead loads are the weights of the platform structure and any permanent equipment and appurtenant structures which do not change with the mode of operation. The total mass of the platform is about 1216800 kg.

Live loads are the loads imposed on the platform during its use and which may change either during a mode of operation or from one mode of operation to another. Since the oil is transferred by the submarine pipelines, the main live loads of the platform include the weight of living quarters and other life support equipment, and the forces exerted on the structure from vessel mooring.

3.2.2. Corrosion

Corrosion may seriously affect the structural capacity of systems due to the thickness reduction of the structural steel. After some time of increasing corrosion, the tensile stress can reach a critical value and can cause corrosion cracks [20, 21], and the bearing capacity of structures will also be reduced. Design features to prevent or reduce the effect of corrosion including the use of coating, cathodic protection and/or a plate thickness allowance. However, because these methods are not fail-safe, corrosion and its negative effects on ultimate strength and fatigue resistance are to be considered during operation.

Studies have shown that the general corrosion rate for steel in seawater is about 0.1 mm/year. However, corrosion rates fluctuate between 0.04 to 1.2 mm/year and in some cases corrosion rates up to 3 mm/year have been found in some ballast tanks in tankers [22]. The corrosion rate exhibits
a very large scatter depending upon location in the structure. The sea environment, i.e. in Bohai Bay as compared to West Africa or Gulf of Mexico, is also a factor of influence on corrosion rates. However this aspect has not yet been quantified. The corrosion is not homogeneous and the maximum corrosion amount of the platform is 2.6mm through the observation.

3.2.3. Fatigue

Fatigue is one of the most common damage types that may occur when structures are subjected to repetitive or cyclic load patterns. Fatigue damage begins with cracks that initiate in structural elements, and may continue with their growth and with initiation of other cracks [23, 24].

Fatigue strength is commonly described by SN-curves that have been obtained by laboratory experiments. Fracture mechanics methods have been adopted to assess the different stages of crack growth, including calculation of residual fatigue life. However, a significant amount of cracks inspections in North Sea fixed platforms suggested that the prediction methods were conservative and that the likelihood of fatigue cracks was much less than initially anticipated. This fact indicates that it is difficult to accurately assess the fatigue cracks growth status [1].

Due to lack of access inside the underwater jacket structure, inspections have been carried out on the outside by divers or by remotely operated vehicles. This fact implies in general high inspection costs. According to the observation of the jacket platform, several small local cracks have been detected on the welding parts of tubular joints.

3.3. Human and Organizational Factors

Surveys conducted by different individuals and safety bodies have revealed that about 75-96% of marine casualties are fully/partially caused by human and organizational errors [25]. Studies [26-28] have shown that HOFs contribute to 75% of fires and explosions, 89-96% of collisions, 75% of allisions, 79% of towing vessel groundings and 84 – 88% of tanker accidents.

From human behavior and organization theory point of view, some researchers investigated human and organizational error classification and grouped them into four basic categories: skill
based errors, decision based errors, perceptual errors, and violation errors [29]. Each basic group could include some factors. Significant factors associated with skill-based errors, for example, may include failed to prioritize attention, inadvertent use of system controls, omitted step in procedure, omitted checklist item, poor technique, and over controlled the system. Decision based errors include improper procedure, misdiagnosed emergency, wrong response to emergency, exceeded ability, inappropriate maneuver, and poor decision. Perceptual errors happen due to misjudged distance and visual illusion. Violation errors may include violated training rules, failed to properly prepare for the mission, not current/qualified for the mission, and intentionally exceeded the limits of the vessel [13].

Fire, blast, and accidental loading events (e.g. collision from ships, dropped objects) are the main accidents caused by HOFs, which could lead to partial or total collapse of an offshore platform resulting in loss of life and/or environmental pollution. For the target production platform in Bohai Bay, there are some oil and gas treatment and measuring equipments, so the fire and blast risks should be considered in the offshore platform safety assessment. Although the platform is unattended most of the time, some periodical inspections and supplies are necessary. Hence the accidental loading events, for instance, collision from ships are also supposed to be considered.

4. Method and Application

4.1. Analytic Hierarchy Process

The analytic hierarchy process (AHP) is a structured technique for organizing and analyzing complex decisions, based on mathematics and psychology. The procedure for using the AHP can be briefly summarized below [30].

