PREDICTION OF CORRODING PIPELINE REMAINING LIFE-TIME USING SEMI-PROBABILISTIC APPROACH

Norhazilan Md Noor ¹, Nordin Yahaya ², Mazura Mat Din³, Nor Apaziah Nor Ozman⁴

^{1,2,4}Department of Structures and Materials, Faculty of Civil Engineering, Universiti Teknologi Malaysia, 81310 Skudai, Johor

³Faculty of Computer Science and Information System, Universiti Teknologi Malaysia, 81310 Skudai, Johor

Abstract: A semi-probabilistic methodology for predicting the remaining strength of submarine pipelines subjected to internal corrosion based on Recommended Practice RP-F101 by Det Norske Veritas (DNV) is described in this paper. It is used to estimate the maximum allowable operating pressure of the corroding pipelines based on series of pigging data, which represents the corrosion pit location and dimension. The introduction of partial safety factors in the DNV code to minimise the effect of uncertainties due to the defect sizing has improved the reliability of pipeline assessment methodology. Nevertheless, the code is still regarded as a fully deterministic approach due to its incapability of predicting the remaining life of corroded pipeline. Thus, we have added prediction capabilities to the capacity equation by introducing a standard deviation model of future defect depth. By doing so, the variation of safety factors of the capacity equation can be fully manipulated where prediction of future pipeline remaining life-time becomes feasible. The paper demonstrates calculation and prediction of pipeline remaining lifetime subjects to internal corrosion. The results shows the standard deviation of corrosion parameter affected the value of partial safety factor as corrosion progressing, hence amplify the conservatism of time to failure. In general, the prediction of pipeline remaining lifetime can effectively assist pipeline operators to evaluate future safe operating strategies including re-inspection and appropriate maintenance schedule. As a result it can minimize the possibility of pipeline failures until it reaches its designed lifetime.

Keywords: Pipeline, DNV RP-F101, corrosion, probabilistic.

1.0 Corrosion

Offshore (marine) and onshore pipelines are one of the safest, economical and as a consequence, the most applied means of transporting oil and gas in the world nowadays. Unfortunately, the increasing number of aging pipelines in operation has significantly increased the number of accidents (Teixera *et al.*, 2008). As pipeline ages, it can be affected by a range of corrosion mechanisms, which may lead to a reduction in its structural integrity and eventual failure (Ahammed, 1997; Netto *et al.*, 2005; Teixera *et al.*, 2008). Corrosion is an imperative form of pipeline deterioration due to aggressive environments (Ahammed and Melchers., 1996). Without practical and effectual corrosion prevention strategy, corrosion will continue to progress and the cost of repairing a deteriorating pipeline will escalate. Significant savings are possible by optimizing the inspection and corrosion prevention strategies (Ainouche, 2006).

2.0 Pipeline Pigging

In line inspection (ILI) tools, also commonly called pipeline inspection gauge or 'pig' are device used by the pipeline industry to survey the condition of the pipeline wall. Intelligent Pig is widely deployed to detect the corrosion defects in pipeline using high resolution magnetic (MFL) or ultrasonic mechanism (UT) to locate and measure the size of a corrosion defect. The past 40 years has seen the development of several methods for assessing the significance of defects. Some of this has been incorporated into industry guidance (Cosham *et al.*, 2007). Metals loss (corrosion) tools are used to detect defects that have resulted in wall thinning in a pipeline. Whereas Magnetic Flux Leakage (MFL) provides a versatile and reliable method for determining the geometry of metal loss in pipelines, UT allows direct and highly accurate measurements of pipeline wall thickness. In our study, the data was collected using MFL tools.

3.0 Research Problem and Methodology

Since corrosion is a complex process involving numerous unknown factors, the prediction task has always been a challenge to pipeline owners especially when vital information is lacking of. The inherent uncertainties embedded within metal loss data plays significant roles in reducing the accuracy of pipeline future assessment. These uncertainties are related to tool imperfect measurement, randomness of environment and variation of operational data. To cater the

