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Phoebe E. Fogelman

University of Tennessee - Knoxville, pfogelma@vols.utk.edu

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**Development of an Efficient Data Processing Procedure for the Prediction of Cleavage Fracture in
Reactor Pressure Vessel Steels Using the J-A₂ Method**

Phoebe E. Fogelman

College of Engineering

Department of Mechanical Engineering

Thesis Advisor: Dr. Larry Sharpe

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Abstract

Nuclear power plants have played an important role in decreasing the world's dependence on fossil fuels. As structures age, however, the hazards of continued operation must be evaluated against the cost of closure or refurbishment. The mechanism of failure for reactor pressure vessel steel is therefore of great concern. Because the competing ductile and brittle failure mechanisms result in a stochastic process, determination of critical values is computationally intensive. Finite element analysis is used to discretize the problem and simulate loading conditions to characterize material behavior. The $J-A_2$ method is a proposed improvement on the Hutchinson, Rice, and Rosengren solution to the failure prediction problem, which has a conservative bias. Because the $J-A_2$ method relies on the solution of a quadratic equation, however, the calculations are much more complicated. In order to continue validating this method, numerous experimental data sets will have to be compared to simulated results. With the former data structure and organization, this validation would be extraordinarily time-consuming, and delegating to research assistants would require extensive training and troubleshooting. The purpose of this project was therefore to develop a more automated and efficient method of processing data and demonstrate that resulting calculations are equivalent to those obtained by the original procedure. Furthermore, an additional data set is analyzed with the $J-A_2$ method, and computed critical values are compared with those experimentally determined at failure. The streamlined data processing procedure does, in fact, generate the same prediction as the previous method when applied to shallow-cracked specimens in 3-point bending. When used to analyze deep-cracked specimens, a curve fit is required to determine properties at the intersection with the material failure curve.

Table of Symbols and Abbreviations

Symbol	Parameter Represented	Units
A_2	See Equation 3	Unitless
a/W	Ratio of crack length to specimen thickness	mm/mm (unitless)
ϵ_0	Strain at yield stress	Unitless
ϵ	General strain	Unitless
HRR	Hutchinson-Rice-Rosengren	N/A
I_n	Integration constant	Unitless
J	Energy density	N/mm
J_c	Critical energy density	N/mm
L	Characteristic length	mm
n	Material constant	Unitless
r	Distance from crack tip	Mm
r_c	Critical distance from crack tip	Mm
r_n	Normalized distance from crack tip	Unitless
S_1	Material constant	Unitless
S_2	Material constant	Unitless
S_3	Material constant	Unitless
$\tilde{\sigma}_{ij}$	Stress at location (i,j)	MPa
σ_0	Yield stress	MPa
σ	General stress	MPa
3PB	3-point bending	N/A
$ x $	Absolute value	operator

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Introduction

Background

Although nuclear power generation as a percent of total energy generation in the United States peaked in the mid-1990s to early 2000s, 99 reactors remain in operation [1]. Furthermore, the European Union relies on nuclear reactors for approximately 27% of its total energy needs [2]. Nuclear power therefore plays an important role in the global effort to decrease dependence on fossil fuels. The future of nuclear power is uncertain, however, as disasters such as the Fukushima meltdown have strongly influenced public opinion and led to re-evaluations of nuclear power plant design within the scientific community. As of March 2015, 24 plants operating in the United States had either filed for license renewal or announced intentions to do so within the coming years [1]. Closure of these plants would significantly decrease the country's nuclear power generation capabilities, and improvements required to continue operation could be extremely expensive. Continued operation without thorough inspection and analysis, however, could have disastrous and even more costly consequences. Among the factors considered by the Nuclear Regulatory Commission when reviewing license renewal requests, structural integrity of steel used in reactor pressure vessels is weighted heavily.