1) Construct a hierarchy containing the target layer and the indicators layer, and the criteria for evaluating the indicators.

2) Establish the judgment matrix based on pairwise comparisons of the indicators.

3) Calculate the weight vectors and check the consistency.

This paper constructs the evaluation index system of safety assessment of aging offshore jacket platform in Bohai Bay, China that includes three levels, namely, the target layer, the
first-grade indicators layer and the second-grade indicators layer. The target layer consists of three first-grade indicators: ocean environments, structure status, and HOFs. Each first-grade indicator is divided into three second-grade indicators. There are nine second-grade indicators in total analyzed in this paper. Ten evaluators are selected to give each second-grade indicator a grade, and then the pairwise comparison matrix is established. The weights of evaluating indicators are determined by computing the greatest eigenvalue and eigenvector of the matrix. The consistency ratio of a pairwise comparison matrix is the ratio of its consistency index to the corresponding random index. The reader is referred to the subject textbook [30] for a detailed treatise on AHP.

4.2. Grey Clustering Method

Grey clustering method adopts the grey whitening functions in common clustering analysis, which is mainly applied to study the uncertainties of system model and make forecasts and decisions. The meaning of grey can be expressed as the characteristics between black and white. The grey system focuses deeply on what partial or limited information the system can provide, and tries to describe its total feature from this [31]. The grey statistic method is proposed and applied to safety assessment of the jacket platform in Bohai Bay, China in this section. The basic definition and approach are described in a step-by-step manner as following.

4.2.1. Grey Whitening Function

Assume \( f(x) \) is a linear monotonic function of \( x \) and \( f(x) \in [0,1] \). Then \( f(x) \) is called the whitening weight function and \( x \) is the grey number. If there are \( g \) evaluation grey cluster, the corresponding whitening function and threshold value is named \( f_1(x), f_2(x), \ldots, f_g(x) \) and \( \lambda_1, \lambda_2, \ldots, \lambda_g \), respectively. The evaluation result depends on the assessment standard which is divided into five grades in this paper: “very high”, “high”, “medium”, “low”, “very low”, \( g=5 \). These qualitative indexes are assigned values accordingly in order to transfer them into quantitative indexes. The score of each grade is 9, 7, 5, 3 and 1, respectively, i.e.,
\[ \{\lambda_1, \lambda_2, \lambda_3, \lambda_4, \lambda_5\} = [9, 7, 5, 3, 1] \]. Their whitening functions are expressed as

\[
\begin{align*}
f_1(x) &= \begin{cases} 
\frac{1}{9}x, & 0 \leq x < 9 \\ 
1, & 9 \leq x \leq 10
\end{cases} 
\tag{1}
\end{align*}
\]

\[
\begin{align*}
f_2(x) &= \begin{cases} 
\frac{1}{7}x, & 0 \leq x < 7 \\ 
\frac{1}{3}(10 - x), & 7 \leq x \leq 10
\end{cases} 
\tag{2}
\end{align*}
\]

\[
\begin{align*}
f_3(x) &= \begin{cases} 
\frac{1}{5}x, & 0 \leq x < 5 \\ 
\frac{1}{5}(10 - x), & 5 \leq x \leq 10
\end{cases} 
\tag{3}
\end{align*}
\]

\[
\begin{align*}
f_4(x) &= \begin{cases} 
\frac{1}{3}x, & 0 \leq x < 3 \\ 
\frac{1}{7}(10 - x), & 3 \leq x \leq 10
\end{cases} 
\tag{4}
\end{align*}
\]

\[
\begin{align*}
f_5(x) &= \begin{cases} 
1, & 0 \leq x < 1 \\ 
\frac{1}{9}(10 - x), & 1 \leq x \leq 10
\end{cases} 
\tag{5}
\end{align*}
\]

