



THE UNIVERSITY OF TEXAS AT EL PASO
COLLEGE OF ENGINEERING

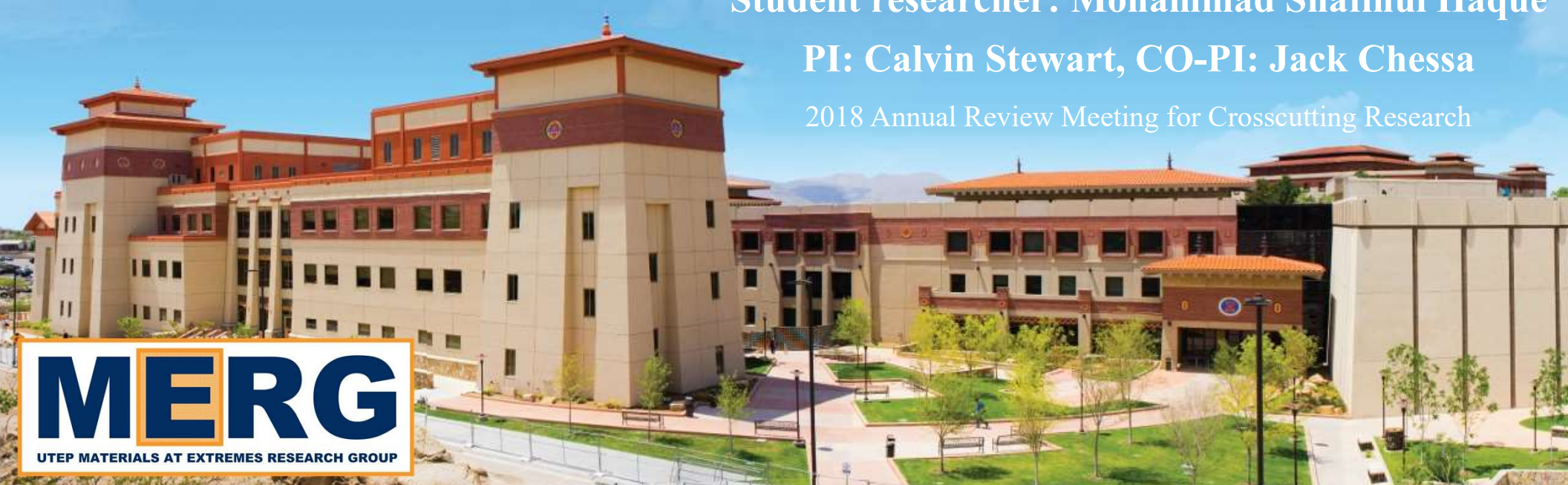


A Guideline for the Assessment of Uniaxial Creep and Creep-Fatigue Data and Models

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PI: **Calvin Stewart**, CO-PI: **Jack Chessa**

2018 Annual Review Meeting for Crosscutting Research



Outline

Motivation

Research Objectives

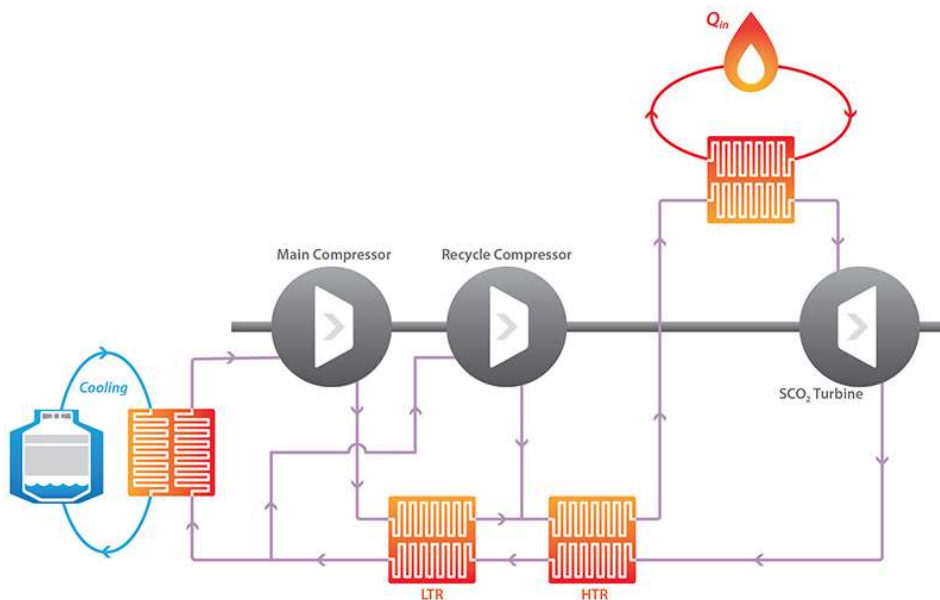
Systematic Approach to Assessment

Results and Accomplishments

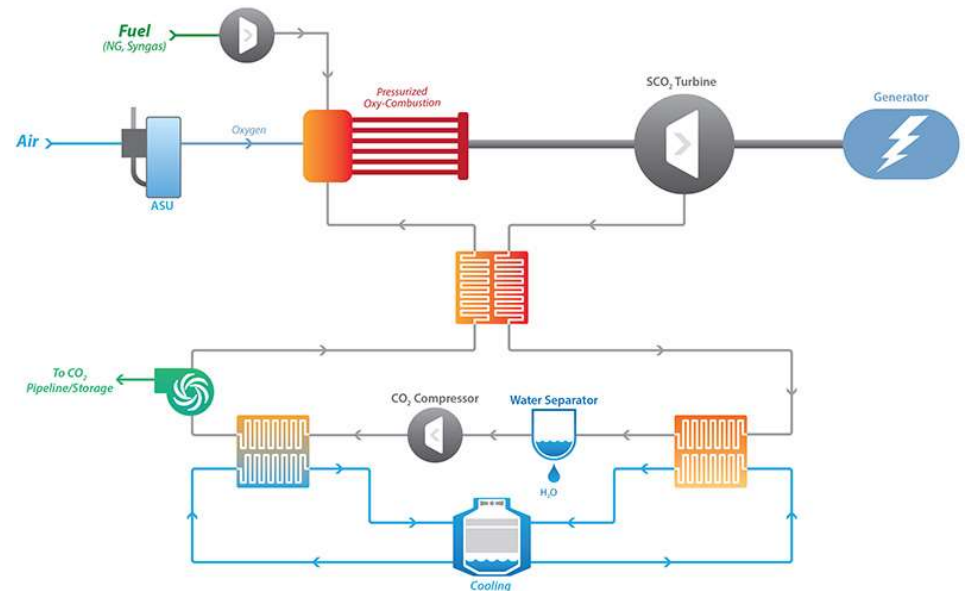
Summary

Motivation

- Recent drives to increase the efficiency of existing fossil energy (FE) power plants and the development of **Advanced Ultrasupercritical (A-USC) power plants**, have led to designs with steam pressures **above 4000 psi** and temperatures **exceeding 1400°F**.



Indirect-Fire Supercritical CO₂ Recompression Brayton Cycle



Oxy-Fueled Directly-Fired Supercritical CO₂ Cycle

Motivation

- The existing FE fleet has an **average age of 40 years**.
- The Department of Energy has outlined a strategy of life extension for US coal-fired power plants where many plants will operate for **up to 30 additional years of service**.

In Service Hours....

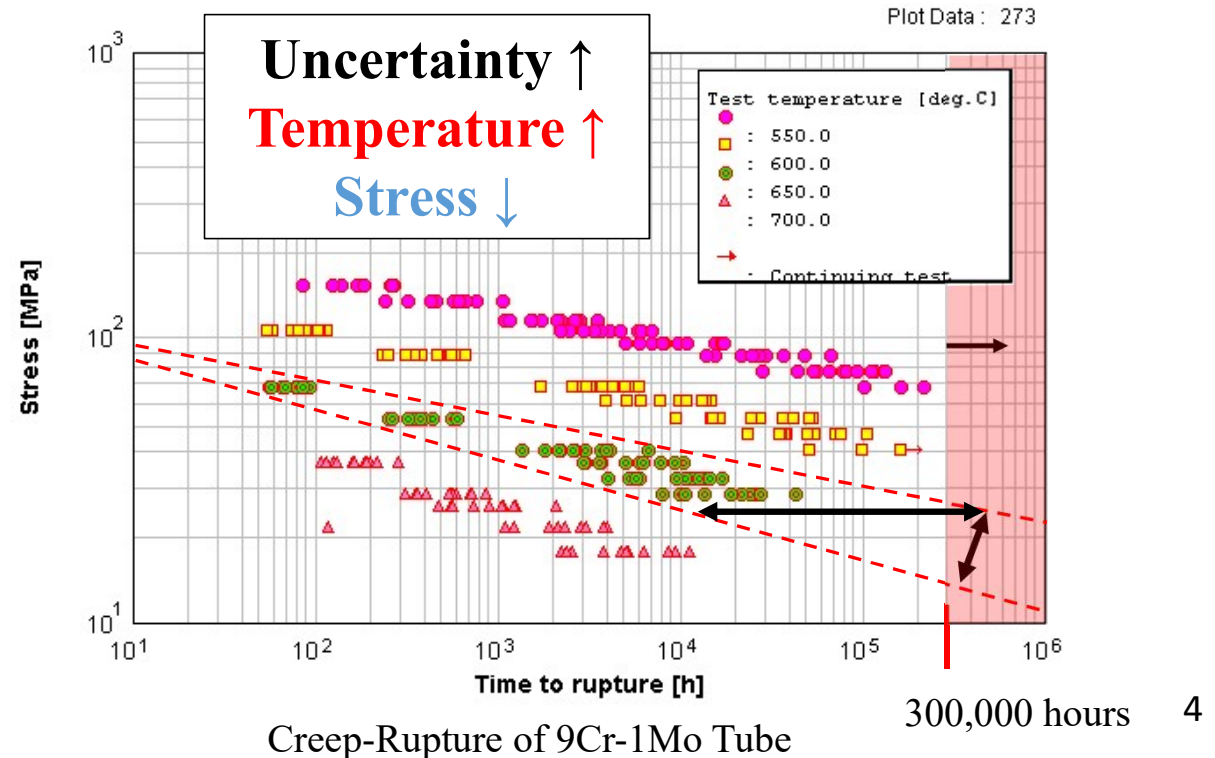
30 Years = 262,974 hours



40 Years = 350,634 hours

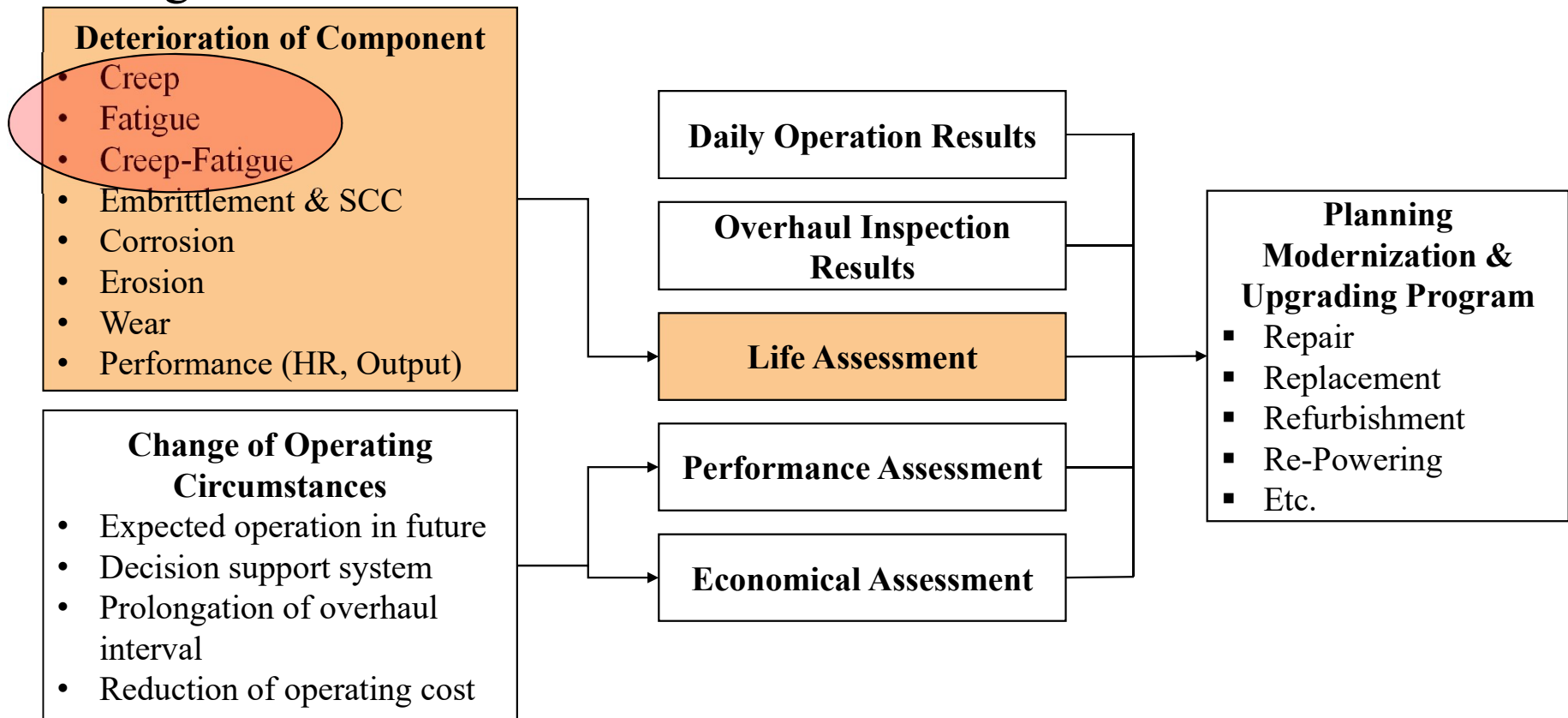


70 Years = 613,607 hours



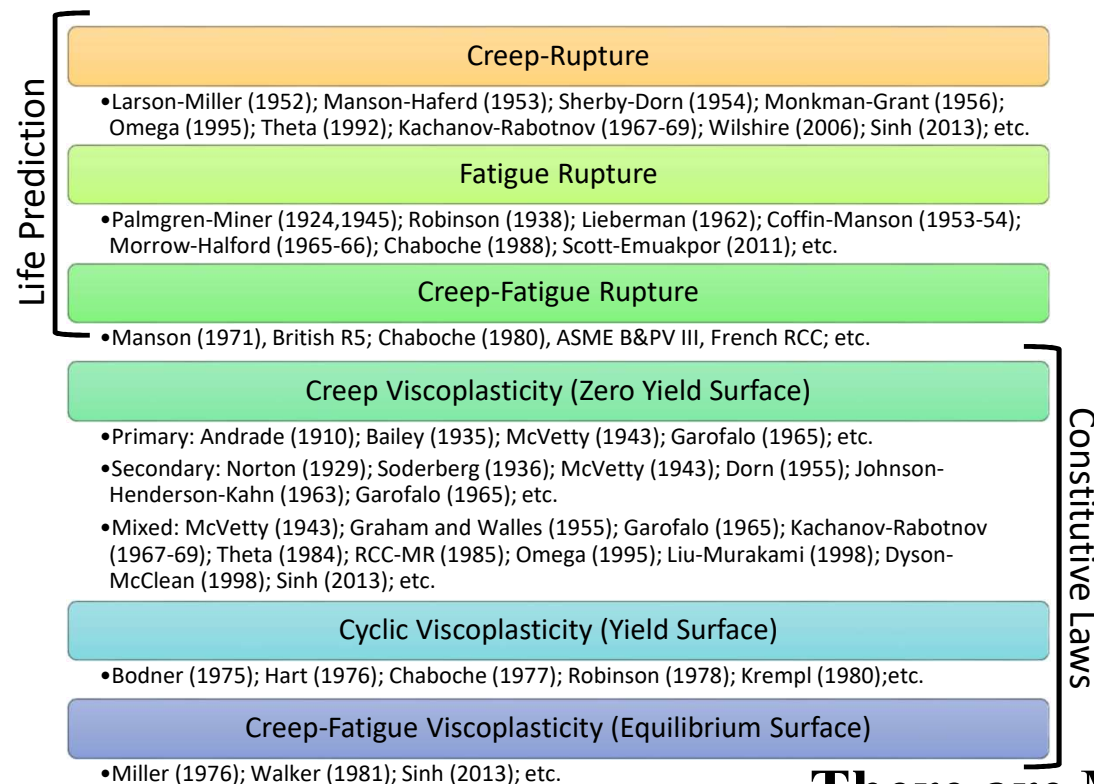
Plant Life Extension Program

- During Life Assessment, the integrity of components is assessed and the remaining service life estimated.



Motivation

- An immense number of models have been developed to predict the deformation, damage evolution, and rupture of structural alloys subjected to Creep and Creep-Fatigue.



There are Many More!...

Research Objectives

- Of primary concern to FE practitioners is a determination of **which constitutive models are the “best”**, capable of reproducing the mechanisms expected in an intended design accurately; as well as **what experimental datasets are proper or “best” to use** for fitting the constitutive parameters needed for the model(s) of interest.