The elastic behavior of ferritic steels such as A508 is paramount, as the transition from ductile to brittle behavior causes catastrophic failure. Materials that exhibit ductile behavior will continue to visibly deform before fracturing, as shown by the relatively large strain at failure for ductile materials (Figure 1). Brittle materials, however, fail unexpectedly when subjected to a load greater than yield strength or cyclic loading beyond the fatigue limit. Another way to express the difference between ductile and brittle materials is that ductile materials are capable of absorbing more energy before failure. The energy absorbed is equal to the area under the stress-strain curve, which (as apparent in Figure 1) is much greater for ductile

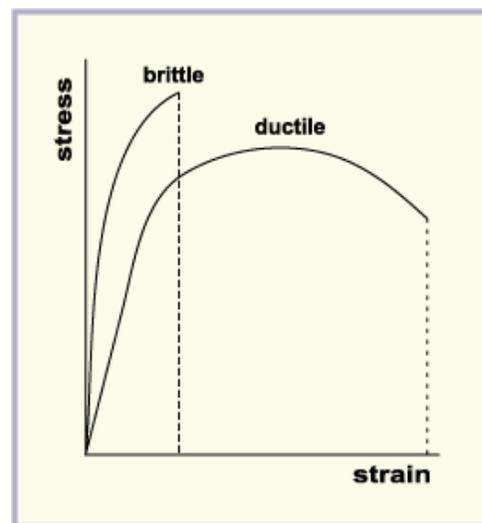


Figure 1. Comparison of stress-strain curves for ductile and brittle materials [3].

materials. Continued use of structures that display surface cracks is therefore permissible if constructed from ductile, but not brittle, material. Ferritic steels in pressure vessel reactors are categorized as ductile materials. However, ductile materials exhibit brittle fracture behavior at low temperatures, and randomly distributed microscopic brittle zones cause macroscopic brittle behavior in alloys under certain loading conditions [4]. When cracks are present in the material, these competing failure mechanisms make actual fracture toughness difficult to quantify [5]. During a criticality, the sudden temperature change caused by the activation of cooling water extends the effects of these micro-zones, further complicating the determination of material properties on a macroscopic level.

Mathematical Models

Scientists and engineers use the term “constraint effect” to denote the degree to which macroscopic behavior is governed by local brittle (also known as plastic) zones. “High constraint” conditions refer to cases in which the plastic behavior is constrained to a small region immediately surrounding the crack tip. Specimens in which plastic behavior predominates in areas far from the crack tip are classified as “low constraint”. High and low constraint are therefore relative terms which are used to classify material behavior based on specimen geometry and loading conditions [6]. Recent studies have shown that high constraint specimens exhibit greater experimental fracture toughness than low constraint specimens, necessitating changes in the mathematical model for failure prediction [7]. It should be noted that failure due to crack propagation through local plastic zones is known as “onset of cleavage fracture”. The critical values of various parameters are therefore defined at this stress state.

Mathematical prediction of cleavage failure was introduced in a seminal paper by Hutchinson, Rice, and Rosengren in 1968. In this paper, stress in a power-law hardening nonlinear material at some distance from the crack tip was expressed as a stress field. The relationship between stress and strain for such a material is defined according to the Ramberg-Osgood equation

$$\frac{\varepsilon}{\varepsilon_0} = \frac{\sigma}{\sigma_0} + \alpha \left(\frac{\sigma}{\sigma_0} \right)^n \quad (1),$$

where both n and α are material constants. To express stress at a certain distance r from the crack tip, the HRR solution employs the J integral, a measure of energy absorbed per fracture surface area. Thus, the stress at location (i,j) is defined as

$$\sigma_{ij} = \sigma_0 \left(\frac{J}{\alpha \varepsilon_0 \sigma_0 I_n r} \right)^{\frac{1}{n+1}} \tilde{\sigma}_{ij}(\theta, n) \quad (2).$$

All variables except σ_{ij} , J , and r are material properties. Thus, these three quantities are used to fully characterize fracture conditions according to the HRR method.