4.2.2. Grey Evaluation Weight

The score of each evaluator is considered as a grey number in grey evaluation model. Assuming \(x\) evaluators give marks \(d_{i1}, d_{i2}, \ldots, d_{ix}\) to evaluation indicator \(U_{ij}\), then the whitening function which evaluators consider \(U_{ij}\) to belong to the \(l\)th grey cluster is \(f_l(d_{i1}), f_l(d_{i2}), \ldots, f_l(d_{ix})\) respectively. The total quantity of whitening function which belongs to the \(l\)th evaluation grey cluster is \(\sum_{k=1}^{x} f_l(d_{ik})\), and then the total quantity of whitening function of the entire evaluation grey cluster is \(\sum_{l=1}^{g} \sum_{k=1}^{x} f_l(d_{ik})\). The grey evaluation weight of the \(l\)th evaluation grey cluster is defined as

\[
\begin{align*}
r_{il} &= \frac{\sum_{k=1}^{x} f_l(d_{ik})}{\sum_{l=1}^{g} \sum_{k=1}^{x} f_l(d_{ik})}, & l = 1, 2, \ldots, g
\end{align*}
\tag{6}
\]

The grey evaluation weight vector of all evaluation grey clusters is

\[
\begin{align*}
r_{ij} &= [r_{ij1}, r_{ij2}, \ldots, r_{ijn}], & i = 1, 2, \ldots, m, & j = 1, 2, \ldots, n
\end{align*}
\tag{7}
\]
The above computational processes can be accomplished by table dispatching method which is listed in Table 1.

Table 1. The table dispatching method calculating the grey evaluation weight

<table>
<thead>
<tr>
<th>Grey cluster</th>
<th>(d_{i1})</th>
<th>(d_{i2})</th>
<th>...</th>
<th>(d_{in})</th>
<th>sum</th>
<th>Evaluation weight</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>(f_1(d_{i1}))</td>
<td>(f_1(d_{i2}))</td>
<td>...</td>
<td>(f_1(d_{in}))</td>
<td>(\sum_1)</td>
<td>(\sum_1/\sum)</td>
</tr>
<tr>
<td>2</td>
<td>(f_2(d_{i1}))</td>
<td>(f_2(d_{i2}))</td>
<td>...</td>
<td>(f_2(d_{in}))</td>
<td>(\sum_2)</td>
<td>(\sum_2/\sum)</td>
</tr>
<tr>
<td>...</td>
<td>...</td>
<td>...</td>
<td>...</td>
<td>...</td>
<td>...</td>
<td>...</td>
</tr>
<tr>
<td>g</td>
<td>(f_g(d_{i1}))</td>
<td>(f_g(d_{i2}))</td>
<td>...</td>
<td>(f_g(d_{in}))</td>
<td>(\sum_g)</td>
<td>(\sum_g/\sum)</td>
</tr>
<tr>
<td>sum</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>(\sum)</td>
<td></td>
</tr>
</tbody>
</table>

* Note that \(r_{ij} = \frac{\sum_i}{\sum}\), where \(\sum_i = \sum_k f_i(d_{ik}), l = 1, 2, \ldots, g, \sum = \sum_1 + \sum_2 + \cdots + \sum_g\).

4.2.3. Grey Assessment Value

Assume that the first-grade indicator objective \(U_i\) is composed of \(U_{i1}, U_{i2}, \ldots, U_{in}\), the weight vectors of \(U_i\) are written as \(A_i = [a_{i1}, a_{i2}, \ldots, a_{in}]\). Note that \(a_{ij}\) is calculated by AHP and should have to meet specific conditions, i.e., \(a_{ij} \geq 0\) and \(\sum_{j=1}^{n} a_{ij} = 1\). Then the grey comprehensive assessment weight vector of \(U_i\) is defined as

\[
B_i = A_i \times R_i = [b_{i1}, b_{i2}, \ldots, b_{ig}],
\]

where the grey evaluation weight matrix \(R_i\) is written as

\[
R_i = \begin{bmatrix}
  r_{i1} \\
  r_{i2} \\
  \vdots \\
  r_{im}
\end{bmatrix} =
\begin{bmatrix}
  r_{i11} & r_{i12} & \cdots & r_{i1g} \\
  r_{i21} & r_{i22} & \cdots & r_{i2g} \\
  \vdots & \vdots & \ddots & \vdots \\
  r_{im1} & r_{im2} & \cdots & r_{img}
\end{bmatrix}, \quad i = 1, 2, \ldots, m
\]
To target layer $U$, consisting of all evaluation indicators $U_1, U_2, \ldots, U_m$, the weight vectors are written as $A = [a_1, a_2, \ldots, a_m]$. Note that $a_i \geq 0$ and $\sum_{i=1}^{m} a_i = 1$. Then the grey comprehensive assessment weight vector of $U$ is defined as

$$B = A \times R = [b_1, b_2, \ldots, b_g],$$

(10)

where the grey evaluation weight matrix $R$ is comprised of $B_i$, written as

$$R = \begin{bmatrix}
B_1 \\
B_2 \\
\vdots \\
B_m
\end{bmatrix}, \quad i = 1, 2, \ldots, m.$$  