uncertainties, DNV RP-F101 has incorporated safety factors which specially tailored for defect depth. Yet, the code is still regarded as a deterministic approach akin to other capacity equation such as ASME B31G and PCORRC since the safety factors represent averaged value and no variation of parameters included in the calculation of pipeline remaining pressure. Unlike conventional safety factors, the value is dependent upon inspection tool accuracy which is defined by the dispersion of corrosion growth rate value and metal loss data. Nevertheless, the code is designed to assess the condition of the line at the time of inspection owing to constant partial safety factor. In order to predict the remaining pressure in the future, the amount of unknown uncertainties which theoretically escalate with time must be taken into account. This is reflected by the increment of partial safety factors as a function of time to represent the influence by these unknown uncertainties related to randomness of corrosion progress, environment as well as material properties (Yahaya et al., 2009). The originality of this paper relies on the introduction of statistical-based equation which able to estimate the new standard deviation value of possible future defect depth. In our works, we manipulate the variation of safety factors in RP-F101 to make the deterministic capacity equation capable of predicting the future growth of defects. The future metal loss data based on prediction is supposed to pose higher variation of its value compared to actual metal loss data. Hence, higher safety factors of defect depth will increase the conservatism of assessment due to rapid reduction of structure capacity which is more realistic.

3.1 Pipeline Inspection Data

In our case study, an extensive amount of pigging data has been gathered through repeated in-line inspection activities using MFL intelligent pig on the same pipelines at different point of times. The transmission pipelines located in North Sea area used to convey crude oil and gas (multiphase line) from central offshore platform to onshore terminal. The data provides valuable information on the internal corrosion defect geometry, such as defect location, depth and length, orientation and types of corrosion regions as displayed in Table 1.

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Spool Length (m)	Relative distance (m)	Absolute distance (m)	d% wt	l (mm)	W (mm)	O'clock	t (mm)	Loc.
11.6	6.6	1016.5	18	32	42	6.00	14.2	Internal
11.5	11.5	1033.0	19	46	64	5.30	14.2	Internal
11.8	10.6	1043.6	12	18	55	5.30	14.2	Internal
11.7	1	1045.8	13	28	83	5.30	14.2	Internal

Description:

Absolute distance : Distance of corrosion from start of pipeline

d%wt : Maximum depth of corrosion in terms of percentage

: Longitudinal extent of corrosion

Lo : Location of corrosion either internal or external.

O'Clock : Orientation of corrosion as a clock position of pipe wall thickness.

Relative distance : Relative distance of corrosion from upstream girth

Spool length : Length of pipe between weld (10m to 12m approximately)

 t_t : Nominal thickness of pipe in pipe spool

W: Extent of corrosion around pipe circumference weld

3.2 Assessment of Corroding Pipeline

In general, the degree of conservatism in regard to structural assessment is due to the implementation of safety factor in deterministic method. This safety factor is associated with load factor or resistance factor (strength of material), which is commonly found in all codes of assessment for corroded pipelines. The uncertainties subjected to structural properties, loading condition, environmental behavior and construction performance are always neglected in the calculation due to the employment of safety factor. In pipeline assessment, deterministic assessment is a straight-forward approach based on codes or developed capacity equation. Generally, the deterministic methods use lower bound data; for instance peak depth of corrosion, maximum corrosion rate and minimum wall thickness without considering the existing uncertainties (Yahaya, 2000). Consequently, it can be over conservative in terms of safety when being implemented to pipelines containing extensive corrosion defects. For example, the prediction of future growth of corrosion defects located in the pipelines will use an average for single rate value without considering the possibility that not all defects will grow at the same rate. The averaged rate is used for the sake of simplicity owing to lack of information pertaining to environmental and material properties.

Assessment method is required to determine the severity of such defects when they are detected in pipelines (Cosham and Hopkins., 2003). The assessment of the condition of existing oil and gas pipeline is necessary in order to protect the public, financial investment and environment from such failures. Systematic and optimized regular inspections of pipelines with state-of-the art tools and procedures can reduce significantly the risk of any undue accident caused by a lack of unawareness of the integrity of the line (Cosham *et al.*, 2007).

3.3 DNV Recommended Practice (RP-F101)

The DNV Recommended Practice for the assessment of corroded pipelines, DNV-RP-F101 was issued in 1999. RP-F101 describes two alternative approaches with different safety philosophy. The equations in RP-F101 were derived by a probabilistic calibration (Bjornoy *et al.*, 2001), taking into account the uncertainties in defect measurements and burst capacity. The equations account directly for the accuracy in sizing the corrosion defect.

3.3.1 RP- F101 Criteria

The RP –F101 recommends the assessment of corroded pipelines subject to internal pressure and internal pressure combined with longitudinal compressive stresses (Bjornoy *et al.*, 2001). Moreover, this new criterion provides an assessment procedure for single defect, interacting defects and complex shaped defects.