While useful, the HRR solution has significant limitations. Pure dependence of stress on the J integral at a given distance r from the crack tip ignores constraint effects. Furthermore, this solution assumes only very small deformations, which may not be the case if high constraint conditions exist. The result of these simplifications is an overly conservative prediction of material toughness. Sharpe and Chao therefore propose an alternate expression which is based on an expansion of the Ramberg-Osgood equation [7]. This expansion adds two terms to the HRR solution, giving the equation

$$\frac{\sigma}{\sigma_0} = \left(\frac{J}{\alpha \varepsilon_0 \sigma_0 I_n L} \right)^{-s_1} \left[\left(\frac{r}{L} \right)^{s_1} \tilde{\sigma}_{ij}^{(1)} \sigma_n - A_2 \left(\frac{r}{L} \right)^{s_2} \tilde{\sigma}_{ij}^{(2)} \sigma_n + A_2^2 \left(\frac{r}{L} \right)^{s_3} \tilde{\sigma}_{ij}^{(3)} \sigma_n \right] \quad (3).$$

When a material failure curve is known, the value of the J integral at failure (J_c) can be determined from the intersection with the Crack Driving Force curve. As explained in [7], the Crack Driving Force curve is generated by plotting multiple $(J, |A_2|)$ pairs at a constant load. This relationship between the empirically determined material failure curve and the simulated Crack Driving Force curve allows the validity of the J- A_2 method to be assessed.

Purpose

The aim of this project is to demonstrate that the same results can be obtained more efficiently and with lower probability of human error by taking advantage of the ability to execute functions in Abaqus finite element analysis software via Python script. Furthermore, the streamlined data processing method is used to generate results for a specimen not included in previous studies. A comparison between

these results and experimental behavior is then used to assess degree to which the J-A₂ method is further validated.

Method

Description of Abaqus Model

The model used in the process development and validation portion of this study (shown in Figures 2 and 3) is the same as that used by Sharpe and Chao. A brief summary of geometry, loading conditions, and mesh properties is provided, and more details can be found in [4] and [7]. Dimensions of the entire part, including crack tip are shown in Figure 2.

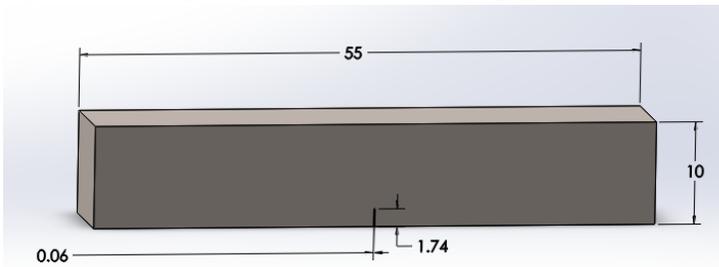


Figure 2. 3D rendering of shallow-cracked specimen (millimeters).

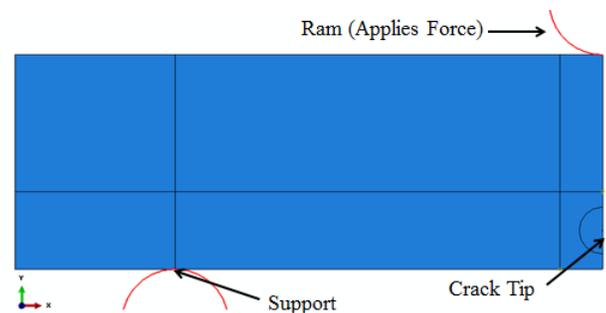


Figure 3. Abaqus model of 3-Point Bending specimen with a shallow crack.

Due to the symmetric part geometry and loading conditions, modeling only the left half of the part saved computation time without compromising the accuracy of results. The left half of the shallow crack specimen was modeled in Abaqus as a 2-dimensional deformable planar part (Figure 3) with the deformation plasticity properties of A508 steel. The ram and the support were also created in 2-dimensional space as analytical rigid parts. To model the conditions of 3-point bending, a surface contact interaction between the specimen and the two loading parts (support and ram) was defined. Furthermore, the support was defined as having zero displacement during the loading process. Since the ram was used to load the specimen, the initial value of displacement was changed in increments of 0.5 for a total of 100 increments. A node on the right edge and 1.8mm from the bottom of the specimen was selected as the crack tip to reflect the a/W ratio of 0.18 in the experimental set-up. The direction of crack propagation was set to the positive y direction along the specimen edge, and the option to model as a half-crack was selected to reflect the symmetry incorporated into the model.