RO1

Development of
**Aggregated Experimental
Databases** of Creep and
Creep-Fatigue Data

RO2

**Computational Validation
and Assessment** of Creep
and Creep-Fatigue
Constitutive Models for
Standard and Non-Standard
Loading Conditions

The Team



Calvin M Stewart, Project PI



Jack F Chessa, Project Co-PI



Mohammad Shafinul Haque
(PhD, Spring 2018)



Christopher Ramirez
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Jimmy Perez
(MS)



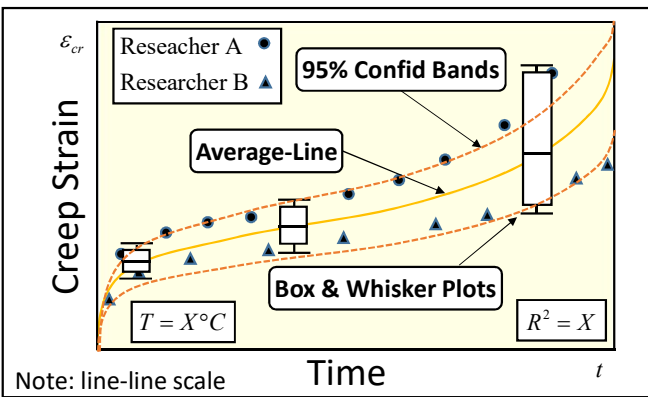
Amanda Haynes
(UG, ACT program)



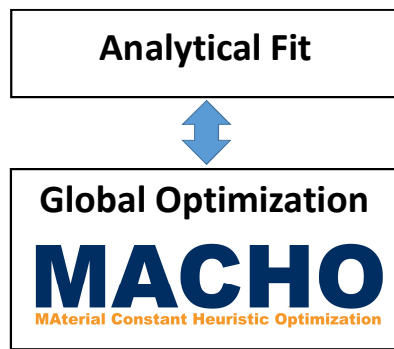
Ricardo Vega
(Undergrad)

Systematic Approach to Assessment

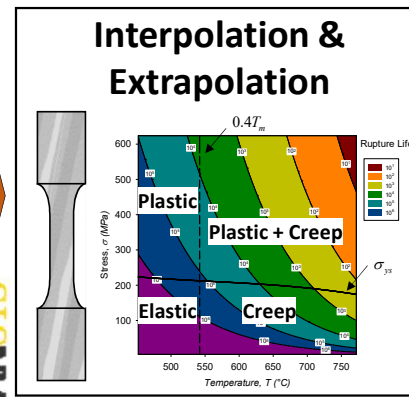
Example for Creep Deformation



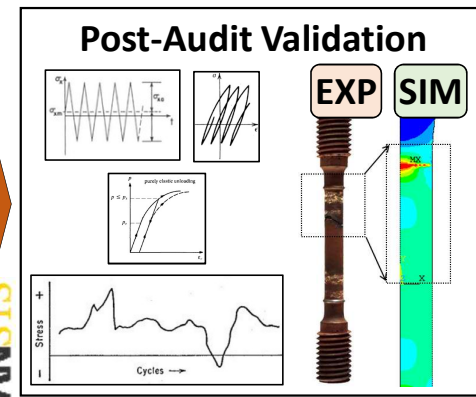
Aggregate Datasets with Uncertainty



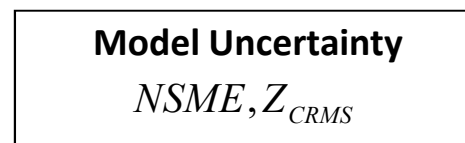
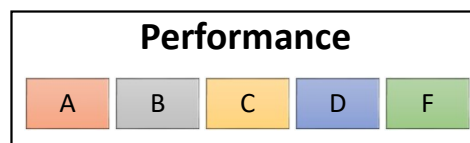
Model Fit to Datasets



Standard Performance



Nonstandard Performance



Systematic Approach to Assessment

Task 1: Locate, Digitize, Sort, Store Experimental Data

Task 2: Uncertainty and Integrity of Experimental Database

Task 3: Mathematical Analysis and FEA of the Models

Task 4: Calibration & Validation – Fit, Interpolation, Extrapolation of the Models

Task 5: Post-Audit Validation of the Models

Task 6: Disparate Data problem and Design Maps

Task 7: Metamodeling: Finding the “best” model

Task 8: Multiaxial Representative Function

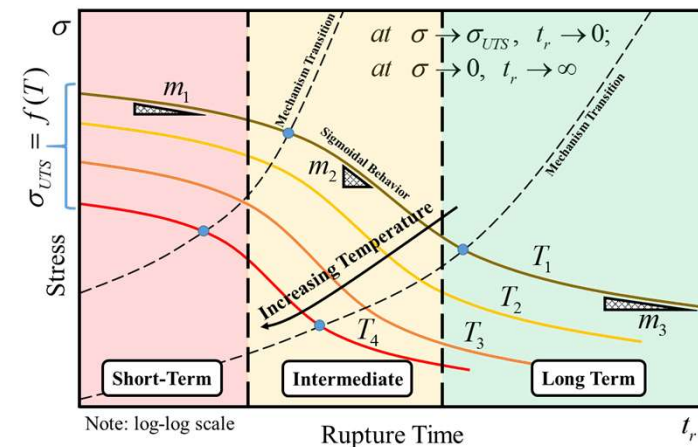
Task 1: Locate, Digitize, Sort, and Store Data

Creep Data

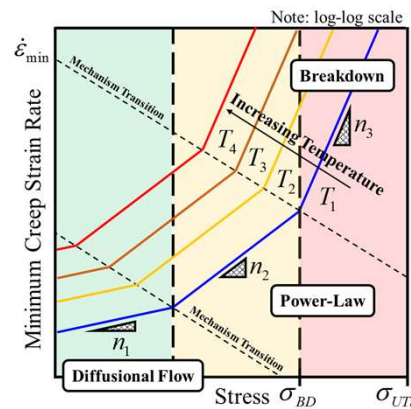
Creep-rupture
 Minimum creep strain rate
 Time to creep strain
 Creep deformation
 Stress relaxation

Creep-Fatigue Data

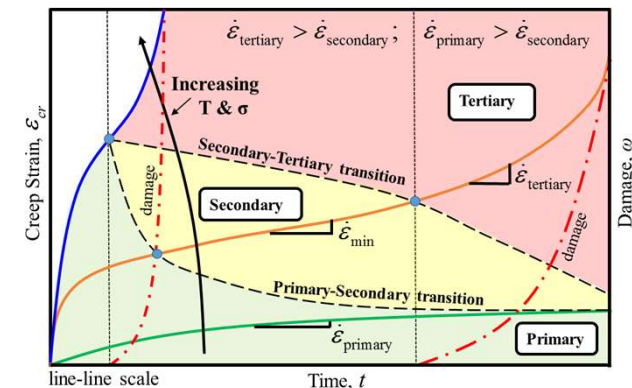
Tensile Hold Tests



Ideal creep rupture curve

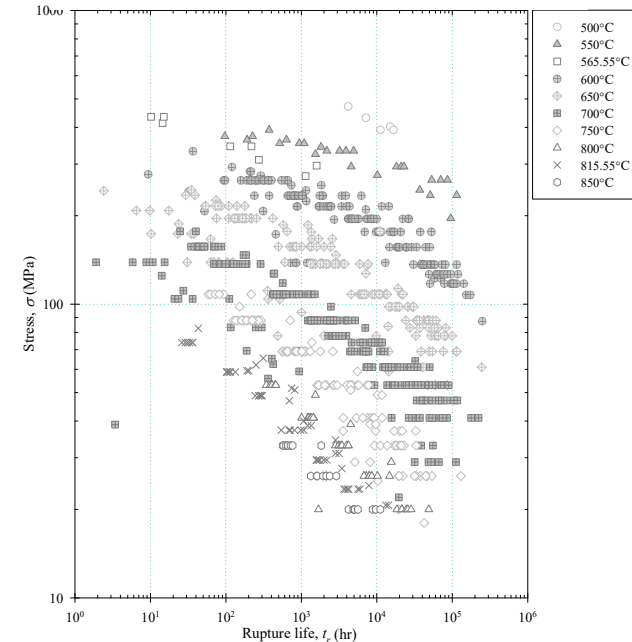
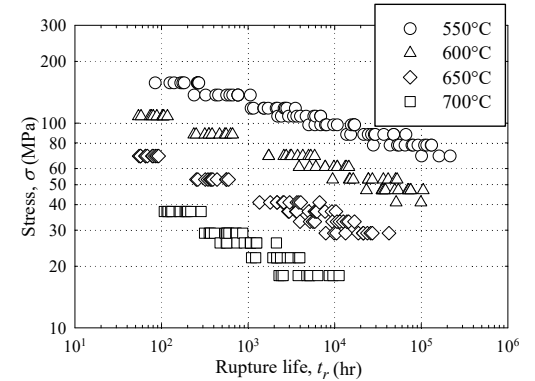
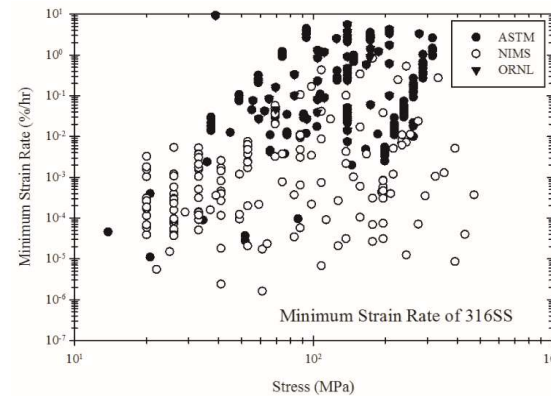
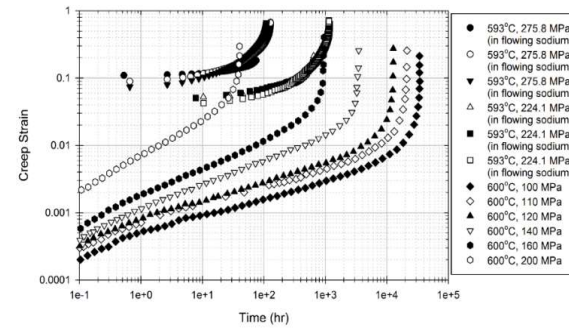
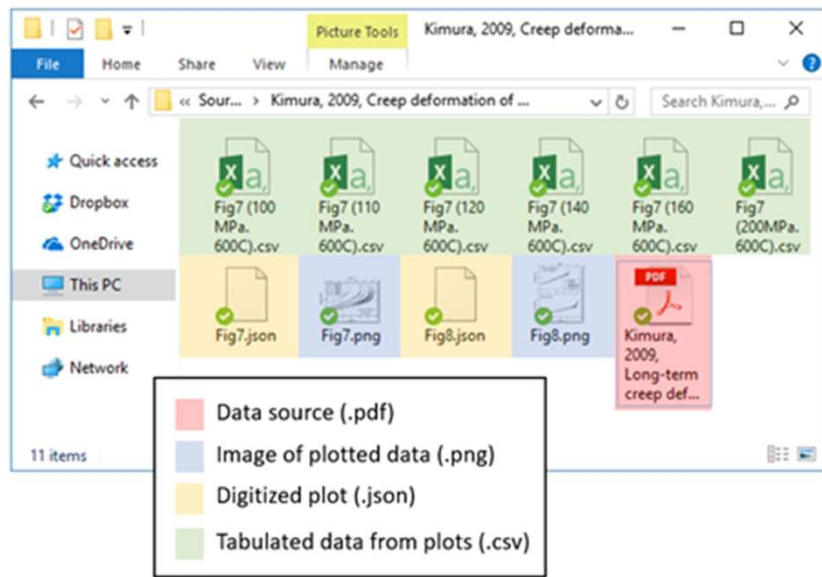


Ideal minimum creep strain rate curve

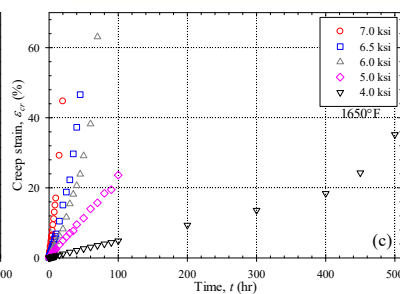
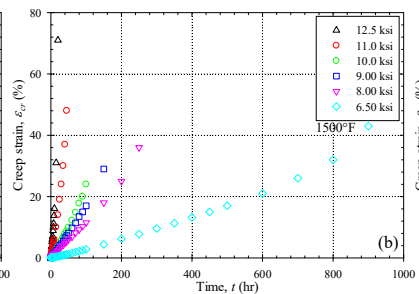
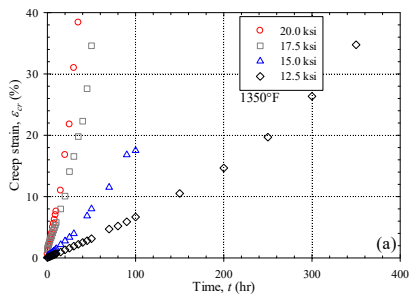


Ideal creep deformation and damage evolution

Task 1: Locate, Digitize, Sort, and Store Data



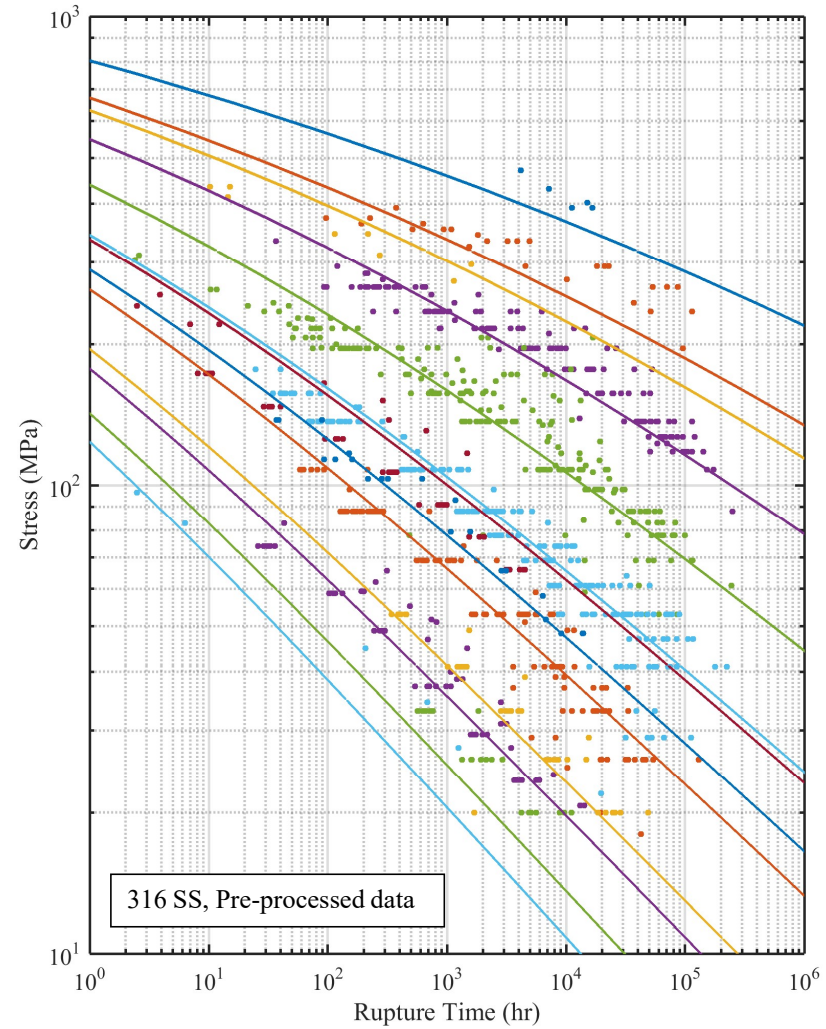
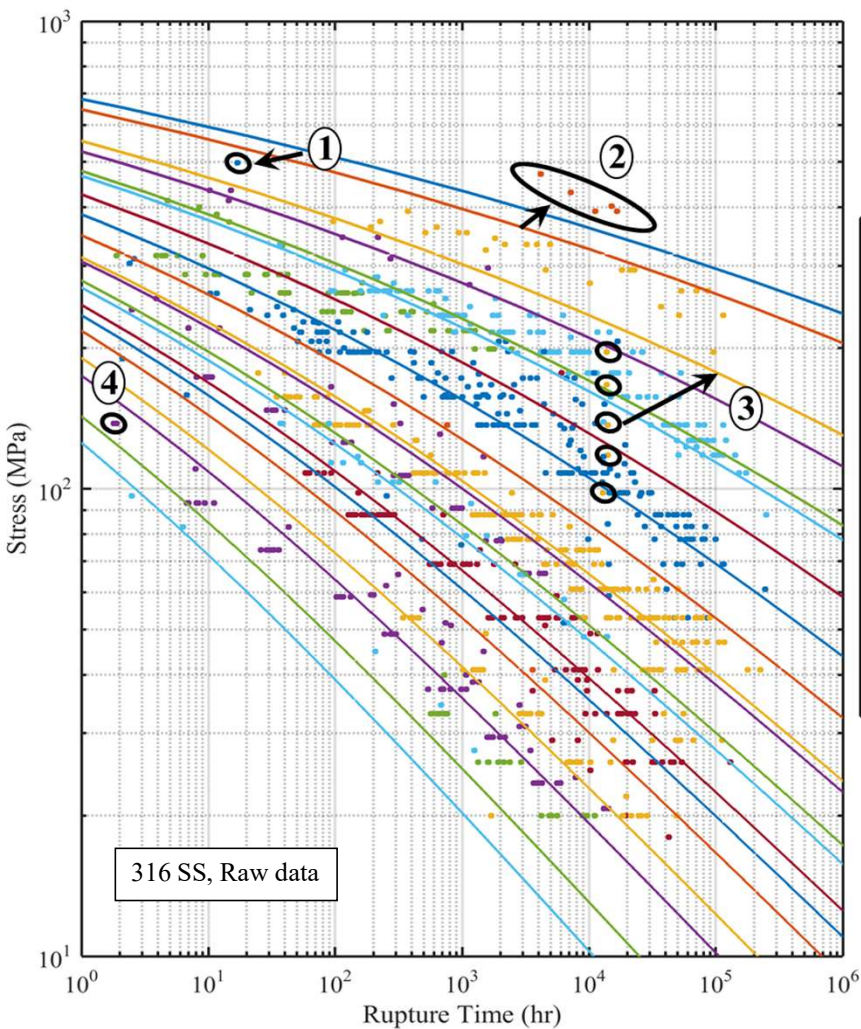
Creep deformation



Task 1: Locate, Digitize, Sort, and Store Data

Table 3 – Data collected by data source									
Source	Creep Deformation	Stress Relaxation	Min. Strain Rate	Time to Cr. Strain	Creep Rupture	Mono. Tensile	Cyclic Hysteresis	Stress Amp/Cycle	Cr.-Fatigue Tensile
ARL-79-33		9							16
ASM Atlas of Creep & Stress-Rupture	20								
ASM Atlas of Fatigue									
ASM Atlas off Stress-Corrosion Fatigue					14			42	14
ASM Atlas of Stress-Strain						106			
ASTM DS-60									
ASTM DS5-S1			144		295				
ASTM STP 124			43	22	85				
ASTM STP 522			135	90	160			24	
Booker and Sikka, 1976				183					
Choudary, 2009			56				13	4	13
Fournier (1), 2008								29	
Fournier (2), 2008							2	12	18
Fournier (3), 2008		13						19	161
Nagesha, 2002									
NIMS Database		210	245	764	1184	207			
ORNL TM-10504	9		38	46	12				
ORNL-101053									
ORNL/TM-6608									
ORNL-5237	6		52		55				
Rau, 2002							30	18	
Rowe, 1963	69		96	78	96				
Shankar, 2006								51	6
Takahasi, 2008									51
Yan, 2015									399
Kimura, 2009			33						

Task 2: Uncertainty Analysis (Data Pre-processing)



NMSE
reduced
from
164.25 to
16.65.