3.3.2 Capacity Equation

The maximum allowed operation pressure in pipelines for a single defect is given as:

$$P_{p} = \frac{\gamma_{m} \times 2 \times t \times SMTS \times (1 - \gamma_{d} (d/t)^{*})}{(D - t)(1 - \gamma_{d} (d/t)^{*} Q^{-1})} \le P_{mao}$$
(1)

where;

$$Q = \sqrt{1 + 0.31 \left(\frac{L}{\sqrt{Dt}}\right)^2} \tag{2}$$

$$(d/t)^* = (d/t)_{meas} + \varepsilon_d StD[d/t]$$
(3)

and;

D = outer diameter

d = depth of corrosion defect t = nominal pipe wall thickness

L = measured length of corrosion defect $(d/t)_{meas}$ = measured relative corrosion depth

partial safety factor for prediction model and safety class $\gamma_{\rm m}$ partial safety factor for corrosion depth = $\gamma_{\rm d}$ factor for defining a fractile value for the corrosion depth $\epsilon_{
m d}$ = P_{mao} maximum allowable operating pressure = standard deviation for measurement (d/t) ratio StD[d/t]= specified minimum tensile strength **SMTS** =

Fundamentally, Equation 1 is similar to ASME B31G. However, the difference between these two criteria is partial safety factors are included in RP-F101 equation to ensure a consistent reliability level for various combinations of material properties, pipe geometries and corrosion defects configurations.

3.3.3 Partial Safety Factors

The partial safety factors γ_m and γ_d , and the fractile value ϵ_d are determined from tables which depend on the safety class classification, the pipe quality, inspection method and sizing accuracy of the inspection tool (DNV, 2004). It was given as functions of the sizing accuracy of the measured defect depth for inspections based on relative depth measurements and for inspections based on absolute depth. The safety class is specified based on Table 2 to Table 4.

Table 2: Partial safety factor γ_m , (DNV, 2004)

Inspection Method	Safety class				
inspection Method	Low	Normal	High		
Relative (e.g. MFL)	$\gamma_m\!=0.79$	$\gamma_m\!=0.74$	$\gamma_m\!=0.70$		
Absolute (e.g. UT)	$\gamma_m\!=0.82$	$\gamma_m\!=0.77$	$\gamma_m\!=0.72$		

Table 3: Standard deviation, StD [d/t], for MFL inspection tool (DNV, 2004)

Relative sizing accuracy	Confidence level				
Relative Sizing accuracy	80%	90%			
Exact \pm (0.0 of t)	StD[d/t] = 0.00	StD[d/t] = 0.00			
± 0.05 of t	StD[d/t] = 0.04	StD[d/t] = 0.03			
± 0.10 of t	StD[d/t] = 0.08	StD[d/t] = 0.06			
± 0.20 of t	StD[d/t] = 0.16	StD[d/t] = 0.12			

Inspection	sizing	accuracy		Safety class		
StD[d/t]			ϵ_{d}	Low	Normal	High
(exact) 0.00			0.0	$\gamma_d = 1.00$	$\gamma_d = 1.00$	$\gamma_d = 1.00$
0.04			0.0	$\gamma_d=1.16$	$\gamma_d=1.16$	$\gamma_d=1.16$
0.08			1.0	$\gamma_d=1.20$	$\gamma_d=1.28$	$\gamma_d=1.32$
0.16			2.0	$\gamma_d=1.20$	$\gamma_d=1.38$	$\gamma_d=1.58$

Table 4: Partial safety factor, γ_d and fractile value factor, ϵ_d (DNV, 2004)

4.0 Pipeline Remaining Lifetime

The steel pipeline has a 14.2 mm wall thickness with outside diameter given as 914.4 mm. The allowable defect size of the pipeline is indicated by the uppermost curve, i.e. the acceptance line (refer to Figures 1 to 5). Once the corrosion point exceeds the acceptance line, the pipeline is considered to be in a critical condition and inspection and repair are recommended to commence. The mean value and standard deviation of the corrosion rate used in this assessment are 0.0405 mm/year and 0.08 mm/year respectively. The acceptance line was constructed using Equation 1. From this equation, the maximum corrosion defect length was estimated by fixing the corrosion depth between 10%wt to 100%wt (percentage of defect depth against wall thickness) and with the different working pressure of 8MPa, 9MPa and 10MPa. By taking out the defect length parameter, an equation of length correction factor, Q can be written as:

$$Q = \frac{\gamma_d \left(\frac{d}{t}\right)^*}{\left[1 - \frac{\gamma_m 2tSMTS\left(1 - \gamma_d \left(\frac{d}{t}\right)^*\right)}{\Phi - t P_p}\right]}$$

$$(4)$$

Since Q can also be represented by Equation 2, therefore the maximum allowable defect length for a given defect depth and working pressure can be calculated as:

$$L_{\text{max}} = \sqrt{\frac{\mathbf{Q}^2 - 1}{0.31}} \times \sqrt{Dt}$$
 (5)

where:

 L_{max} = maximum allowable defect length

The Q expression in Equation 5 can be calculated by inserting Equation 4 into Equation 5 so the allowable line / acceptance line to evaluate pipeline condition can be constructed. The increment value of Std [d/t], γ_d and ε_d in the future can be estimated using Table 4 as given in the DNV RP-F101 code.

4.1 Semi-Probabilistic Equation

The next section of this paper exhibits the efforts of deriving a semi-probabilistic equation intended to calculate the future variation of corrosion defects. The term semi-probabilistic means that the proposed equation still requires averaged values (fixed value) of parameters instead of random values. Nonetheless, the estimated value is related to the probable variation of corrosion defects in the future. Moreover, the equation is derived based on the principal of variance of probability distribution. Hence, the terms semi-probabilistic was chosen to signify the probabilistic element within deterministic equation. The augmentation of prediction capabilities by increasing the standard deviation of predicted can be explained mathematically by referring to linear growth rate model. The corrosion rate equation can be written as:

$$CR = \frac{d_{t_{1+1}} - d_{t_i}}{T} \tag{6}$$

where, $T = t_{i+1} - t_i$ and is a constant value.

If corrosion depth d is assumed statistically to be varied, the variation of corrosion rate can be expressed as:

variance
$$\mathbb{C}R$$
 = variance $\left(\frac{d_{t_{l+1}} - d_{t_i}}{T}\right)$ (7)

Since the time interval, *T* is a single value with no variation, Equation 7 can be rewritten as:

$$\sigma_{CR}^2 = \frac{1}{T^2} . \text{variance } \left(d_{t_{1+1}} - d_{t_i} \right)$$
 (8)

and, simplified into:

$$\sigma_{CR}^2 = \frac{1}{T^2} \left\{ \left\{ \left[\left[\frac{T}{T_{t+1}} - \left[\left[\left[\frac{T}{T_{t+1}} \right] \right] \right] \right] \right\} \right\} \right\}$$
 (9)

Therefore, the relationship between inspection time interval and the variation in corrosion growth rate can be presented as:

$$\sigma_{CR} = \frac{1}{T} \sqrt{\left(\left(\left(\left(\frac{T}{L_{+}} \right) - \sigma \left(\left(\frac{T}{L_{+}} \right) \right) \right) \right)}$$

$$\tag{10}$$

Since $\sigma = Std$, therefore

$$Std_{CR} = \frac{1}{T} \sqrt{ \left(td \right) \left(\frac{2}{L_{+1}} - Std \right) \left(\frac{2}{L_{+}} \right)}$$

$$\tag{11}$$

Inspection data of metal loss from MFL pig tools usually is represented as a ratio of defect depth to wall thickness, d/t. By replacing the exact metal loss value, d with metal loss ratio, d/t, Equation 11 can be rewritten as follows:

$$Std\left[\frac{cr}{t}\right] = \frac{1}{T}\sqrt{\left(Std\left[\frac{d}{t}\right]_{t_{i+1}}^{2} - Std\left[\frac{d}{t}\right]_{t_{i}}^{2}\right)}$$
(12)

Equation 12 now can be reshuffled to make it as a standard deviation model of predicted depth. The new form of equation as a function of variation of defect from previous inspection, inspection time interval and variation of corrosion rate is as follows;

$$T.Std\left[\frac{cr}{t}\right] = \sqrt{Std\left[\frac{d}{t}\right]_{t_{i+1}}^{2} - Std\left[\frac{d}{t}\right]_{t_{i}}^{2}}$$
(13)

$$Std\left[\frac{d}{t}\right]_{t=1} = \sqrt{Std\left[\frac{d}{t}\right]_{t=1}^{2} + T^{2}.Std\left[\frac{cr}{t}\right]^{2}}$$
(14)

By assuming the wall thickness, t as a fixed value with no variation, the conclusive equation can be presented as

$$Std \left[V/t \right] = \sqrt{Std \left[V/t \right]^2 + \frac{T^2}{t^2} Std[cr]^2}$$
 (15)

where:

 $Std[d/t]_o$ = Standard deviation of inspection tool in first year

assessment.