The mesh shown in Figure 4 was created by the authors of [7] and used in this analysis without modification. Figure 4a shows the global mesh applied to areas far from the crack tip. Because the area immediately surrounding the crack tip was the most important, a semicircle with a diameter of $4\mu\text{m}$ contained 640 quadrilateral quadratic elements with reduced integration. To reduce computation time, a total of only 272 quadrilateral quadratic reduced integration elements were defined over a region of 267.8 mm^2 that was considered outside of the plastic zone. The larger semicircular mesh, which is visible in Figure 4, contained 1024 elements of the same type. A transition region between the fine and coarse meshes used 105 triangular quadratic elements of increasing size to maintain continuity between the two sections of differently-sized quadrilateral elements.

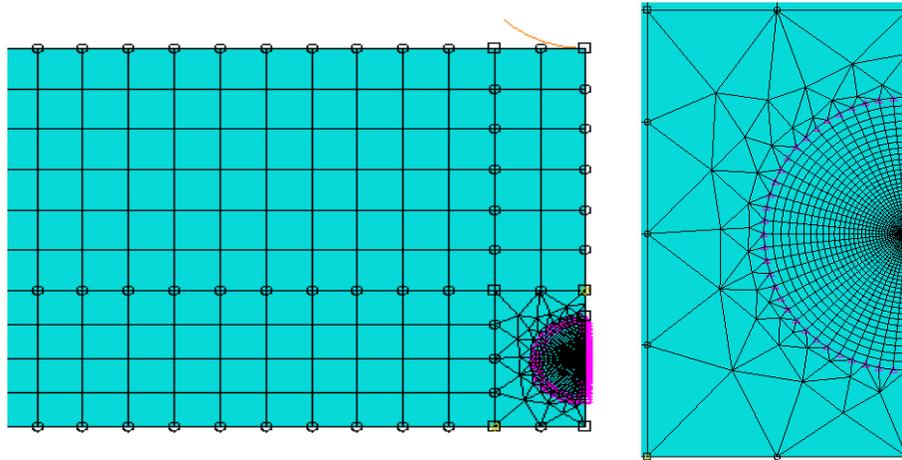


Figure 4. (a) Specimen with mesh. (b) Detail view of mesh surrounding crack tip.

Before running the Abaqus analysis, both field and history outputs were defined. Field outputs are those for which values depend on position, while history outputs are those for which values depend on time. Stress, displacement, and strain, along with other common physical parameters were specified as outputs. Because these values would increase throughout the loading process, the end of the loading process was specified as the output time. An additional output was requested in order to model crack growth. As the load is increased, energy is dissipated by both the crack growth and the deformation of the surrounding region. This energy release rate is reflected by the J-integral [8]. The area over which the J-integral is calculated is referred to as a contour. While Abaqus automatically calculates contours when a

crack is defined, the user must select the number to be used for the J-integral calculation. For this analysis, 30 contours were used, and the J-integral was calculated at contour 30.

Data Processing

The first step in data processing was to define a path that included the nodes along the line of crack propagation, as shown in Figure 5. The value of stress in the x-direction (S11) is then calculated at each node along the path for selected time points in the loading process. The previous method of data analysis required the user to manually select which frames (time points) to use in the analysis. For each frame, the S11 value at each node along the path was saved as an X-Y dataset, with distance from crack tip (r) on the x-axis and stress on the y-axis. Each X-Y dataset was

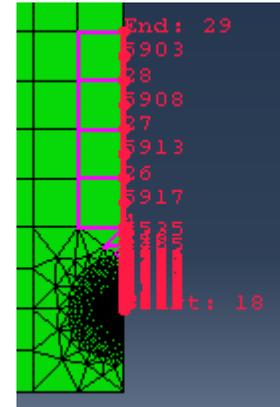


Figure 5. Crack propagation path selected in Abaqus.

copied manually into Excel for calculations. Additionally, the J-integral at contour 30 for each of the selected frames was recorded. Based on material constants and the additive solution of the quadratic in Equation 3, the value of A_2 was calculated at each node. To determine which points along the path should be included in the calculation of the composite A_2 value at each time, an additional parameter r_n was introduced as

$$r_n = r * \frac{\sigma_0}{J} \quad (4).$$

Points for which the value of r_n was greater than 2 and less than 5 were included, and the A_2 value was the average of the value calculated at each point. The J-integral was plotted against A_2 for the selected time points to form the crack driving force curve. Therefore, for each point on the graph, a dataset would have to be manually generated in Abaqus and copied into Excel. Several intermediate calculations were then required to reach the final value.