(11)

Usually the grey grade can be determined according to the maximum principle. However, sometimes the result will be distorted because of losing too much information. The assessment weight vector $B$ will be dealt with by making it single-value in this paper. The grey comprehensive assessment value will be calculated as follow

$$W = B \times \begin{bmatrix}
\lambda_1 \\
\lambda_2 \\
\vdots \\
\lambda_g
\end{bmatrix}^T = b_1 \lambda_1 + b_2 \lambda_2 + \ldots + b_g \lambda_g$$

(12)

Taking the grey assessment value $W$ as the grey number, then the whitening weight function of $W$ is denoted as $f_1(W), f_2(W), \ldots, f_g(W)$. The grey grade is determined according to the maximum principle.

4.3. Application

According to the principles of the AHP, the pairwise comparison matrix and the weights are calculated based on the overall consideration and statistical analysis of the multiple risk factors of the aging jacket platform. The weight vectors of first-grade evaluation indicators affecting safety of the platform are written as $A = [0.2667, 0.4011, 0.3322]$. The consistency ratio is 0.005, which indicates that the weighted coefficients are reasonable and efficient. The importance ranking for the first-grade evaluation indicator is structure status $>$ HOFs $>$ ocean environments. So the structure status is the most noticeable factor in the safety assessment of the jacket platform. The
The grey evaluation weight vectors of second-grade evaluation indicators are written as $A_1 = [0.2047, 0.4370, 0.3583]$, $A_2 = [0.2720, 0.3584, 0.3696]$, $A_3 = [0.3747, 0.3337, 0.2816]$ for ocean environments, structure status, and HOFs, respectively. The consistency ratios are 0.006, 0.005 and 0.006, respectively. The importance ranking for the second-grade evaluation indicator are earthquakes > scour > environmental loads, fatigue > corrosion > dead loads and live loads, and fire > blast > accidental loading events, respectively.

The safety assessment of the jacket platform has the characteristics of grey. The grey clustering method is applied in this paper to evaluate the safety of the platform. The grey evaluation weight vectors of indicators are calculated by the table dispatching method. Take for example, the grey evaluation weights of environmental loads are given in Table 2.

<table>
<thead>
<tr>
<th>Grey cluster</th>
<th>Evaluation weight</th>
<th>2</th>
<th>3</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>1</th>
<th>3</th>
<th>2</th>
<th>Σ</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td></td>
<td>$f_1(2)$</td>
<td>$f_1(3)$</td>
<td>$f_1(1)$</td>
<td>$f_1(2)$</td>
<td>$f_1(3)$</td>
<td>$f_1(1)$</td>
<td>$f_1(3)$</td>
<td>$f_1(2)$</td>
<td>2.4444</td>
</tr>
<tr>
<td>2</td>
<td></td>
<td>$f_2(2)$</td>
<td>$f_2(3)$</td>
<td>$f_2(1)$</td>
<td>$f_2(2)$</td>
<td>$f_2(3)$</td>
<td>$f_2(1)$</td>
<td>$f_2(3)$</td>
<td>$f_2(2)$</td>
<td>3.1429</td>
</tr>
<tr>
<td>3</td>
<td></td>
<td>$f_3(2)$</td>
<td>$f_3(3)$</td>
<td>$f_3(1)$</td>
<td>$f_3(2)$</td>
<td>$f_3(3)$</td>
<td>$f_3(1)$</td>
<td>$f_3(3)$</td>
<td>$f_3(2)$</td>
<td>4.4000</td>
</tr>
<tr>
<td>4</td>
<td></td>
<td>$f_4(2)$</td>
<td>$f_4(3)$</td>
<td>$f_4(1)$</td>
<td>$f_4(2)$</td>
<td>$f_4(3)$</td>
<td>$f_4(1)$</td>
<td>$f_4(3)$</td>
<td>$f_4(2)$</td>
<td>7.3333</td>
</tr>
<tr>
<td>5</td>
<td></td>
<td>$f_5(2)$</td>
<td>$f_5(3)$</td>
<td>$f_5(1)$</td>
<td>$f_5(2)$</td>
<td>$f_5(3)$</td>
<td>$f_5(1)$</td>
<td>$f_5(3)$</td>
<td>$f_5(2)$</td>
<td>8.6667</td>
</tr>
<tr>
<td>Σ</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>25.9873</td>
</tr>
</tbody>
</table>
Then the grey evaluation weight matrix of $U_i$ can be written as