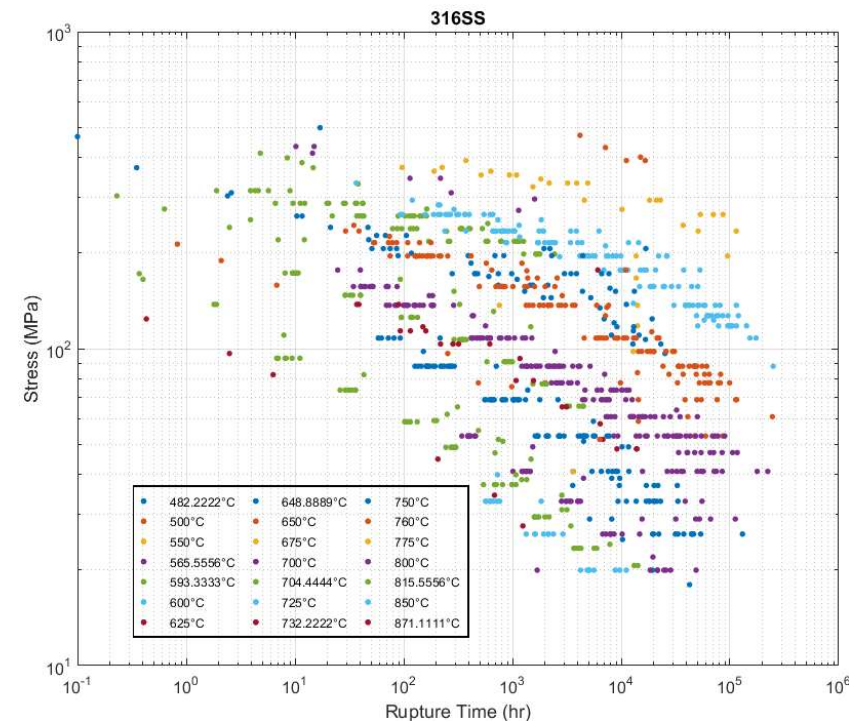
Task 2: Uncertainty Analysis (Metadata)

- **Metadata** include: form, thermomechanical processing, source, chemistry, geometry, laboratory code, etc.

Form	Data points
Bar	438
Bar and Plate	17
Pipe	9
Plate	210
Tube	321

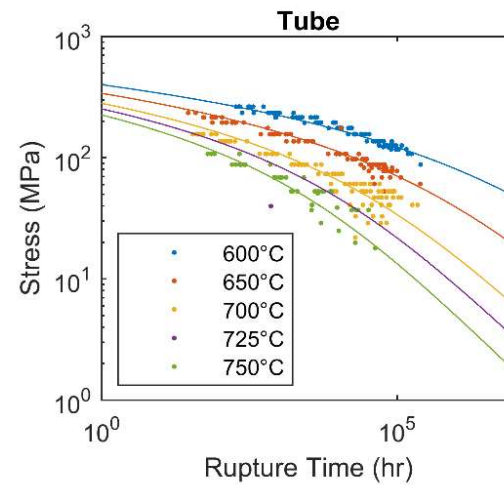
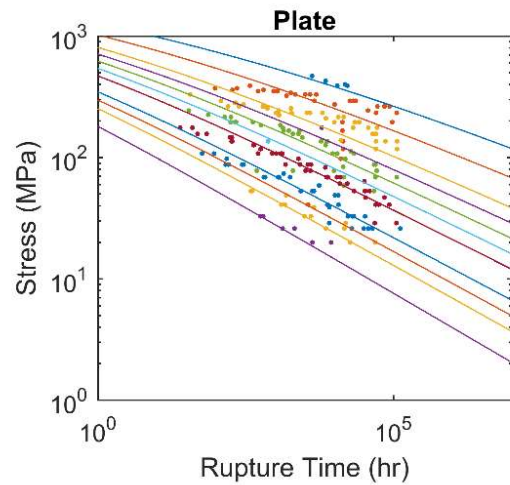
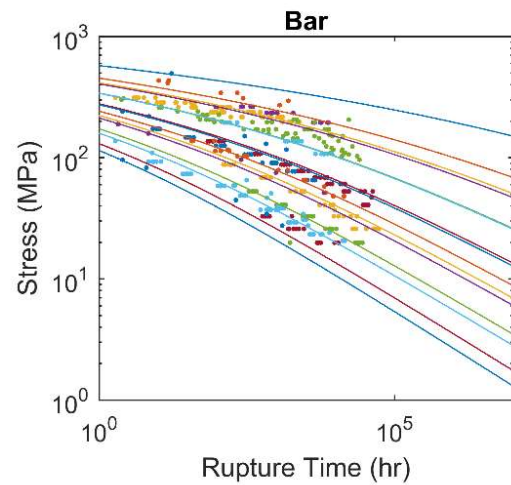
Source	Data points
ASTM DS5-S1	289
ASTM STP 552	9
NIMS online database	697

Processing	Data points
Annealed	11
Forged	19
Hot extruded	9
Hot extruded and cold drawn	167
Hot rolled	520
Quenched	115
Rotary pierced and cold drawn	154



Combined data

Data parsing



$$Z = t_{r,exp} - t_{r,LMP}$$

$$\log(Z) = \log(P_{exp}) - \log(P_{LMP})$$

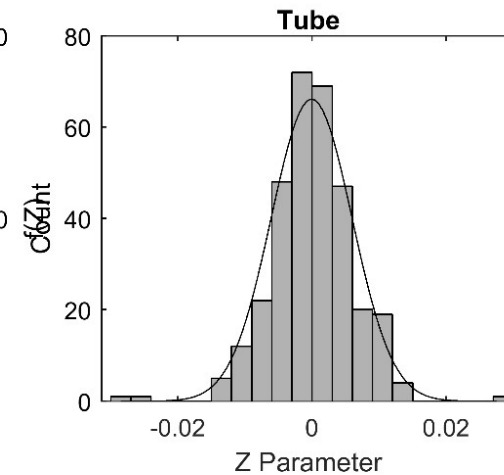
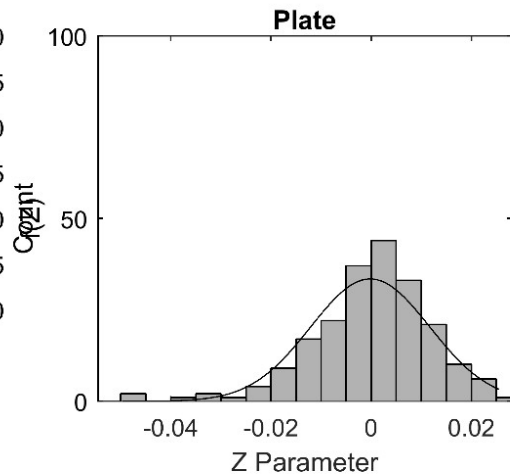
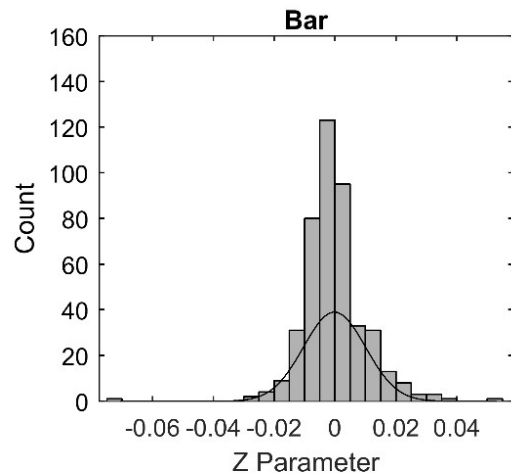
Rupture Prediction
(100MPa, 500°C)

Form	Rupture time (hrs)
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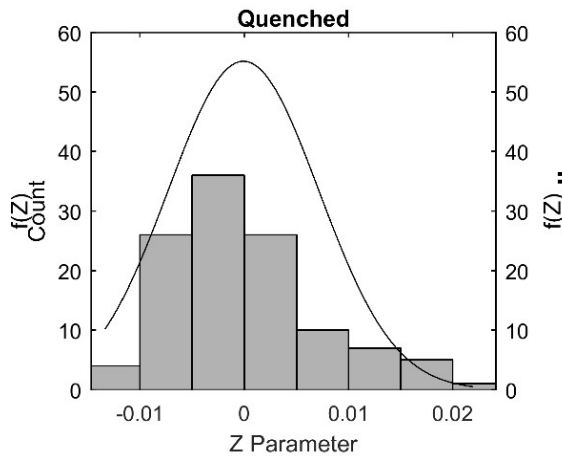
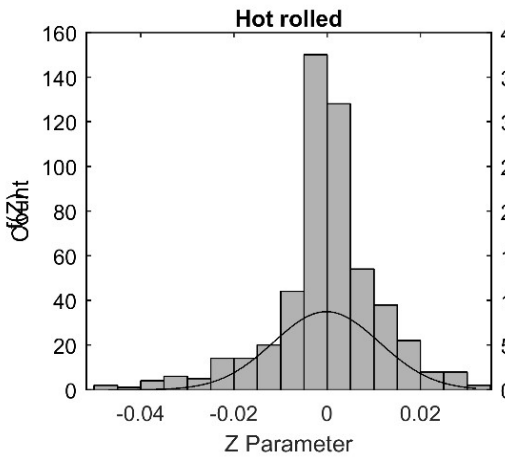
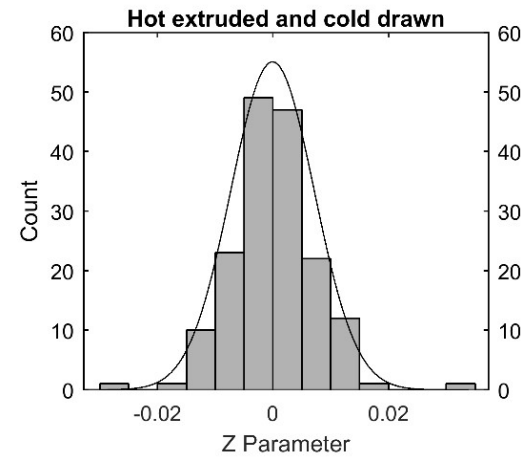
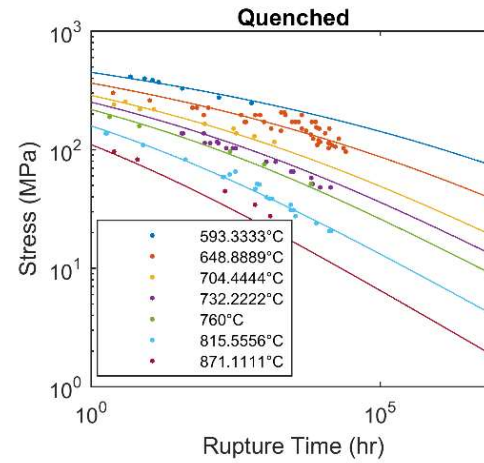
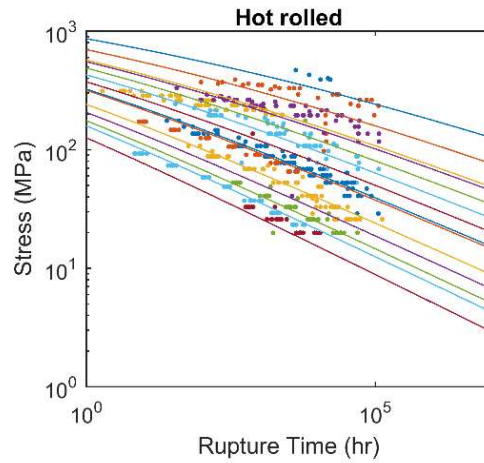
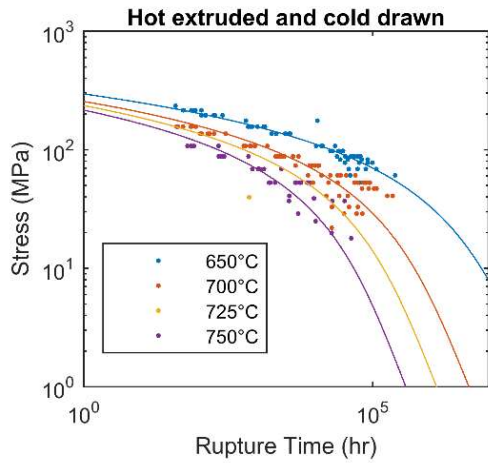
Bar	6.77E7
-----	--------

Plate	2.57E7
-------	--------

Tube	28.2E7
------	--------



Data parsing



Rupture Prediction (100MPa, 500°C)	
TMP	Rupture time (hrs)
HE & CD	14.5E7
HR	5.05E7
Quenched	105E9

Data Cull

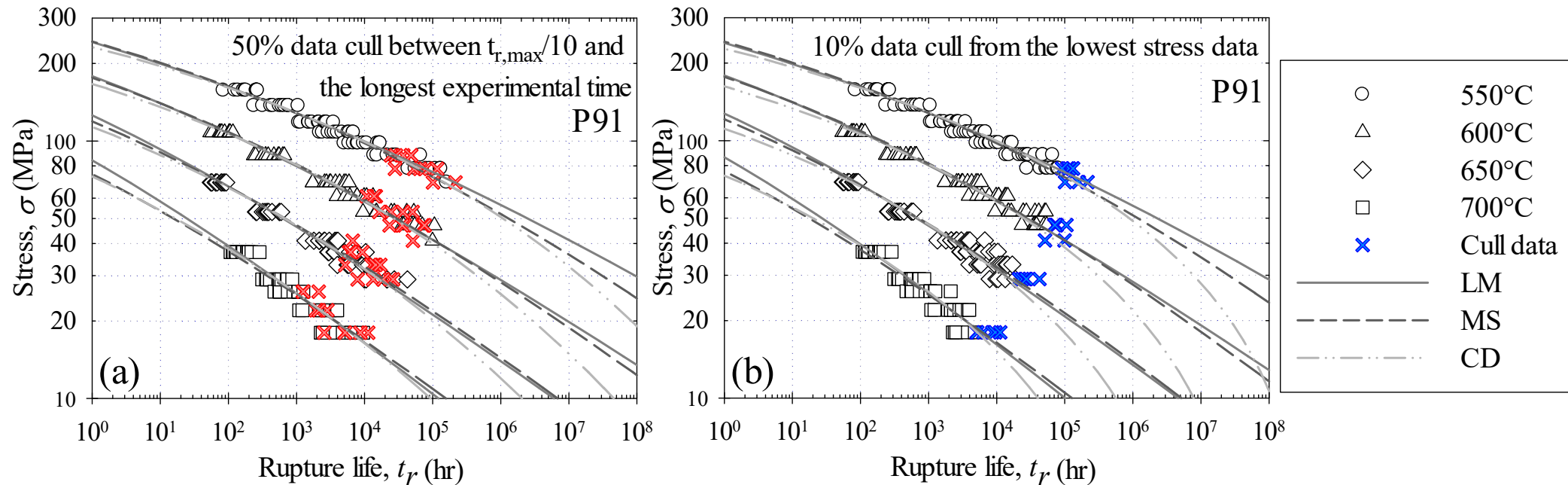


Figure 10 – Rupture prediction of LM, MS, and CD against (a) 50% data cull between $t_{r,max} / 10$ and the longest time and (b) 10% data cull from the lowest stress data

Task 3: Mathematical Analysis

Stress-Rupture

- **Eight** commonly used TTP models
- **Four** newly developed model

Creep-deformation

- Omega model
- Theta Projection

Minimum Strain rate

- Norton Power Law
- McVetty Law
- **Four** additional model

Continuum Damage Mechanics

- Kachanov-Rabotnov
- Sin-hyperbolic
- Liu-Murakami

Multiaxial Representative Stress Functions

- Hayhurst
- Huddleston
- Additional **five** model

Rupture, Deformation, and Steady-state models

Table: Creep-Rupture Models

Model	Year	Parametric equation
Larson-Miller	1952	$P_{LMP} = T(\log(t_r) + t_a)$
Manson-Haferd	1953	$P_{MH} = \frac{\log(t_r) - \log(t_a)}{T - T_a}$
Manson-Brown	1953	$P_{MB} = \frac{\log(t_r) - \log(t_a)}{(T - T_a)^n}$
Orr-Sherby-Dorn	1954	$P_{OSD} = \log(t_r) - Q / RT$
Manson-Succop	1959	$P_{MS} = \log(t_r) - BT$
Graham-Walles	1955	$P_{GW} = \frac{\log(t_r)}{(T - T_a)^n}$
Chitty-Duval	1963	$P_{CD} = mT - \log(t_r)$
Goldhoff-Sherby	1968	$P_{GS} = \frac{\log(t_r) - \log(t_a)}{1/T - 1/T_a}$
Modified Manson-Haferd	--	$P_{MMH} = \frac{\log(t_r) - \log(t_a)}{T}$
Modified Graham-Walles	--	$P_{MGW} = \frac{\log(t_r)}{(1/T - 1/T_a)^n}$
Modified Chitty-Duval	--	$P_{CD} = \frac{m}{T} - \log(t_r)$
Modified-Goldhoff-Sherby	--	$P_{GS} = \frac{\log(t_r) - \log(t_a)}{(1/T - 1/T_a)^n}$

Table: Creep deformation

Omega model

$$\varepsilon = \theta_1[1 - \exp(-\theta_2 t)] + \theta_3[\exp(\theta_4 t) - 1]$$

$$\dot{\varepsilon}_t = \theta_3 \theta_4 \exp(\theta_4 t)$$

Theta model

$$\dot{\varepsilon} = \dot{\varepsilon}_0 \exp(\varepsilon \Omega)$$

$$\omega = \frac{t}{t_r} = \frac{\dot{\varepsilon} \Omega t}{1 + \dot{\varepsilon} \Omega t}$$