While the Excel spreadsheet approach was sufficient for processing data from a single analysis, the process would have to be repeated each time a feature of the model was changed. This limitation was a major deterrent to research on specimens of different shapes or materials. Python and Matlab, when

used in conjunction, provided a solution to this problem. First, a Python script containing a series of commands (a macro) is executed in the Abaqus environment. While the user must define the path before running the macro, the only other interaction required is entry of the job name and the path name. If a job has already been analyzed, and the user does not wish to delete the resulting files, an alternate unique identifier can be selected. The macro automatically uses frames from the second half of the loading process to generate the same datasets as those created manually. The J-integral dataset is written to a .dat file to be easily imported into Matlab. The other data sets are written to text files, which require additional parsing in Matlab. In order to ensure that the correct files are read in Matlab, the macro creates an additional text file which records the job name and the frames being analyzed. In Matlab, the user runs the analysis script and enters the name of the job or identifier. Matlab then creates a data structure called 'Calcs' which has a field for each frame. Each of the fields then has several subfields where the data from Abaqus and intermediate calculations are stored. After performing the calculations, a matrix of J values and A_2 values is produced and graphed, with J values on the y axis, as shown in Figure 8.

Analysis of Deep-Cracked Specimen

The validity of the J- A_2 method for predicting failure under higher constraint loading conditions was also evaluated by analyzing a deep-cracked specimen subjected to 3-point bending. Also made of A508 steel and tested at -85°C , data for this specimen was available in [4]. The deep-cracked specimens used in the study had a ratio of crack length to specimen length (a/W) of approximately 0.53, as opposed to the 0.18 shallow-cracked specimens. The authors of [7] had created an Abaqus model for this specimen with the crack tip location adjusted to 5.3 mm along the right edge, as shown in Figure 6. However, the investigators were previously unable to obtain reasonable

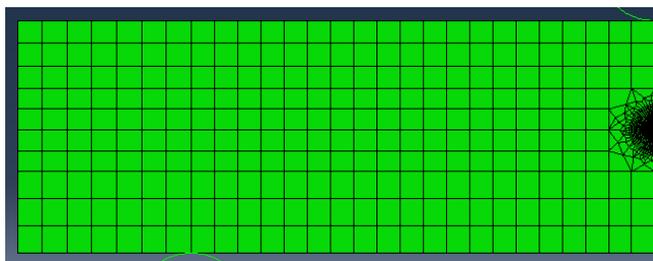


Figure 6. Abaqus model of 3-Point Bending specimen with a deep crack.

results using this model. In the time between publication and the present study, other researchers have

warned against the use of reduced integration elements around a crack tip. Reduced integration quadrilateral quadratic elements have eight nodes instead of nine due to the elimination of the center node. While numerical integration is faster with reduced elements, the mesh surrounding the crack tip was changed to be composed of full integration quadrilateral quadratic elements. Because of the high constraint conditions and resulting lower material toughness, the rate of loading was also reduced from 0.5 mm/s to 0.2 mm/s. In all other respects, the analysis of the deep-cracked specimen was the same as that used with the shallow-cracked specimen.

Results and Discussion

Process Validation

Before analyzing additional datasets using the Python and Matlab scripts, the results had to first be compared to those obtained with the original method. That is, the crack driving force curve was required to be the same shape and intersect the material failure curve at the same value in order for the process to be considered valid. Figures 7 and 8 show the driving force curve plotted with the material failure curve for the previous procedure and the procedure developed in this study, respectively. The use of different frames had a slight effect on the values used to create the curve, and the range of values for used in the new procedure was smaller. It should be noted that the point labeled in Figure 7 is not the actual intersection point but the experimental data point closest to the intersection. The intersection clearly lies slightly above this data point at a y value of approximately 55 N/mm, which is consistent with the value in Figure 8. This data processing procedure was therefore validated against the original process, allowing more efficient analysis of other models.