 $Std[d/t]_T$ = Standard deviation of inspection tool in the future.

Std[cr] = Standard deviation of corrosion rate.

T = prediction interval in year

The equation depicts relationship between deviation of predicted data and the interval of prediction. The longer the prediction interval, the higher the variation of future metal loss, hence the higher the partial safety factors for metal loss. Table 5 shows the equations required to estimate the partial safety factors for metal loss and fractile value according to the range of metal loss standard deviation, Std[d/t].

Table 5: Polynomial equation for partial safety factor (defect depth) and fractile value [DNV, 2004]

Safety Factors	$\gamma_{d\ and}\ \epsilon_{d}$	Range
Low	$\gamma_d = 1.0 + 4.0 \text{ StD}[d/t]$	StD[d/t] < 0.04
	$\gamma_d = 1 + 5.5 \text{ StD}[d/t] - 37.5 \text{ StD}[d/t]^2$	$0.04 \le \text{StD}[d/t] < 0.08$
	$\gamma_d = 1.2$	$0.08 \le \text{StD}[d/t] \le 0.16$
Normal	$\gamma_d = 1 + 4.6 \text{ StD}[d/t] - 13.9 \text{ StD}[d/t]^2$	$StD[d/t] \le 0.16$
High	$\gamma_d = 1 + 4.3 \text{ StD}[d/t] - 4.1 \text{ StD}[d/t]^2$	$StD[d/t] \le 0.16$
(all)	$arepsilon_d = 0$	$StD[d/t] \le 0.04$
	$\varepsilon_d = -1.33 + 37.5 \text{ StD}[d/t] - 104.2 \text{ StD}[d/t]^2$	$0.04 \le \text{StD}[d/t] \le 0.16$

5.0 Results and Discussion

Pipeline time to failure was determined using the DNV RP-F101 capacity equation (Part A), also known as a semi-probabilistic assessment (DNV, 2004). Figures 1 to 5 show the prediction result of pipeline assessment subjects to internal corrosion from year t_0 to t_{10} . These predictions were based on gathered

corrosion data from pigging inspection done in year t_0 . From the assessment result in year t_0 and t_2 , the measured corrosion defect is within the acceptance criteria for all operating pressures where there are no defects exceeding the acceptance line as shown in Figures 1 and 2. Further prediction was carried out from year t_5 until t_{10} . As can be seen in Figures 3 to 5, the corrosion depth begins to exceed the acceptance line starting from year t_5 when the pipeline is operated under 9MPa and 10MPa of operating pressure. The pipeline is considered fail or in critical condition due to bursting in year t_5 when the acceptance criteria for all operating pressure were exceeded by the projected defects. Based on the result, it can be concluded that the pipeline should be inspected no later than year t_5 for every condition of operating pressure. The acceptance line predicted by the corrosion defect in year t_5 until t_{10} was found to be lower than the one estimated for the t_0 and t_2 prediction. This is due to the increment of uncertainties related to the averaged corrosion growth rate. Figures 6 to 8 show the increment of StD[d/t], fractile value, ε_d and safety factor, γ_d over the year, t_n . Equation 15 and Table 4 were used to recalculate the abovementioned values as corrosion progress in time. Hence, reduce the maximum allowable defect length and depth.