Table: Minimum-creep-strain-rate model

Source	Creep law
Norton, 1929	$\dot{\varepsilon}_{cr} = A(\sigma / \sigma_0)^n$
Soderberg, 1936	$\dot{\varepsilon}_{cr} = A\{\exp(\sigma / \sigma_0) - 1\}$
McVetty, 1943	$\dot{\varepsilon}_{cr} = A \sinh(\sigma / \sigma_0)$
Dorn, 1955	$\dot{\varepsilon}_{cr} = A \exp(\sigma / \sigma_0)$
JHK, 1963	$\dot{\varepsilon}_{cr} = A_1(\sigma / \sigma_0)^{n_1} + A_2(\sigma / \sigma_0)^{n_2}$
Garofalo, 1965	$\dot{\varepsilon}_{cr} = A\{\sinh(\sigma / \sigma_0)\}^n$



Figure: Turbine blade inspection

CDM models

Strain rate and Min. Strain rate	Damage Rate and Rupture Life
Kachanov-Rabotnov (KR) model	
$\dot{\epsilon}_{cr} = A \left(\frac{\sigma}{1-\omega} \right)^n$ $\dot{\epsilon}_{min} = A\sigma^n$ <p>$A, n =$ Norton power law constants</p>	$\dot{\omega} = \frac{M_K \sigma^\chi}{(1-\omega)^{\phi_K}}$ $t_r = [(\phi_K + 1)M_K \sigma^\chi]^{-1}$ <p>$M_K, \chi, \phi_K =$ tertiary creep damage constants</p>
Liu-Murakami (LM) model	
$\dot{\epsilon}_{cr} = A\sigma^n \exp(\rho\omega^{3/2})$ $\dot{\epsilon}_{min} = A\sigma^n$ $\rho = (2n+2) / (\pi\sqrt{1+3/n})$	$\dot{\omega} = \frac{M_L [1 - \exp(-\phi_L)]}{\phi_L} \sigma^q \exp(\phi_L \omega)$ $t_r = [M_L \sigma^q]^{-1}$ <p>$M_L, q, \phi_L =$ tertiary creep damage constants</p>
Sin-hyperbolic (Sinh) model	
$\dot{\epsilon} = B \sinh(\sigma/\sigma_s) \exp(\lambda\omega^{3/2})$ $\dot{\epsilon}_{min} = B \sinh(\sigma/\sigma_s)$ <p>$B, \sigma_s =$ secondary creep constants</p> $\lambda = \ln(\dot{\epsilon}_{final} / \dot{\epsilon}_{min})$	$\dot{\omega} = \frac{M_S [1 - \exp(-\phi_S)]}{\phi_S} \sinh\left(\frac{\sigma}{\sigma_t}\right)^\chi \exp(\phi_S \omega)$ $t_r = \left[M_S \sinh\left(\frac{\sigma}{\sigma_t}\right)^\chi \right]^{-1}$ <p>$M_S, \sigma_t, \phi_S =$ tertiary creep damage constants</p>

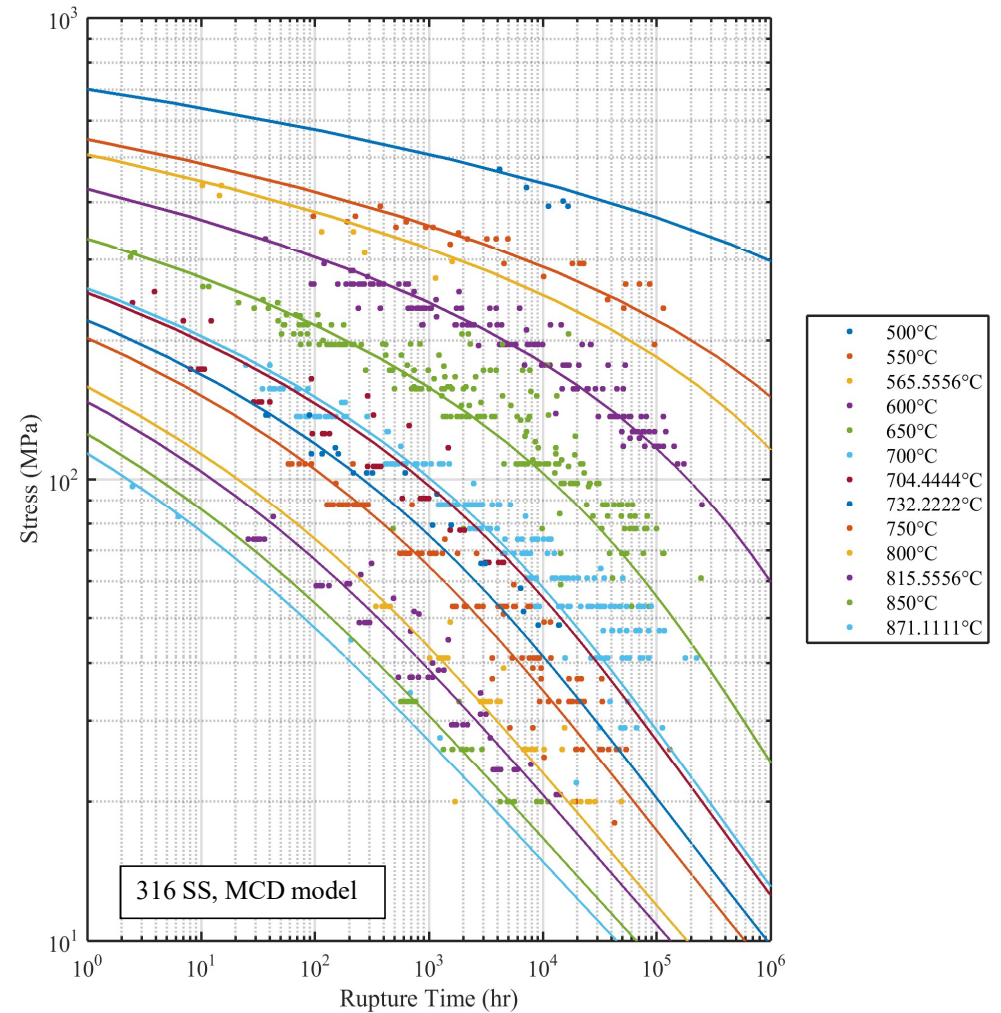
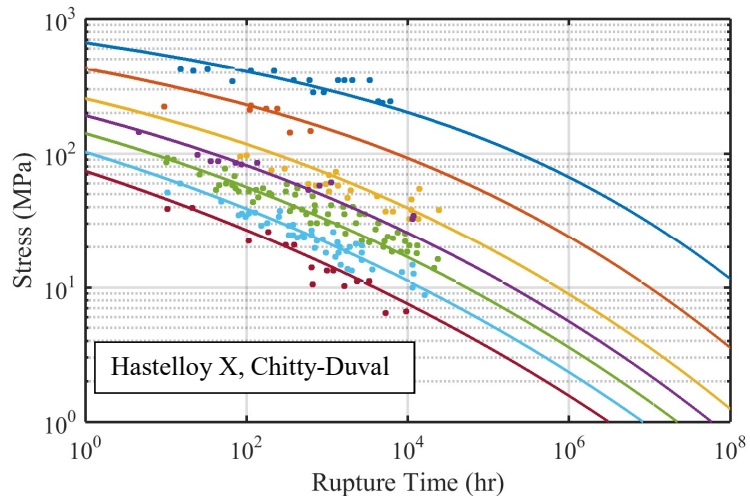
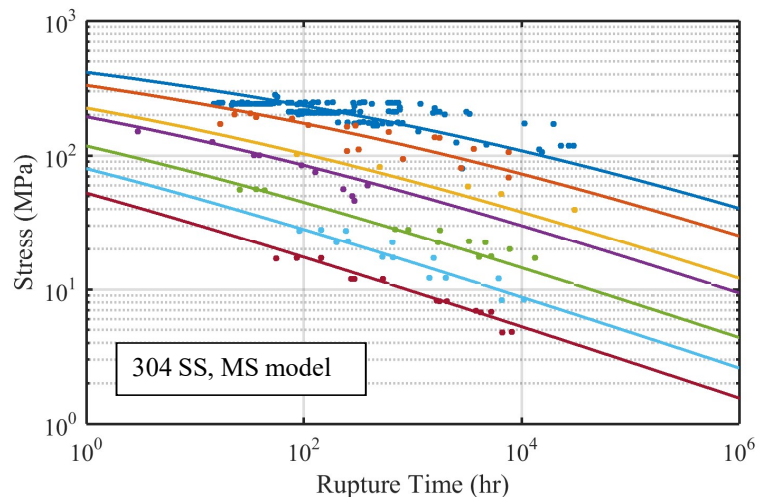
Representative Stress Functions

Sodbyrev	$\sigma_{rep} = \alpha\sigma_1 + (1-\alpha)\sigma_{vm}, \quad 0 \leq \alpha \leq 1$
Hydrostatic	$\sigma_{rep} = 3\beta\sigma_m + (1-\beta)\sigma_{vm}, \quad 0 \leq \beta < 1$
Hayhurst	$\sigma_{rep} = \alpha\sigma_1 + 3\beta\sigma_m + (1-\alpha-\beta)\sigma_{vm}$ $0 \leq \alpha + \beta \leq 1$

Dyson, Webster and Cane	$\sigma_{rep} = \left(\frac{\sigma_1}{\sigma_{VM}}\right)^{\gamma/\nu} \sigma_{VM}, \quad 0 \leq \gamma \leq \nu$
Hydrostatic	$\sigma_{rep} = \left(\frac{3\sigma_m}{\sigma_{VM}}\right)^{\gamma/\nu} \sigma_{VM}, \quad 0 \leq \gamma < \nu$
Combined	$\sigma_{rep} = \left(\frac{\sigma_1}{\sigma_{VM}}\right)^{\gamma/\nu} \left(\frac{3\sigma_m}{\sigma_{VM}}\right)^{\mu/\nu} \sigma_{VM}, \quad 0 \leq \gamma + \mu < \nu$

Huddleston	$\sigma_{rep} = \frac{3}{2} S_1 \left(\frac{2\sigma_{VM}}{3S_1} \right)^a \exp \left[b \cdot \left(\frac{J_1}{S_s} - 1 \right) \right]$ $S_s = \sqrt{\sigma_1^2 + \sigma_2^2 + \sigma_3^2}, \quad (S_1 = \sigma_1 - J_1/3), \quad (J_1 = \sigma_1 + \sigma_2 + \sigma_3)$
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Analytic fit (Stress-rupture)



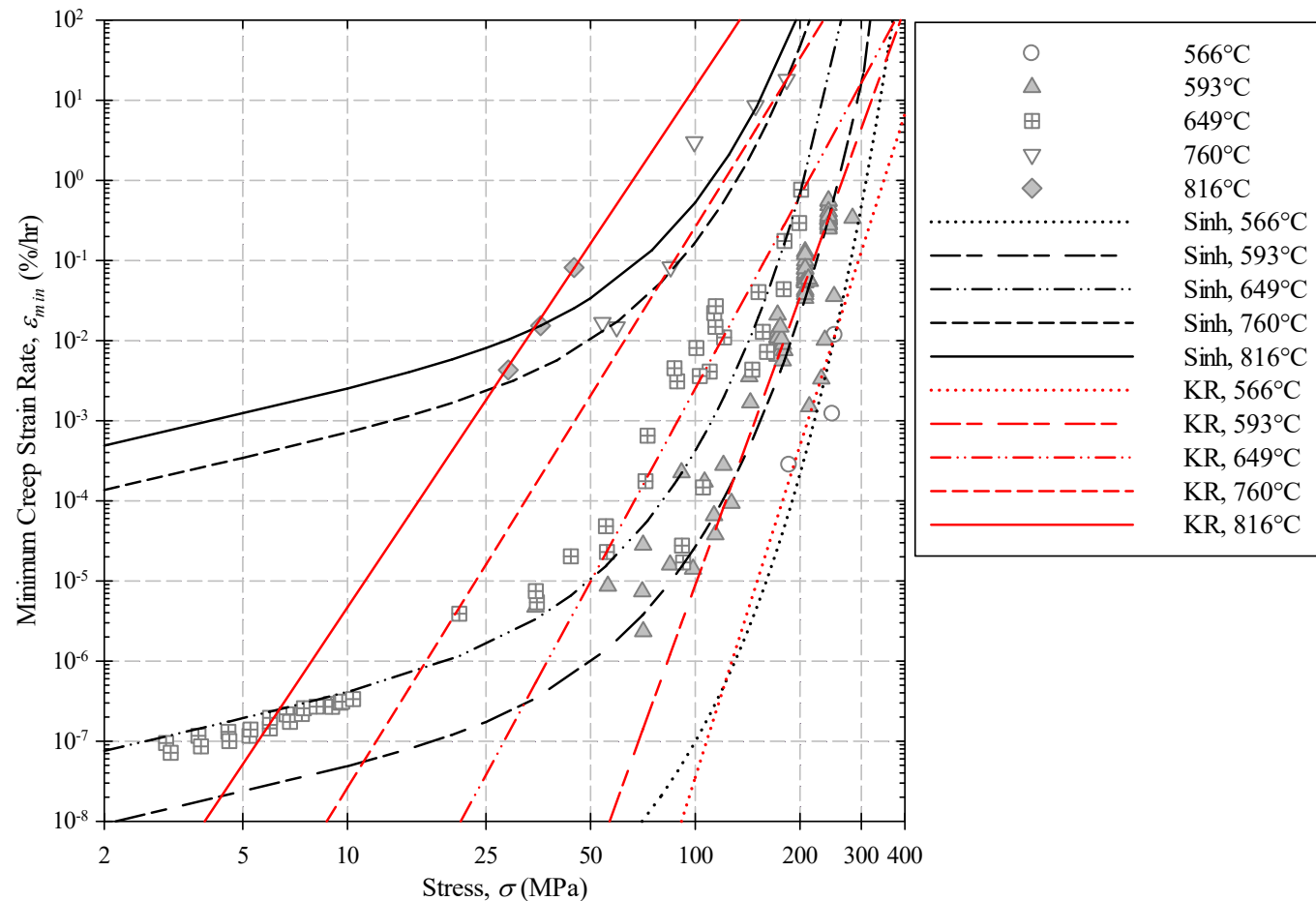
Minimum Creep Strain Rate

Norton Power Law (KR) $\dot{\epsilon}_{\min} = A\sigma^n$

McVetty (Sinh) $\dot{\epsilon}_c = B \sinh(\sigma/\sigma_s)$

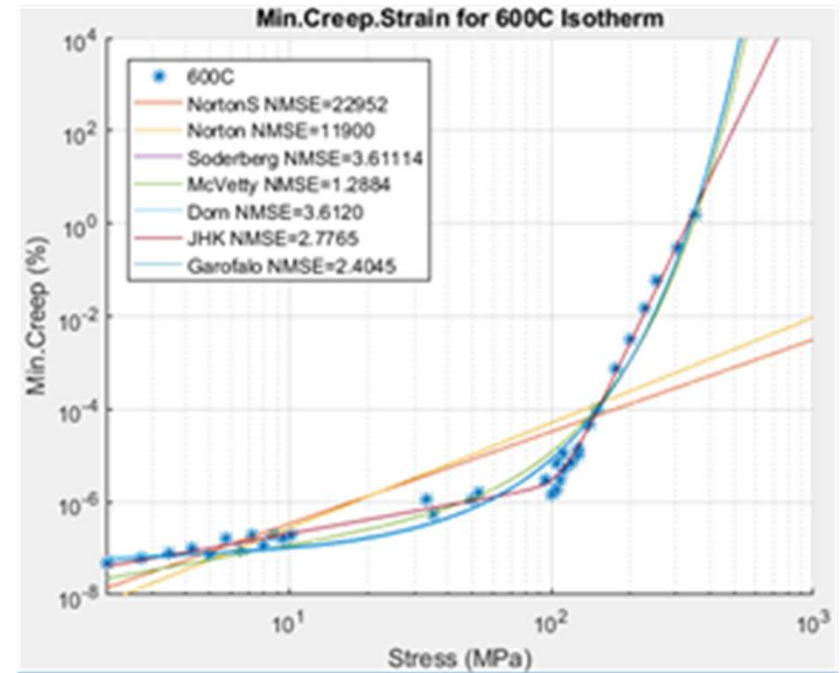
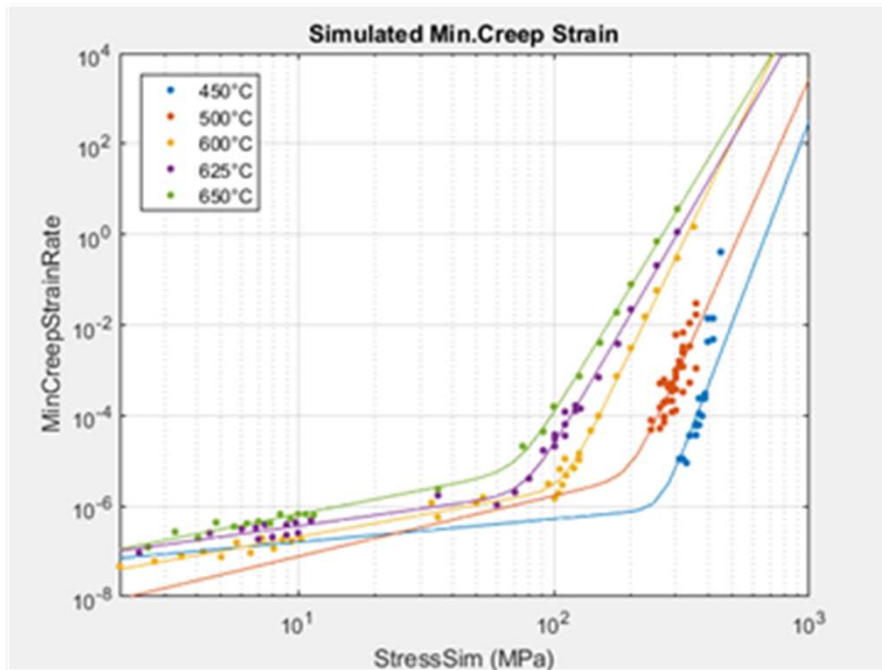
KR limitations: The KR minimum creep strain rate predictions are linear on a log-log scale and thus are not able to accurately model the sigmoidal behavior observed in the experimental data.