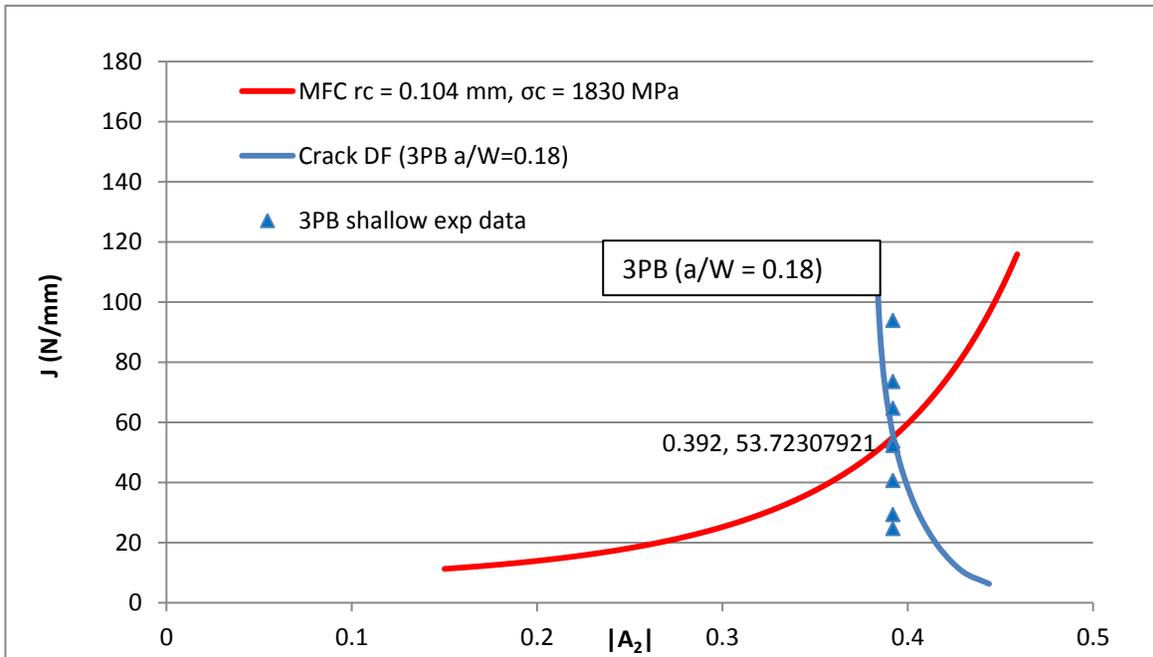


Figure 7. Graphical representation of Excel analysis results for shallow-cracked specimen.

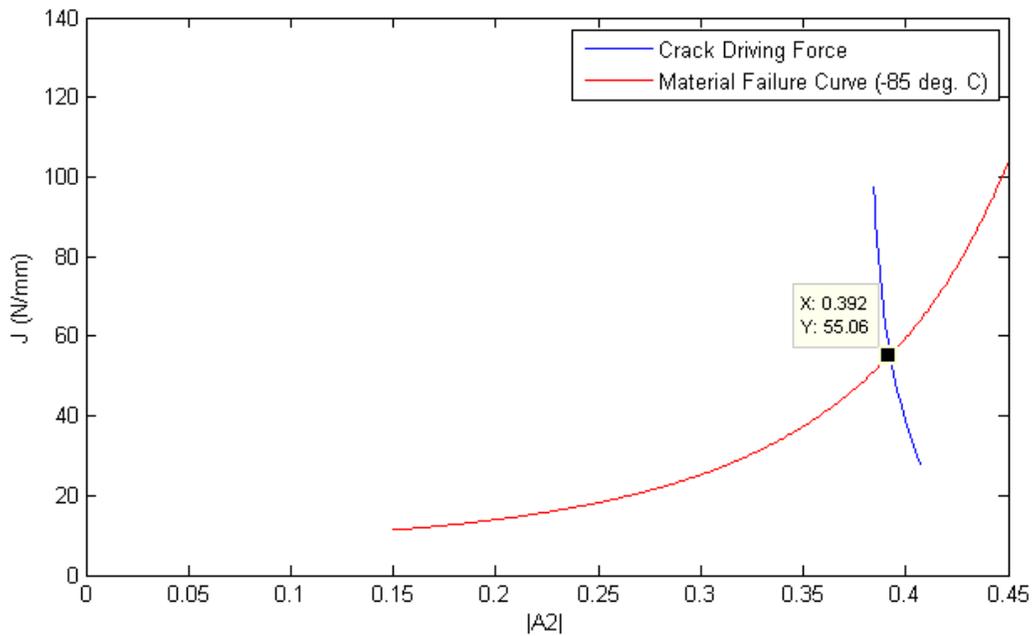


Figure 8. Graphical representation of Matlab analysis results for shallow-cracked specimen.