$$r_{31} = [0.3608, 0.2706, 0.1624, 0.1160, 0.0902]$$

$$r_{32} = [0.3089, 0.2926, 0.1756, 0.1254, 0.0975]$$

$$r_{33} = [0.2362, 0.2885, 0.2094, 0.1496, 0.1163]$$

The grey comprehensive assessment weight vector of $U_i$ is

$$B_1 = A_i \times R_1 = [0.1270, 0.1632, 0.2185, 0.2499, 0.2415]$$

$$B_2 = A_i \times R_2 = [0.3307, 0.2240, 0.1680, 0.1560, 0.1214]$$

$$B_3 = A_i \times R_3 = [0.3048, 0.2803, 0.1784, 0.1274, 0.0991]$$

Accordingly, the grey evaluation weight matrix of $U$ can be written as

$$R = \begin{bmatrix} B_1 \\ B_2 \\ B_3 \end{bmatrix} = \begin{bmatrix} 0.1270 & 0.1632 & 0.2185 & 0.2499 & 0.2415 \\ 0.3307 & 0.2240 & 0.1680 & 0.1560 & 0.1214 \\ 0.3048 & 0.2803 & 0.1784 & 0.1274 & 0.0991 \end{bmatrix}$$

and the grey comprehensive assessment weight vector of $U$ is

$$B = A \times R = [0.2678, 0.2265, 0.1849, 0.1715, 0.1460]$$

Finally, the grey comprehensive assessment value is calculated as follow

$$W = B \times [\lambda_1, \lambda_2, \ldots, \lambda_5]^T = 5.5807.$$
\[ f_1(W) = 0.6201, f_2(W) = 0.7972, f_3(W) = 0.8839, f_4(W) = 0.6313, f_5(W) = 0.4910. \] (19)

Obviously, \( f_3(W) \) is the maximum value and it can be concluded that the grey grade of the failure risk of the platform is “medium”. That means the failure of the aging offshore jacket platform is occurred possibly which is consistent with the test report from a third party. Regular maintenance or periodically detect for the platform and facilities should be carried out in order to ensure safety and extend the service life.

5. Conclusions

1. The safety evaluation index system of an aging offshore jacket platform in Bohai Bay, China is proposed, which consists of three levels and nine risk factors in ocean environments, structure status, and HOFs, respectively.

2. AHP method is adopted to weigh the risk factors and grey clustering method is applied to estimate the failure risk of the jacket platform qualitatively and quantitatively. The evaluation results suggest that the grade of the risk is medium and the failure is possible to occur which is in agreement with the test report from a third party. It indicates this evaluation method is objective and efficient.

3. The impact factors of the safety of the aging platform in Bohai Bay are comprehensive; especially structure status and HOFs will lead to direct or potential hazards. In order to extend the service life, regular maintenance or periodically detect for the platform and facilities should be implemented.

Acknowledgments: The research works are supported by the National Natural Science Foundation of China (No. 51309239), Fundamental Research Funds for the Central Universities (Grant No. 16CX02057A), National Key Research and Development Plan (No. 2016YFC0303800), High School Science and Technology Project of Shandong Province (No. J15LB55), and Science and Technology Project of Qingdao Huanghai University (No. 2015DXKJ19).

Author Contributions: Bao, X.X. and Cao, A.X. analyzed the data; Bao, X.X. wrote the paper.
and Qin, F.Z. checked the whole paper.

Conflicts of Interest: The authors declare no conflict of interest.

References


© 2016 by the authors; licensee *Preprints*, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons by Attribution (CC-BY) license (http://creativecommons.org/licenses/by/4.0/).