Sinh advantage: The Sinh minimum creep strain rate predictions bend on a log-log scale and are able to accurately model the sigmoidal behavior.



Comparison of Norton and McVetty model against 304 SS data

Minimum Creep Strain Rate



Comparison of various data model against P91 data

Creep deformation

Comparison of Theta, Omega, and Sinh model against **Hastelloy X** creep data

Theta Projection (tertiary strain rate)

$$\varepsilon = \theta_1[1 - \exp(-\theta_2 t)] + \theta_3[\exp(\theta_4 t) - 1]$$

$$\dot{\varepsilon}_t = \theta_3 \theta_4 \exp(\theta_4 t)$$

Omega Model

$$\dot{\varepsilon} = \dot{\varepsilon}_0 \exp(\varepsilon \Omega)$$

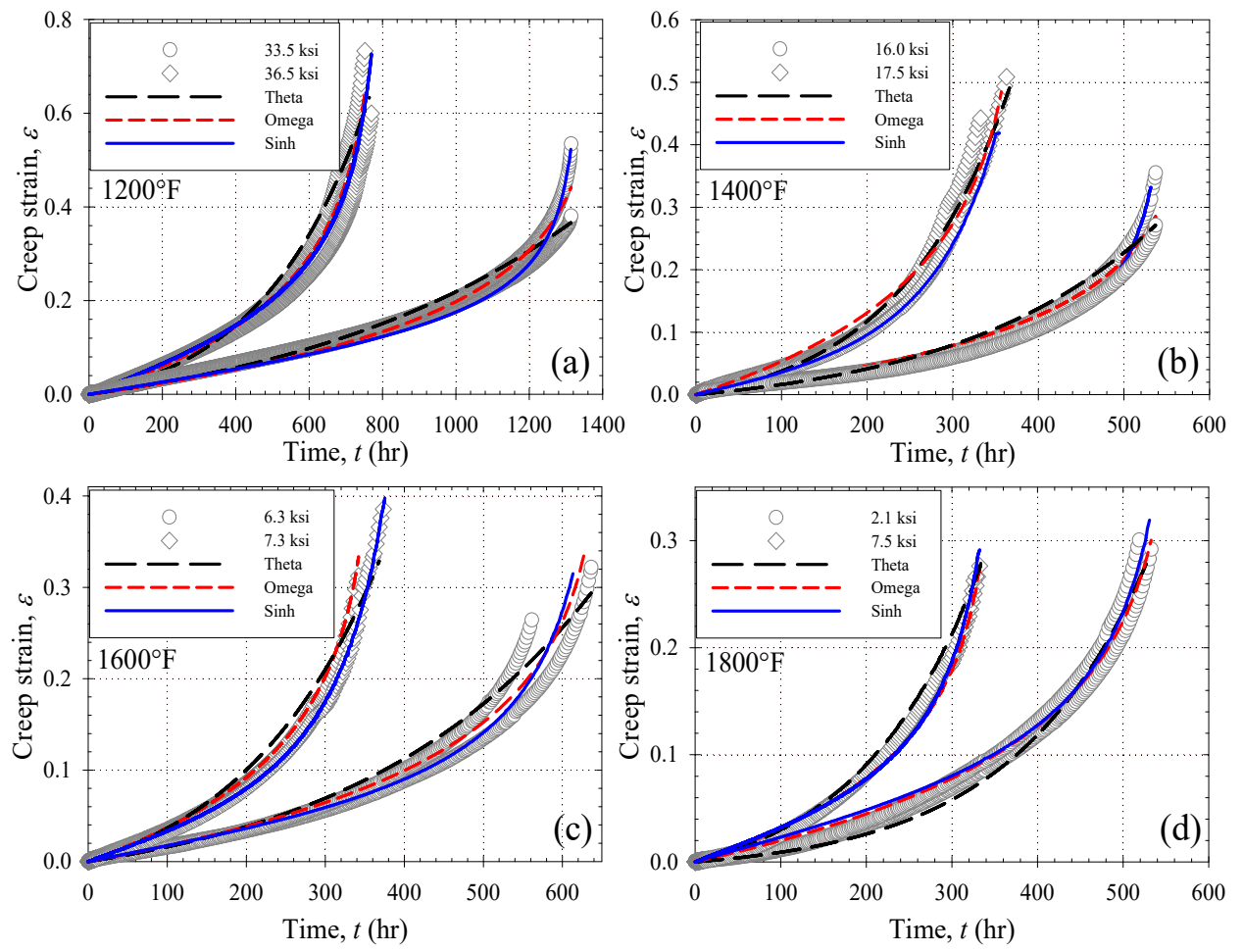
$$\omega = \frac{t}{t_r} = \frac{\dot{\varepsilon} \Omega t}{1 + \dot{\varepsilon} \Omega t}$$

Sin-Hyperbolic Model

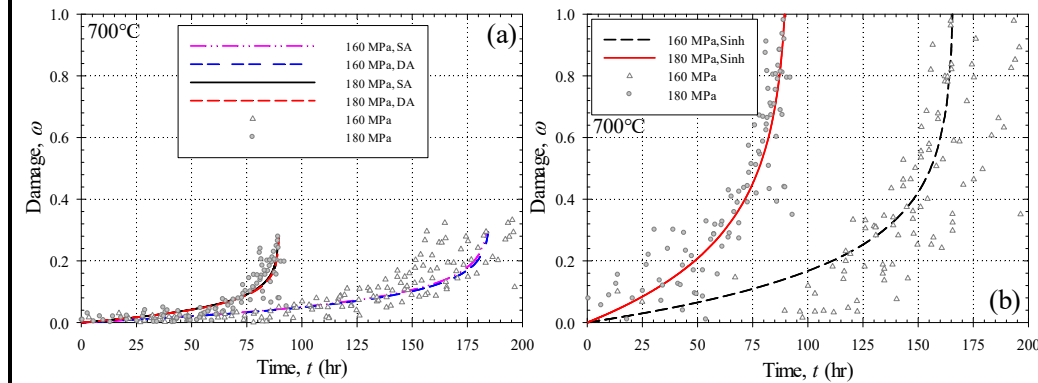
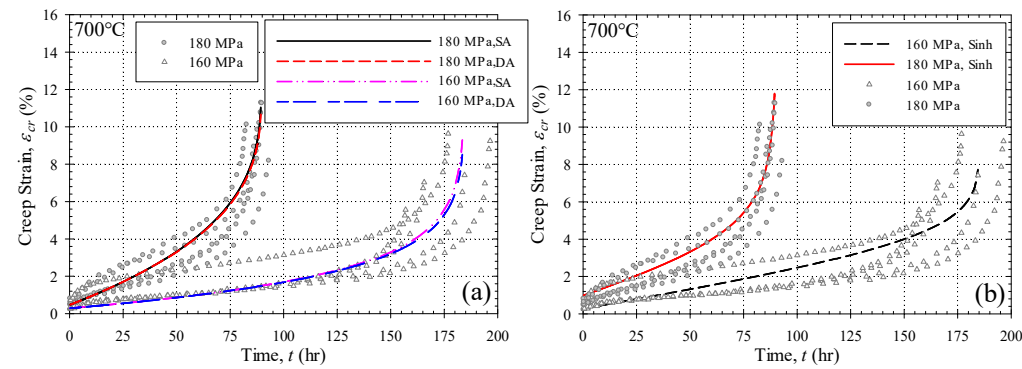
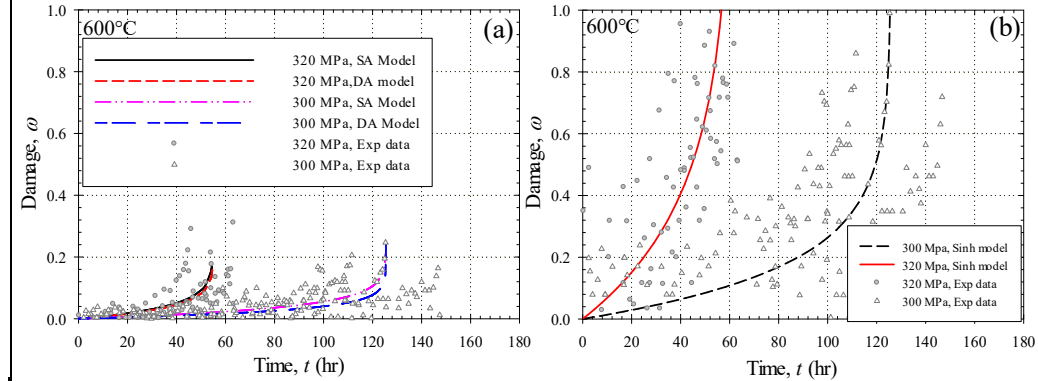
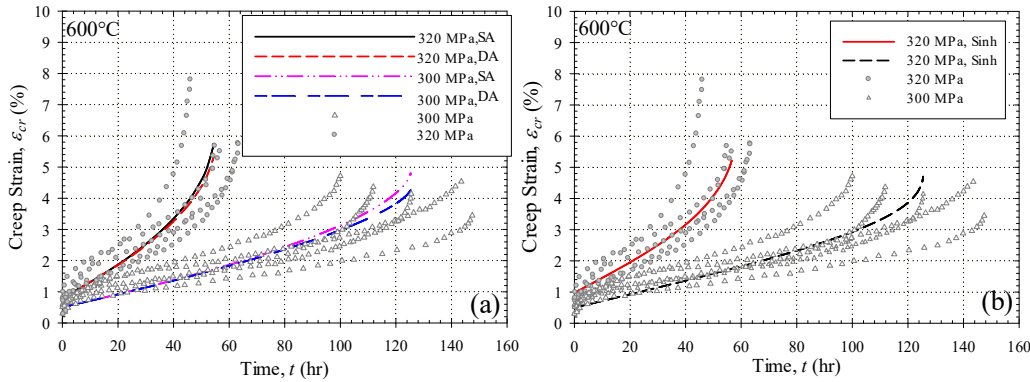
$$\dot{\varepsilon} = A \sinh(\sigma / \sigma_s) \exp(\lambda \omega^{3/2})$$

$$\dot{\omega} = \frac{M [1 - \exp(-\phi)]}{\phi} \sinh\left(\frac{\sigma}{\sigma_t}\right)^\chi \exp(\phi \omega)$$

Ref: [4]



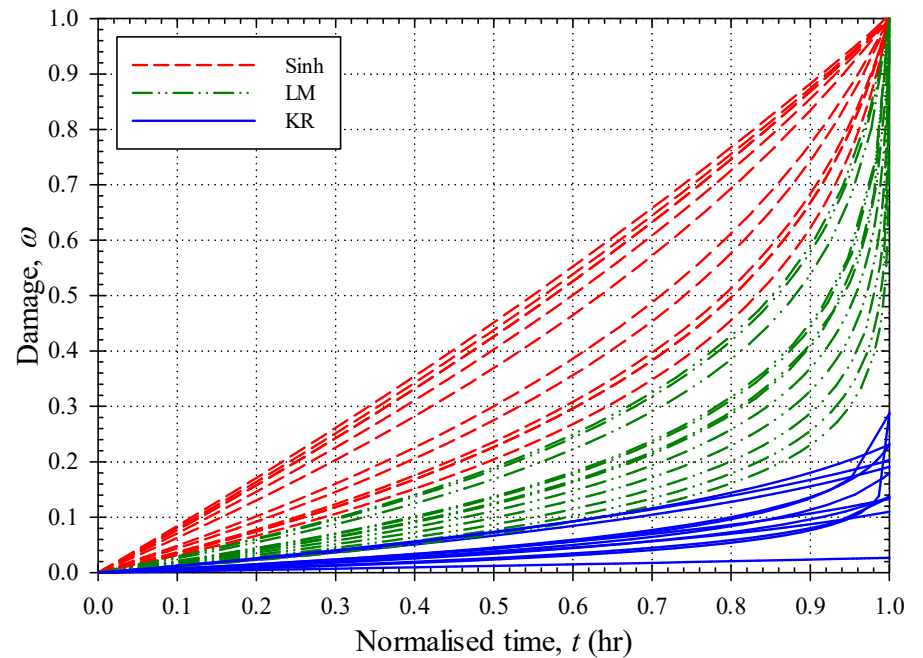
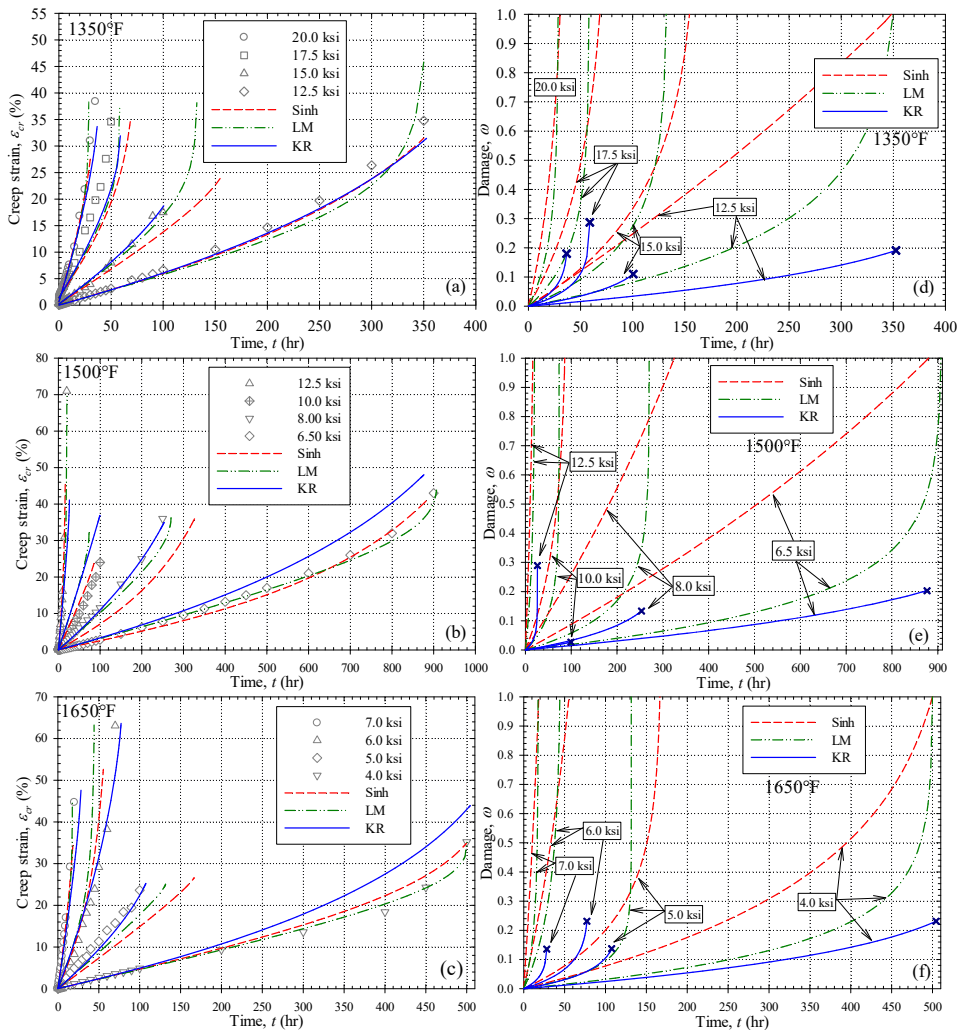
CDM model fit



Comparison of Sinh and KR model against 304 SS creep strain data

Comparison of Sinh and KR model damage evolution against 304 SS analytic damage

CDM model fit



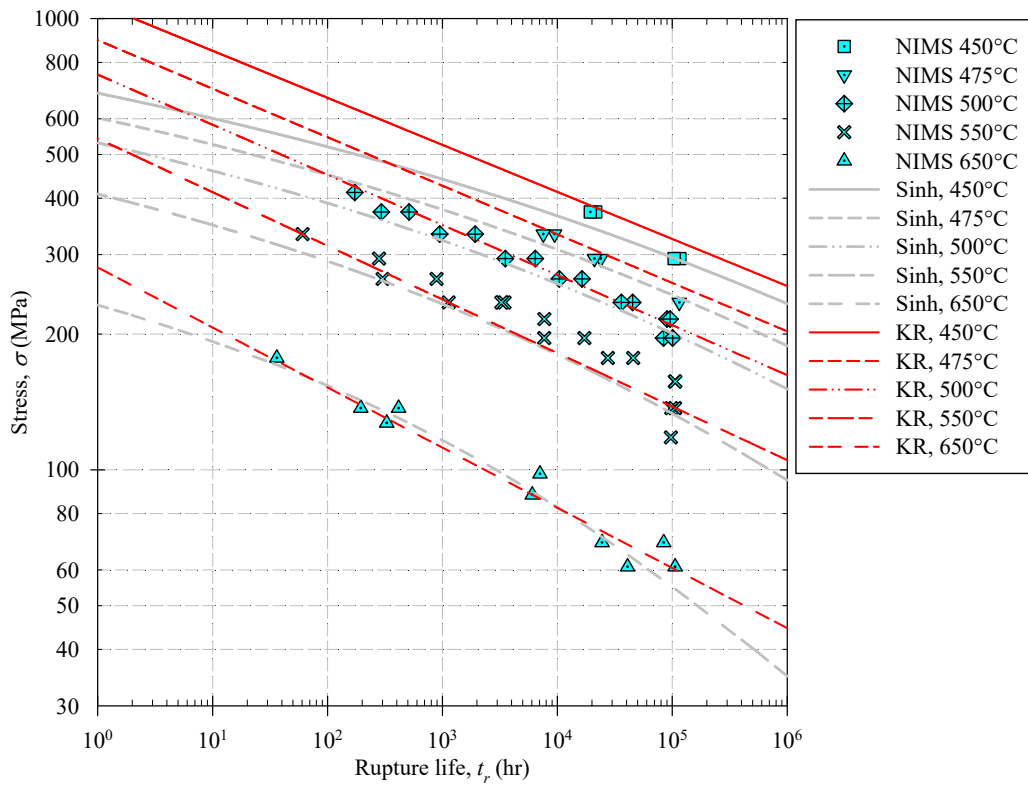
Comparison of Sinh, LM and KR model against 316 SS creep strain data

[1] Haque, M. S., and Stewart, C. M. (2015). Comparison of a New Sin-Hyperbolic Creep Damage Constitutive Model With the Classic Kachanov-Rabotnov Model Using Theoretical and Numerical Analysis. In *TMS 2015 144th Annual Meeting & Exhibition* (pp. 937-945). Springer, Cham.