Deep-Cracked Specimen

In the paper that introduced the $J-A_2$ method for fracture prediction, the authors validated the results by noting that the intersection between the crack driving force curve and the material failure curve fell in the middle of the range of experimental failure values [7]. Figure 7 illustrates this relationship between the three datasets. Although the material used for the deep-cracked specimen is identical, the same relationship does not hold true. As shown in Figure 9, the crack driving force curve intersects the material failure curve well below actual failure points. However, the experimental failure data is not centered on the material failure curve, as was the case in Figure 8. The rightward shift of the failure curve relative to the actual data points indicates that the model would predict lower constraint, and therefore greater material toughness than demonstrated experimentally. The scatter of experimental data above the material failure curve, however, suggests that this failure curve may not accurately represent experimental conditions. A fourth order polynomial best-fit line is used to show that the intersection between the driving force curve and the experimental data would occur at approximately the midpoint of the spread. However, as best-fit lines were not used in previous analyses, the conclusions that can be drawn about the validity of the $J-A_2$ method are extremely limited.

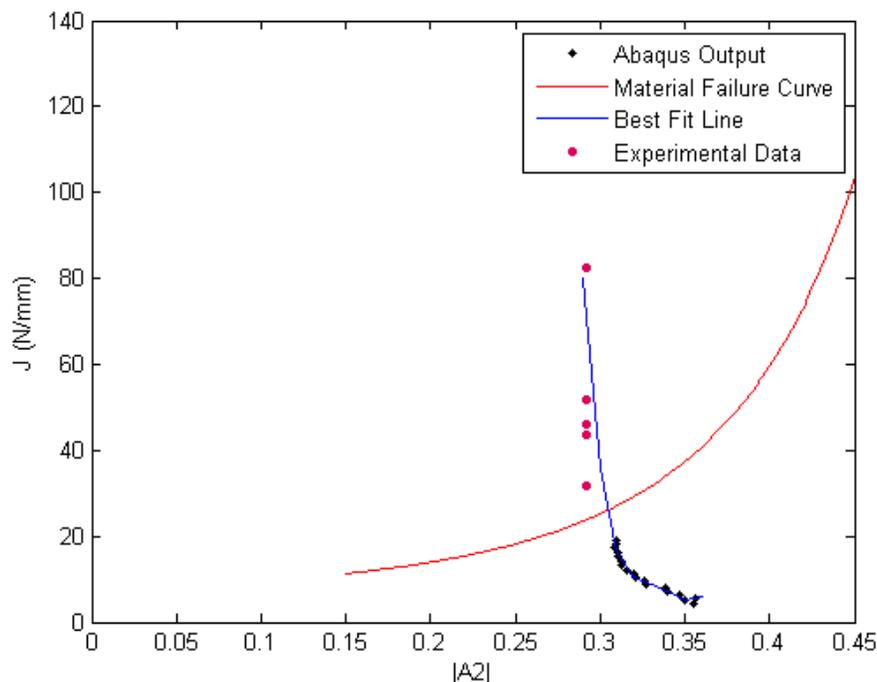


Figure 9. Graphical representation of results from analysis of deep-cracked specimen.

Limitations and Future Directions

The original objective of this investigation was to obtain more data to evaluate the validity of the J-A₂ method. However, the cumbersome process of data analysis was a major hindrance and likely to deter future students from continuing the project. The scope of the project therefore shifted, although the analysis of the deep-cracked specimen partially fulfilled the original goal. In order to continue evaluating the J-A₂ method, more data will have to be gathered. The Matlab and Python scripts are robust enough to handle different materials (with a few adjustments) and differently shaped specimens, making future analyses much more efficient. The availability of test data is somewhat limited, however, which may necessitate material testing on campus. After more research is conducted, the accuracy and potential limitations of the J-A₂ method can be more adequately discussed.

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