6.0 Conclusion

The deterministic approach has the distinct advantage of simplicity that capable to be applied in an entire pipeline or collection of pipelines easily. The disadvantage of the deterministic approach may often, but not entirely be linked to inaccuracies in the input data, but notably the inability to deal with uncertainties in the input data. The uncertainties of the deterministic approach are caused mainly by the averaged value of each parameter in the calculation i.e. lower bound of data. In the deterministic assessment the population of the corrosion dimension was assumed to grow at the same rate. A semi-probabilistic theory was introduced in the DNV RP-F101 code by estimating the standard deviation of inspection tool error and defect sizing. Nevertheless, the uncertainties in theory still inherently exist owing to a single or averaged value of parameters used in the calculation. These averaged or single values are unable to eliminate the uncertainties that might have occurred and increased over the years of service. Therefore, the prediction of pipeline integrity by using a deterministic assessment cannot fulfill the cost saving requirement by the operators. This leads to the condition whereby the operators have to inspect their pipelines frequently in order to obtain accurate information on pipeline condition. However, deterministic assessment is still widely used, but only to assess the current condition of the pipelines, owing to the lack of its prediction capabilities. The introduction of partial safety factors to minimise the effect of uncertainties due to the defect sizing and standard deviation model of future metal loss has improved the capability of predicting the future growth of corrosion defects deterministically. To improve prediction capability within deterministic framework, the inclusion of semi-probabilistic equation of future defect depth variation can be perceived as the right tool in reducing the stagnant state of structure resistance. In fact, structure assessment must take into account that when pipeline is aging, so does the structure resistance or capacity.

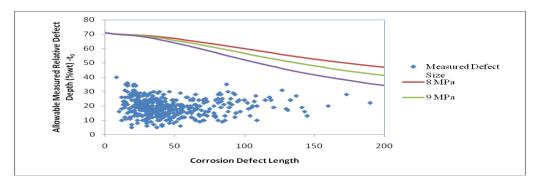


Figure 1: Projection of corrosion depth in year t₀ using DNV RP-F101 Assessment code

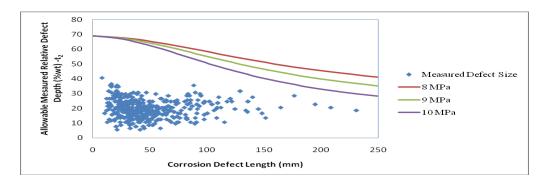


Figure 2: Projection of corrosion depth in year t₂ using DNV RP-F101 Assessment code

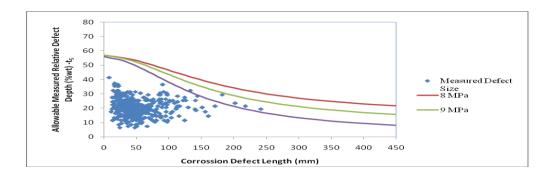


Figure 3: Projection of corrosion depth in year t₅ using DNV RP-F101 Assessment code

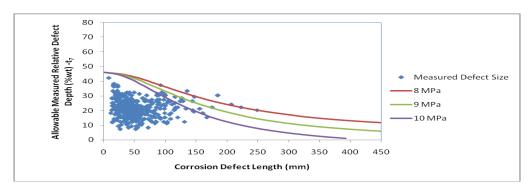


Figure 4: Projection of corrosion depth in year t₇ using DNV RP-F101 Assessment code

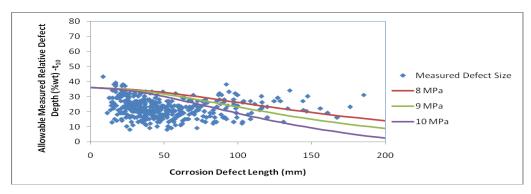


Figure 5: Projection of corrosion depth in year t_{10} using DNV RP-F101 Assessment code

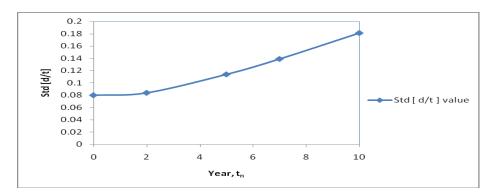


Figure 6: The increment of StD[d/t] over time

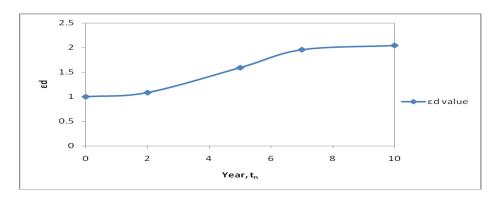


Figure 7: The increment of fractile value, ε_d over time

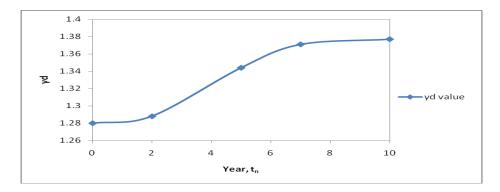


Figure 8: The increment of safety factor, γ_{d} over time

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