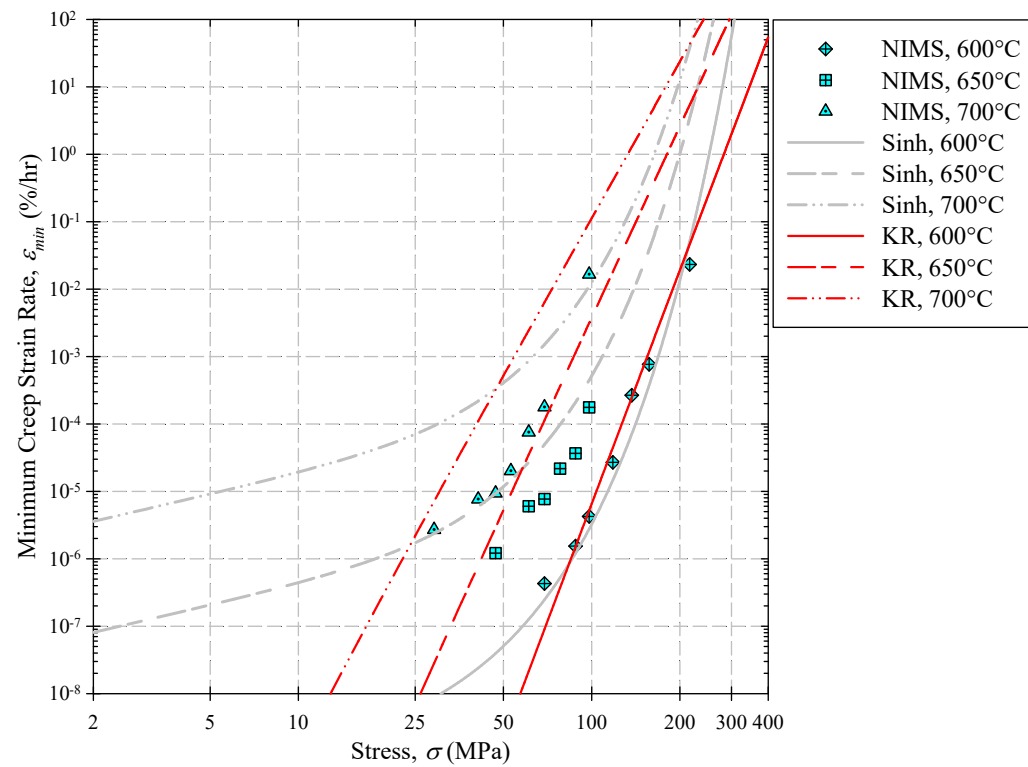
[14] Haque, M. S., and Stewart, C. M. (2017). Metamodeling of the Kachanov-Rabotnov, Liu-Murakami, and Sin-Hyperbolic Continuum Damage Mechanics based Creep Models. **(40% complete)**

Task 5: Post-Audit Verification

Post-audit validation with additional data that is not used in calibration



Stress-Rupture



Minimum-creep-strain-rate

Task 6: Disparate Data problem

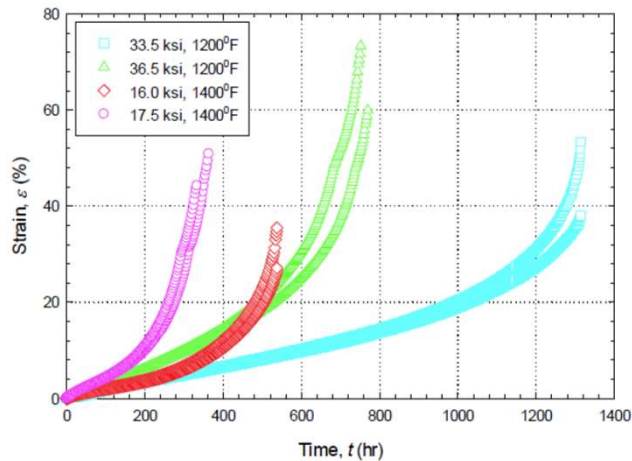
Gap exist in creep data.

Data is not available in the regime or form of interest.

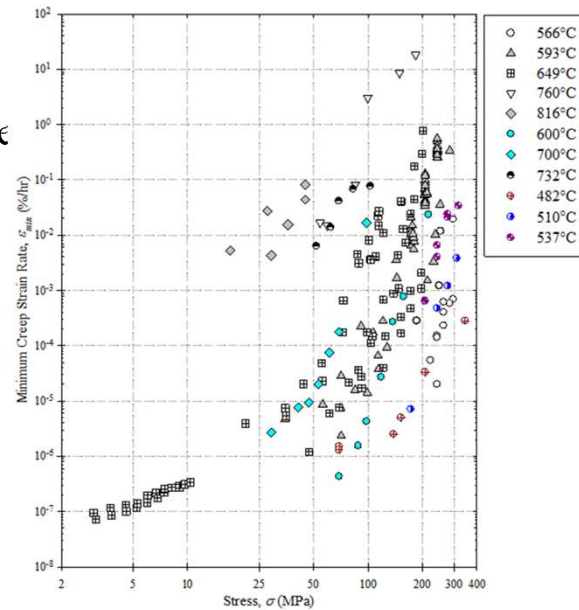
Creep deformation extrapolation using short-term data may lead to unreliable prediction.

Cross-calibration to any and every form of data can improve extrapolation.

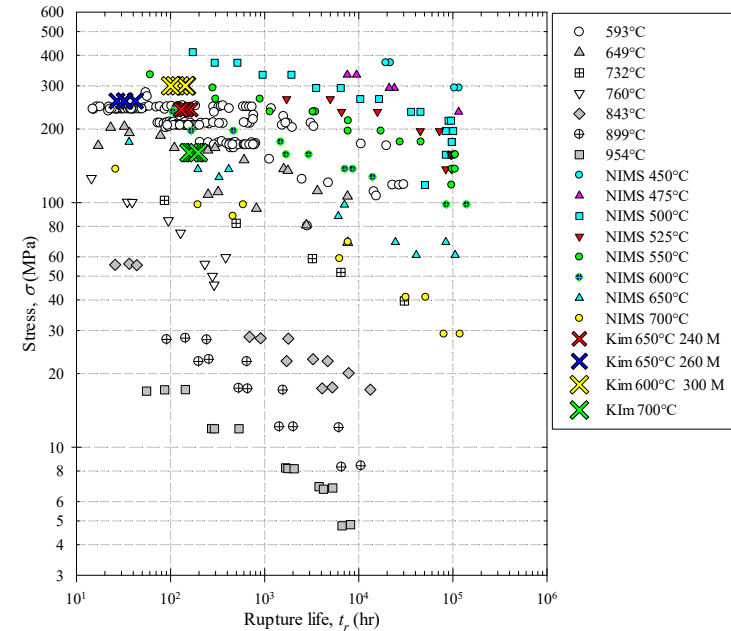
Data exists in diaspora
Unreliable extrapolation
Limited up to certain multiple of the longer



Creep deformation



Minimum-creep-strain-rate

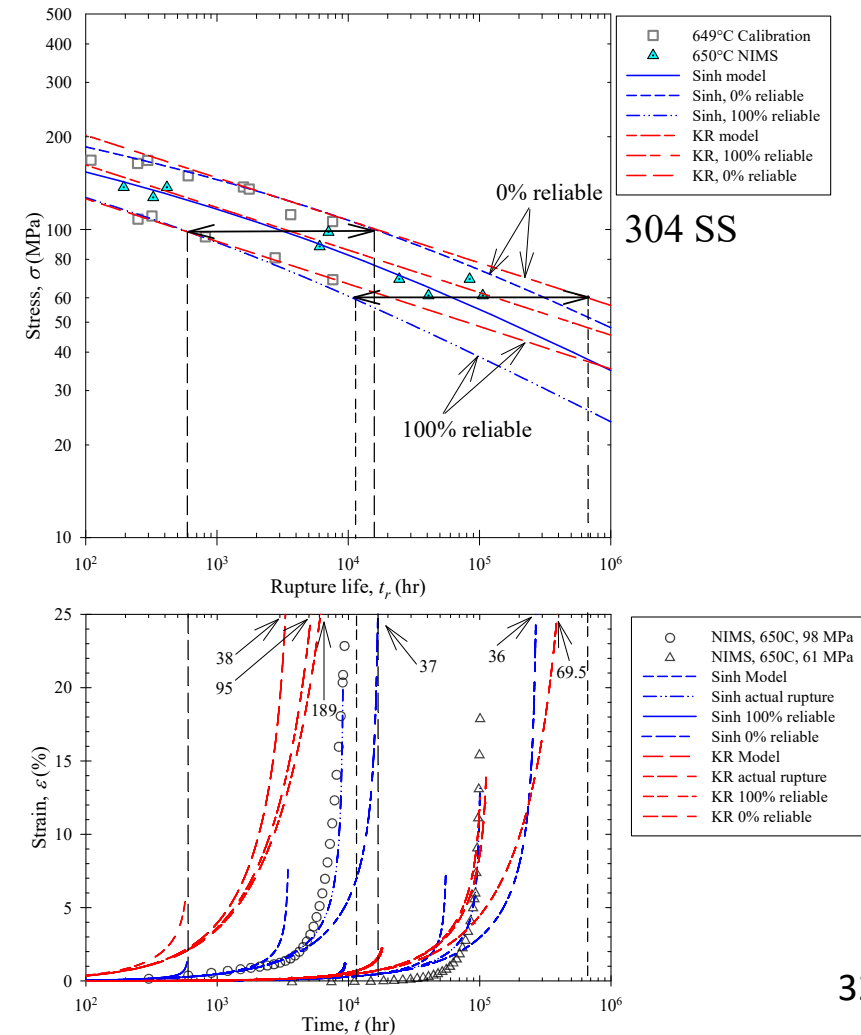


Stress-Rupture

Disparate Data problem: Improved Extrapolation

Calibration procedure for CDM models to disparate creep data

Model Selection	Select the creep model of interest. (Kachanov-Rabotnov, Sinh, etc.)
Segregation	Segregate the model into equations related to each type of creep data. (Minimum-Creep-Strain-Rate, Stress-Rupture, Creep Deformation, etc.)
Calibration	Calibrate the material constants of the equations using the creep data.
Regression	Use regression analysis to convert material constants into temperature-dependent functions. $f_i(T)$
Validation	Take the pre-calibrated model and compare it to additional data not used in the calibration process.
Design	Make interpolative and extrapolative prediction of creep behavior. Plot these predictions as Creep Design Maps.



Development of a Design Map

- Tensile properties
- The variable of interest
- Design envelope

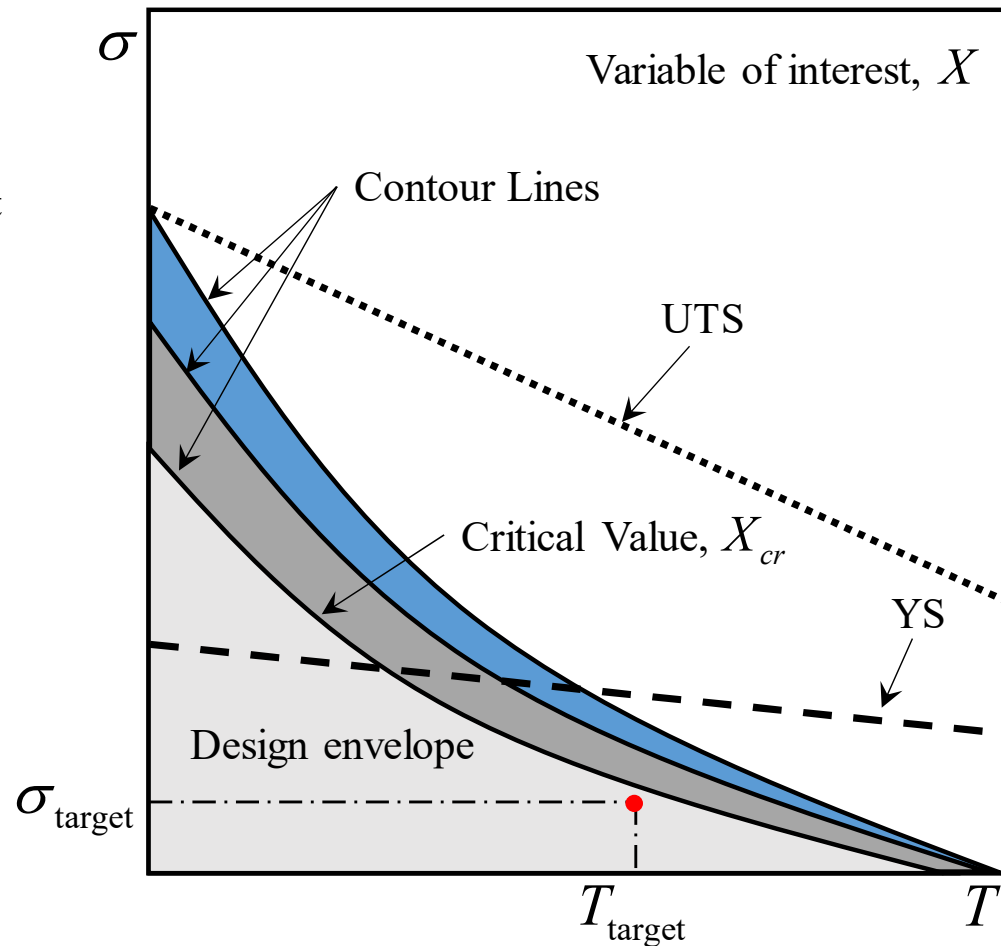
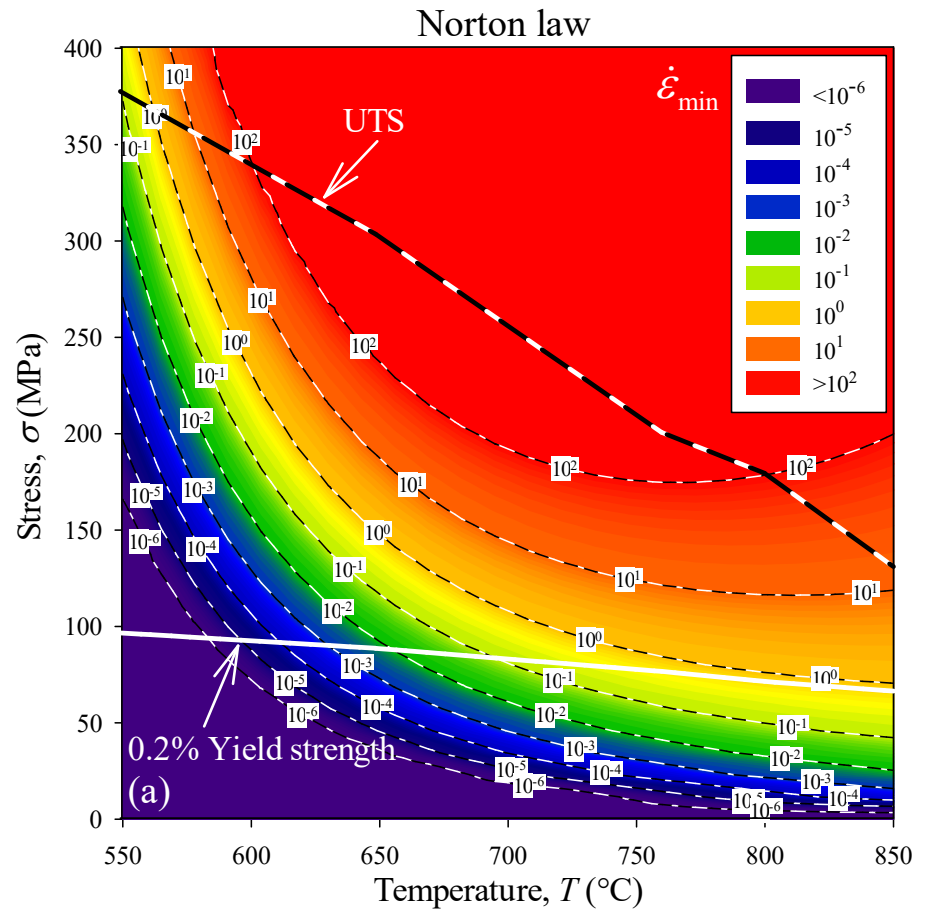
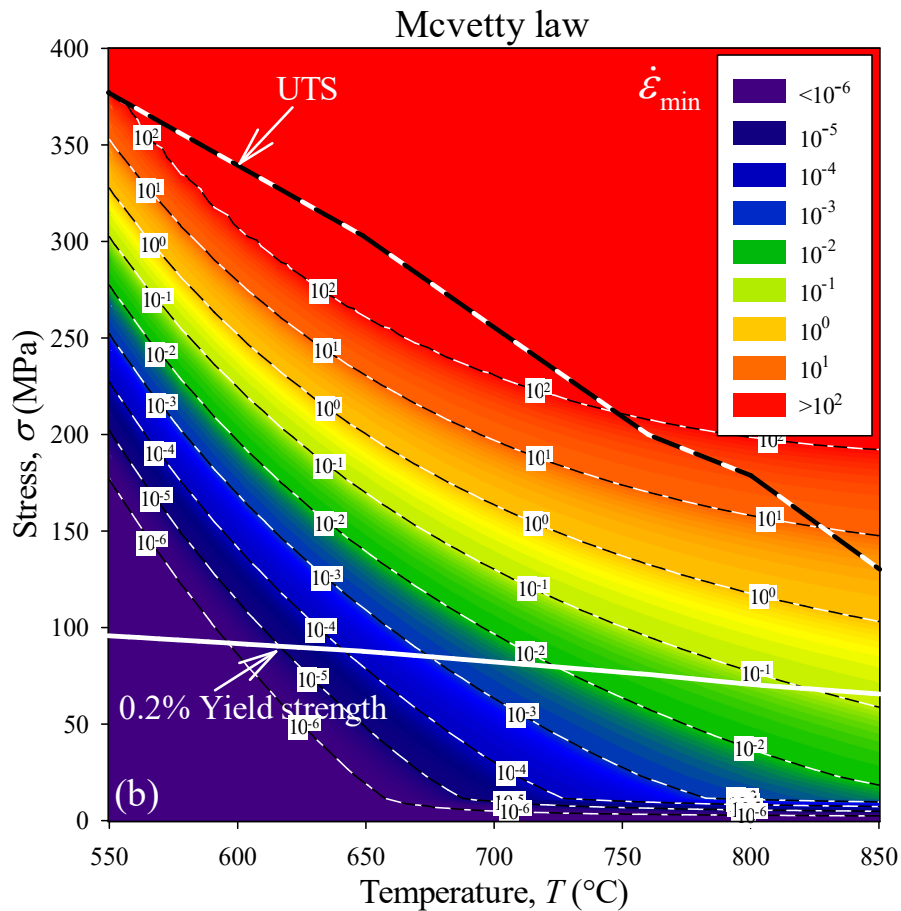


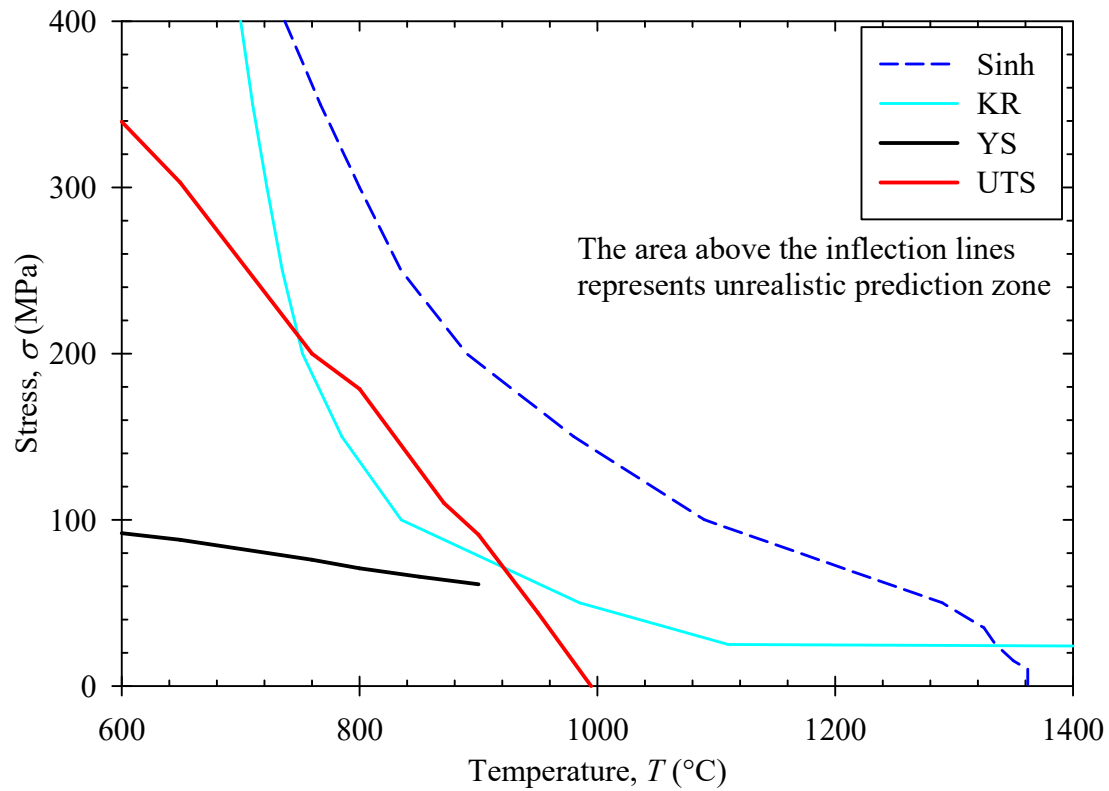
Illustration of design map

Extrapolation and Interpolation: Design Maps



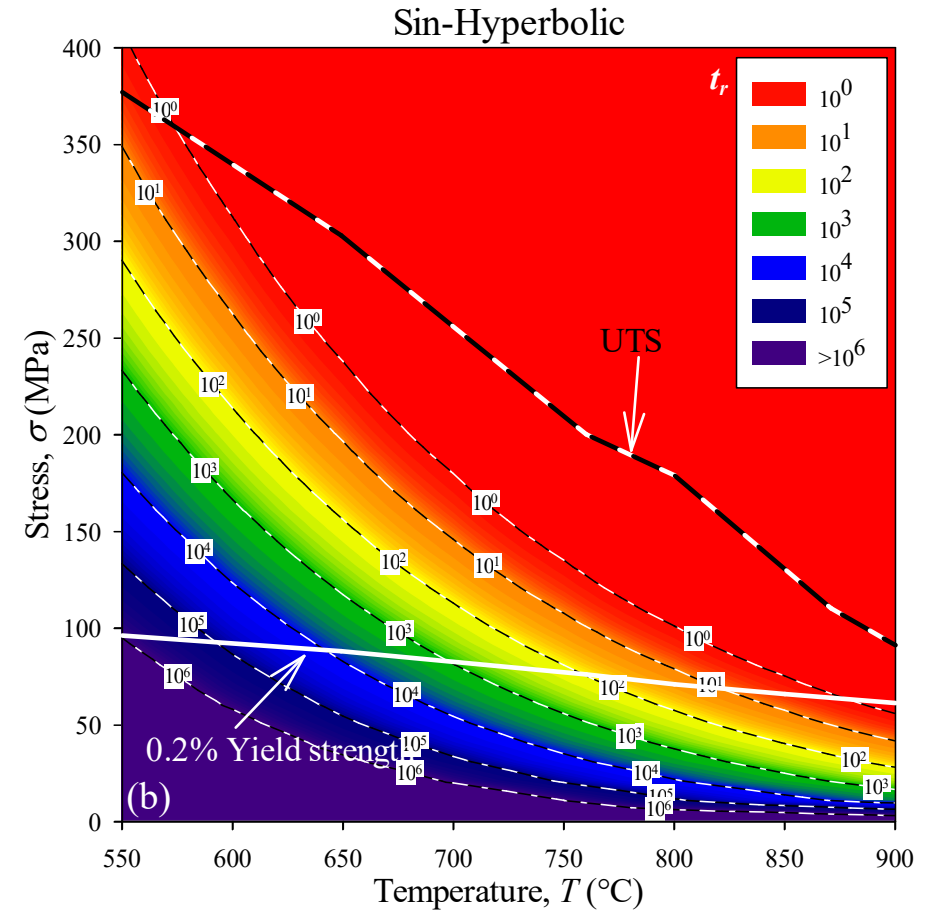
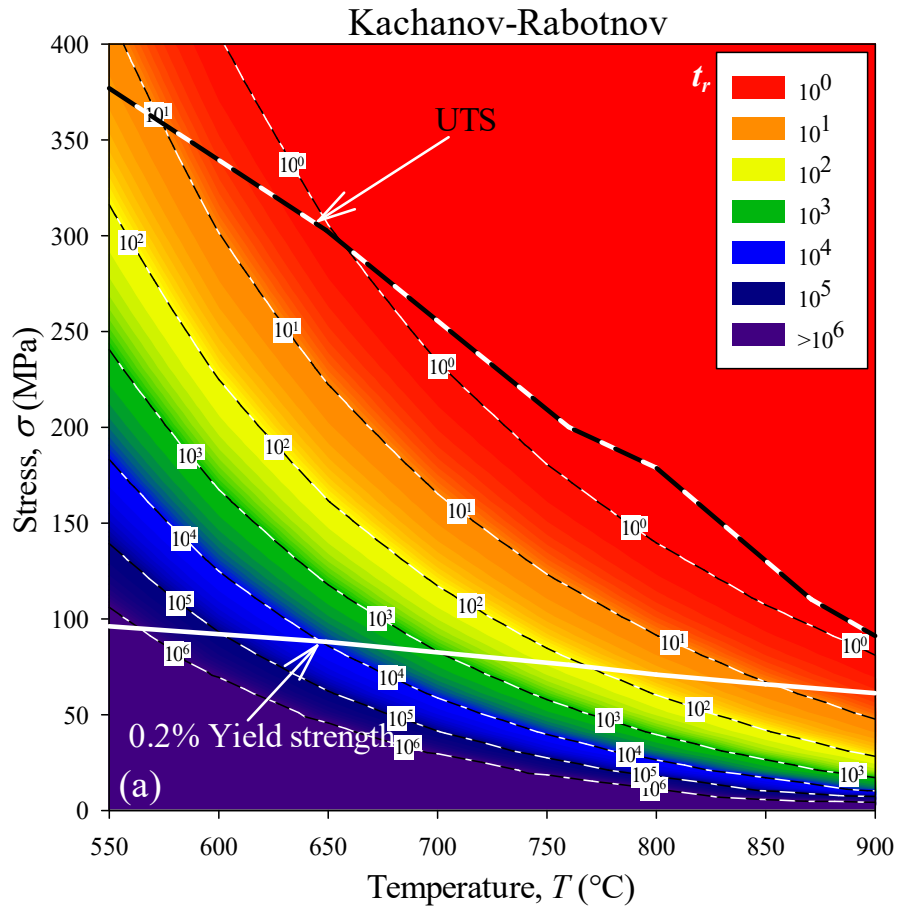
Minimum-creep-strain-rate design maps

Extrapolation and Interpolation: Design Maps



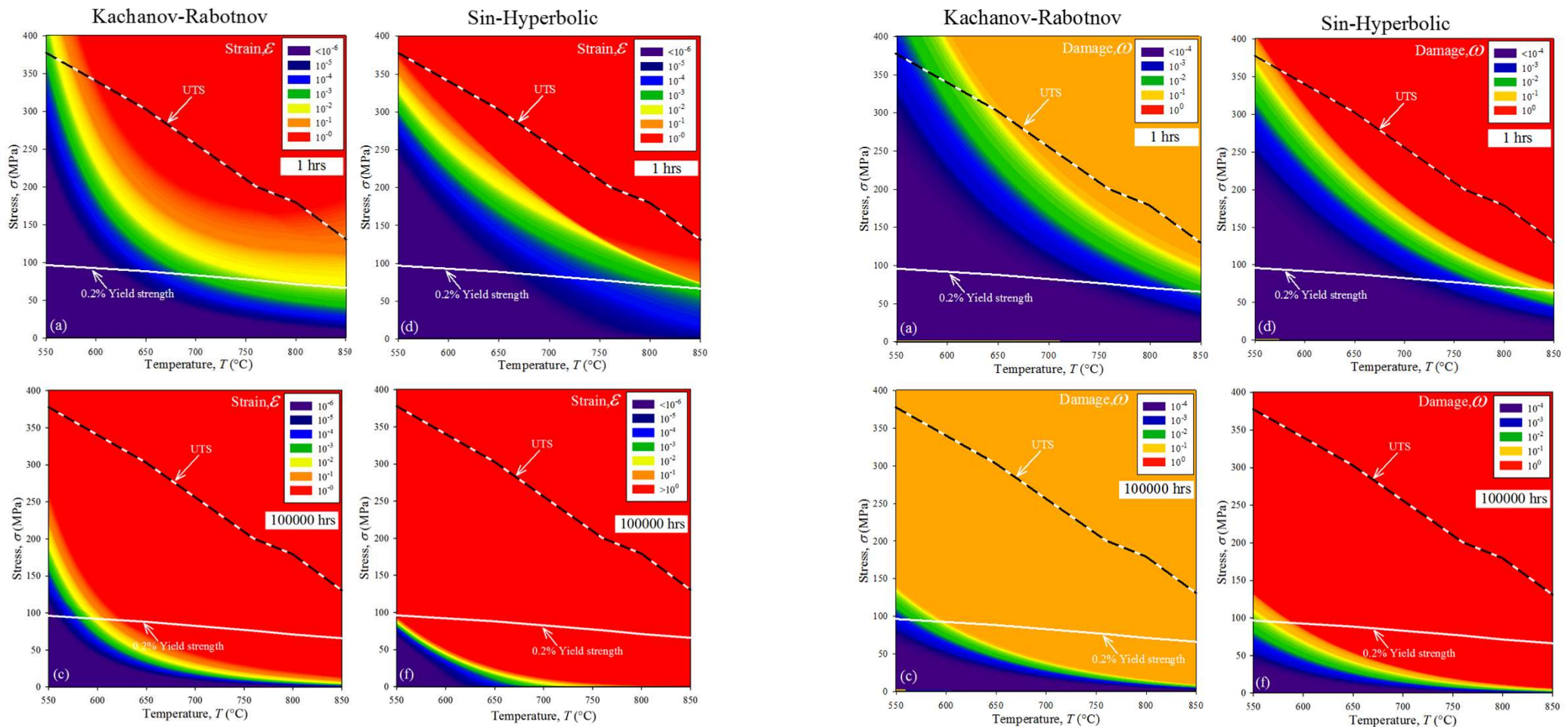
Minimum-creep-strain-rate Inflection

Design Maps



Stress-Rupture design maps

Design Maps



Deformation design maps

Damage design maps

Task 7: Metamodeling: Finding the “best ” model

A “**metamodel**” can be described as a combinational model, derived from rearranging, modifying, and/or expanding the functional relationships between different models.

Stress-Rupture

- **Eight** commonly used TTP models
- **Four** newly developed model

Creep-deformation

- Omega model
- Theta Projection
- Sin-hyperbolic model

Continuum Damage Mechanics

- Kachanov-Rabotnov
- Sin-hyperbolic
- Liu-Murakami

- A detail explanation of stress-rupture metamodeling will be presented.
- Summary and progress of the other metamodeling group is reported.

Metamodeling: Stress-Rupture

- Metamodel:

$$P_{HS} = \frac{\log(t_r) - \alpha_0 - \alpha_1 T^r}{(T^r - \alpha_2^r)^q}$$

- Parent model:
Larson-Miller

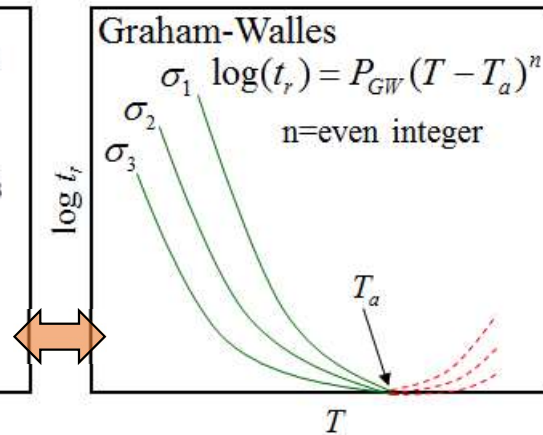
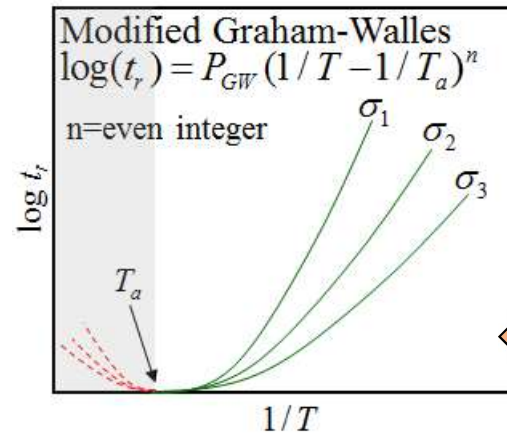
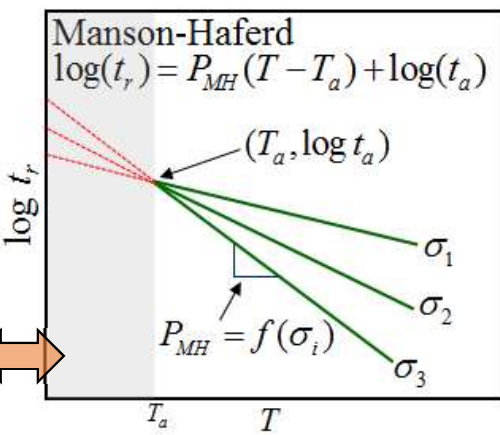
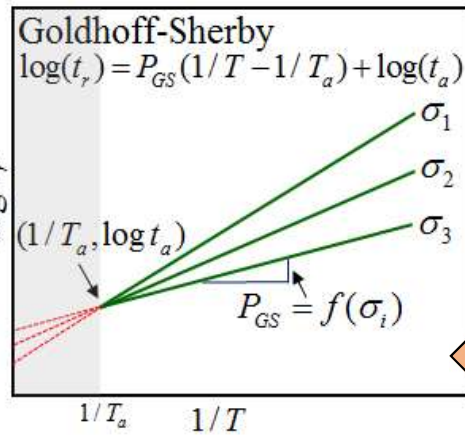
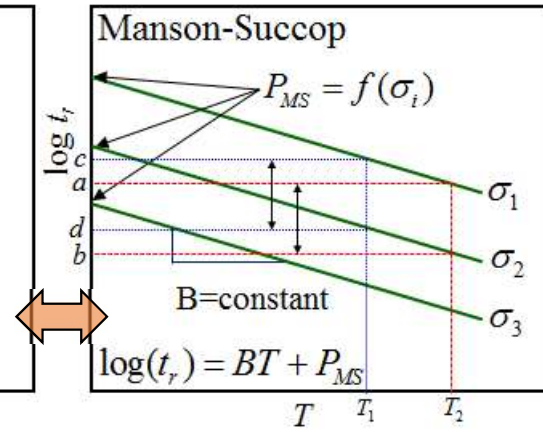
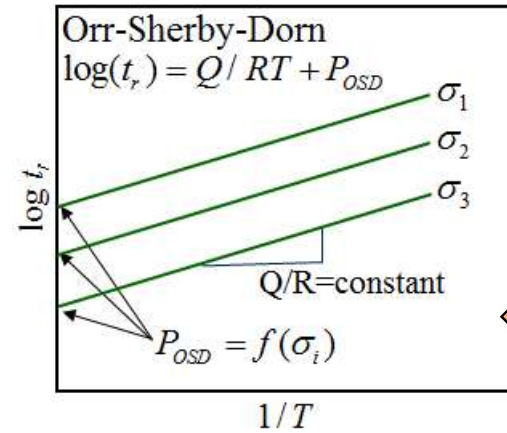
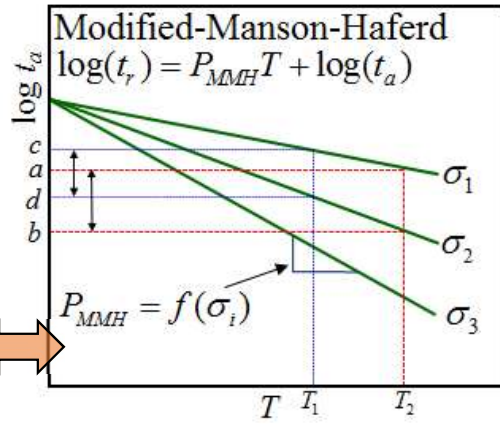
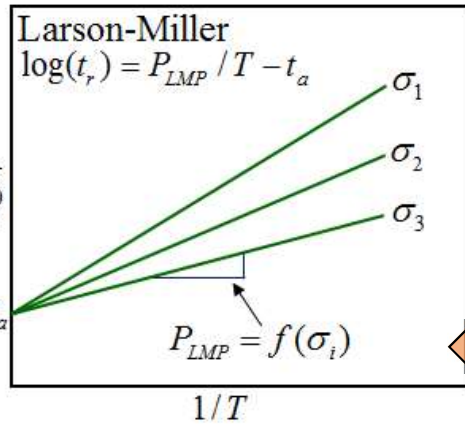
$$P_{LMP} = T(\log(t_r) + t_a)$$

$$P_{unified} = \frac{\log(t_r) - \alpha_0 - \alpha_1 T^r}{(T^r - \alpha_2^r)^q}$$

Model	Year	Parametric equation	Iso-stress equation	Material constants	Characteristics	Metamodel condition	Ref.
Larson-Miller	1952	$P_{LMP} = T(\log(t_r) + t_a)$	$\log(t_r) = \frac{P_{LMP}}{T} - t_a$	$P_{LMP}(\sigma), t_a$	Cy, L, NP, Sp	$\alpha_2 = \alpha_1 = 0$ $r = -1, q = 1$	11
Manson-Haferd	1953	$P_{MH} = \frac{\log(t_r) - \log(t_a)}{T - T_a}$	$\log(t_r) = P_{MH}(T - T_a) + \log(t_a)$	$P_{MH}(\sigma), T_a, t_a$	Cxy, L, NP, Sn	$\alpha_1 = 0$ $r = q = 1$	3
Manson-Brown	1953	$P_{MB} = \frac{\log(t_r) - \log(t_a)}{(T - T_a)^n}$	$\log(t_r) = P_{MB}(T - T_a)^n + \log(t_a)$	$P_{MB}(\sigma), T_a, t_a, n$	Cxy, NL, NP, Sny	$\alpha_1 = 0, r = 1,$ $q = n$	12
Orr-Sherby-Dorn	1954	$P_{OSD} = \log(t_r) - Q/RT$	$\log(t_r) = Q/RT + P_{OSD}$	$P_{OSD}(\sigma), Q, R$	NC, L, P, Sp=constant	$\alpha_0 = 0, r = -1,$ $q = 0$	13
Manson-Succop	1959	$P_{MS} = \log(t_r) - BT$	$\log(t_r) = BT + P_{MS}$	$P_{MS}(\sigma), B$	NC, L, P, Sn=constant	$\alpha_0 = \alpha_2 = 0,$ $r = 1, q = 0$	14
Graham-Walles	1955	$P_{GW} = \frac{\log(t_r)}{(T - T_a)^n}$	$\log(t_r) = P_{GW}(T - T_a)^n$	$P_{GW}(\sigma), T_a, n$	Cx, NL, NP, Sny	$\alpha_0 = \alpha_1 = 0,$ $r = 1, q = n$	15
Chitty-Duval	1963	$P_{CD} = mT - \log(t_r)$	$\log(t_r) = mT - P_{CD}$	$P_{CD}, m = a\sigma^b$	NC, NP, L, Sn	$\alpha_0 = \alpha_2 = 0,$ $r = 1, q = 0$	16
Goldhoff-Sherby	1968	$P_{GS} = \frac{\log(t_r) - \log(t_a)}{1/T - 1/T_a}$	$\log(t_r) = P_{GS}(1/T - 1/T_a) + \log(t_a)$	$P_{GS}(\sigma), T_a, t_a$	Cxy, L, NP, Sp	$\alpha_1 = 0, r = -1,$ $q = 1$	17
Modified Manson-Haferd	--	$P_{MMH} = \frac{\log(t_r) - \log(t_a)}{T}$	$\log(t_r) = P_{MMH}T + \log(t_a)$	$P_{MMH}(\sigma), t_a$	Cy, L, NP, Sn	$\alpha_2 = \alpha_1 = 0,$ $r = 1, q = 1$	--
Modified Graham-Walles	--	$P_{MGW} = \frac{\log(t_r)}{(1/T - 1/T_a)^n}$	$\log(t_r) = P_{MGW}(1/T - 1/T_a)^n$	$P_{MGW}(\sigma), T_a, n$	Cx, NL, NP, Spv	$\alpha_0 = \alpha_1 = 0$ $r = -1, q = n$	--
Modified Chitty-Duval	--	$P_{CD} = \frac{m}{T} - \log(t_r)$	$\log(t_r) = m/T - P_{CD}$	$P_{CD}, m = a\sigma^b$	NC, NP, L, Sp	$r = -1, q = 0,$ $\alpha_0 = \alpha_2 = 0$	--
Modified-Goldhoff-Sherby	--	$P_{GS} = \frac{\log(t_r) - \log(t_a)}{(1/T - 1/T_a)^n}$	$\log(t_r) = P_{MGS}(1/T - 1/T_a)^n + \log(t_a)$	$P_{MGS}(\sigma), T_a, t_a$	Cxy, NL, NP, Sp	$r = -1, q = n,$ $\alpha_1 = 0$	--

New models

Mirror Models



Mirror Models (Cont.)

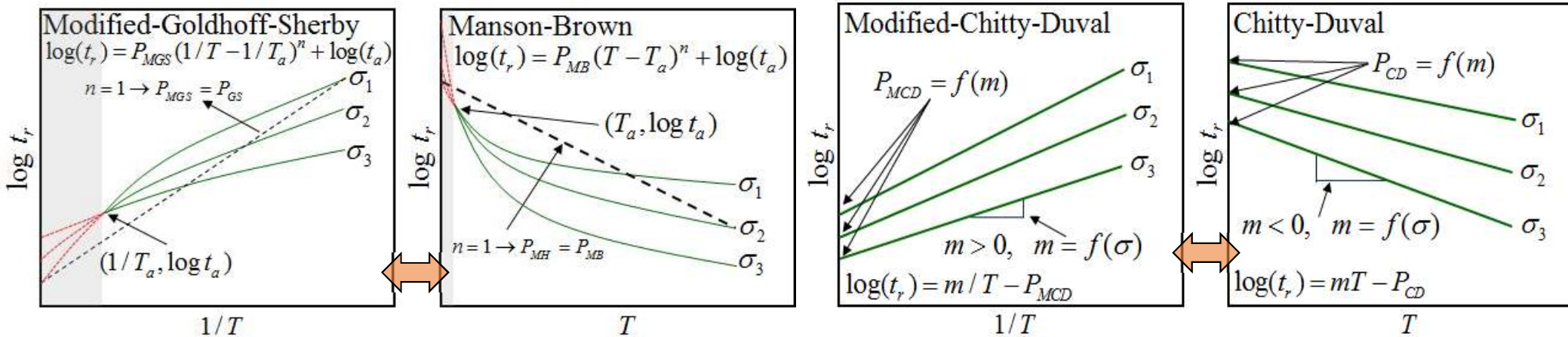
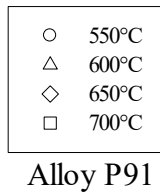


Table – List of the “mirror pairs” that have similar mathematical form whose isostress lines are inverse/mirror image due to the temperature variable T being inverted

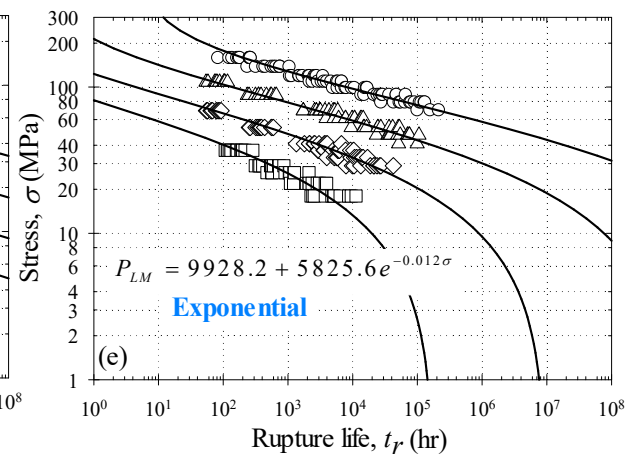
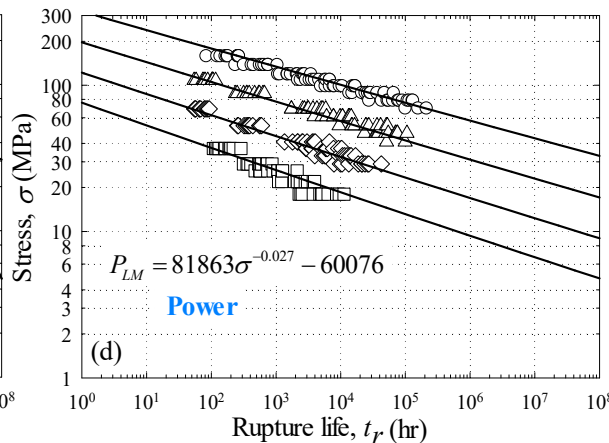
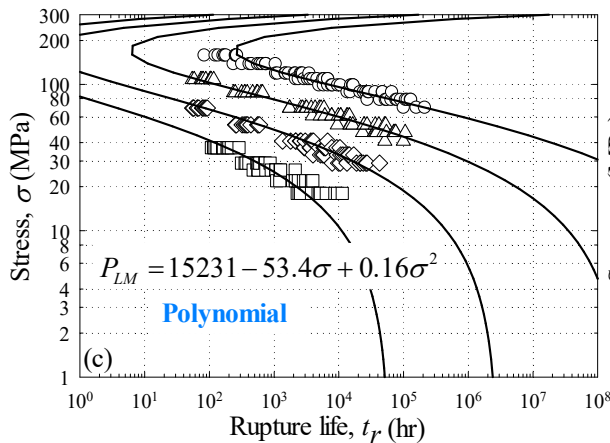
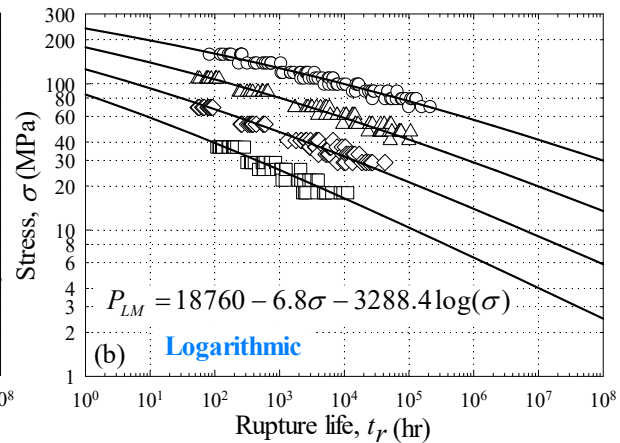
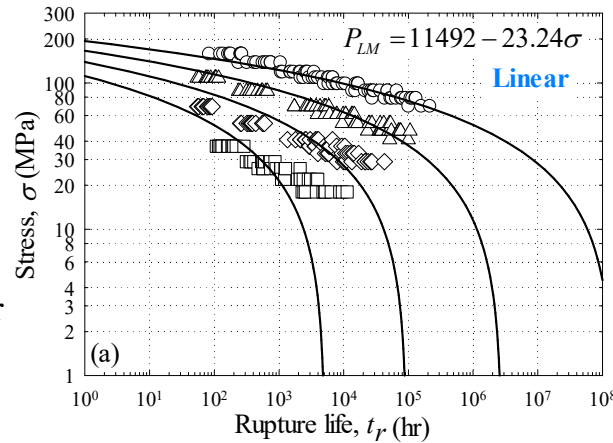
(1)	Larson-Miller	\leftrightarrow	Mod-Manson-Haferd
(2)	Goldhoff-Sherby	\leftrightarrow	Manson-Haferd
(3)	Mo-Goldhoff-Sherby	\leftrightarrow	Manson-Brown
(4)	Orr-Sherby-Dorn	\leftrightarrow	Manson-Succop
(5)	Mod-Graham-Walles	\leftrightarrow	Graham-Walles
(6)	Mod-Chitty-Duval	\leftrightarrow	Chitty-Duval

Stress-parameter function

- The **stress-parameter function** is a mathematical expression of the master curve such that a temperature invariant parameter can be obtained for a given stress.



Larson-Miller model fit using five Stress-parameter function



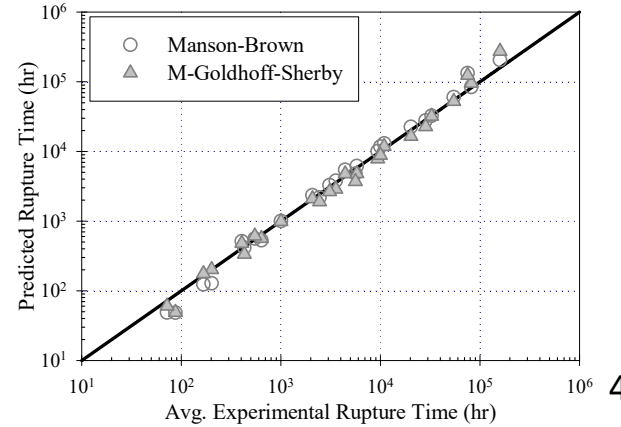
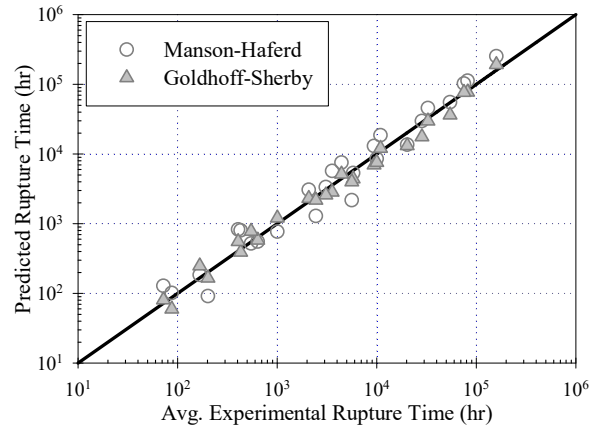
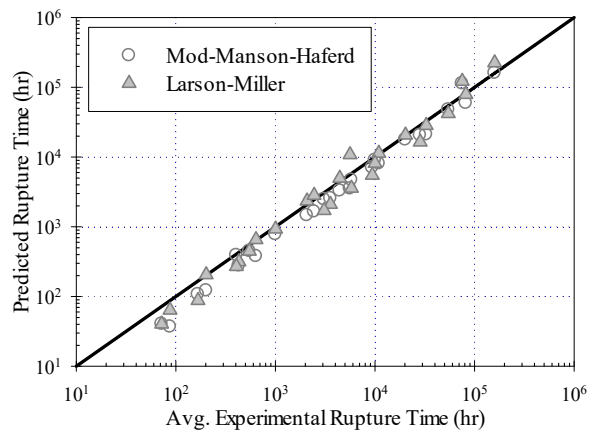
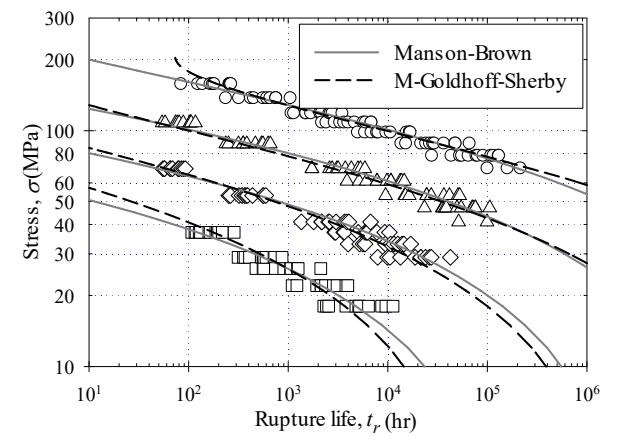
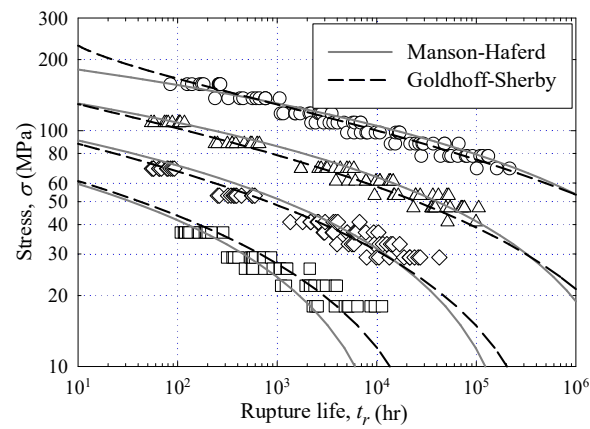
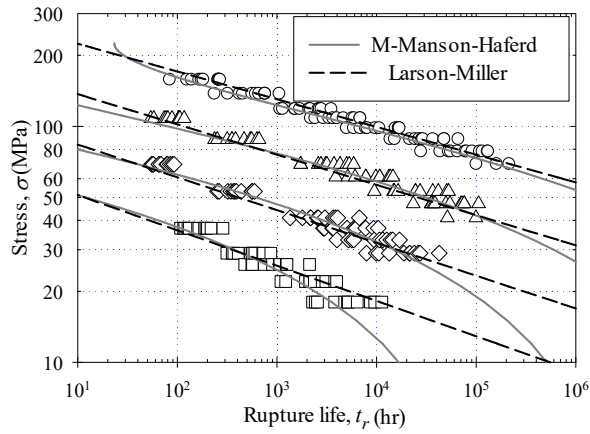
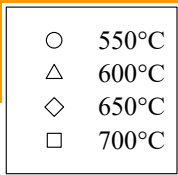
Stress-parameter function

Table – Effect of typical stress-parameter functions during extrapolation on Larson-Miller

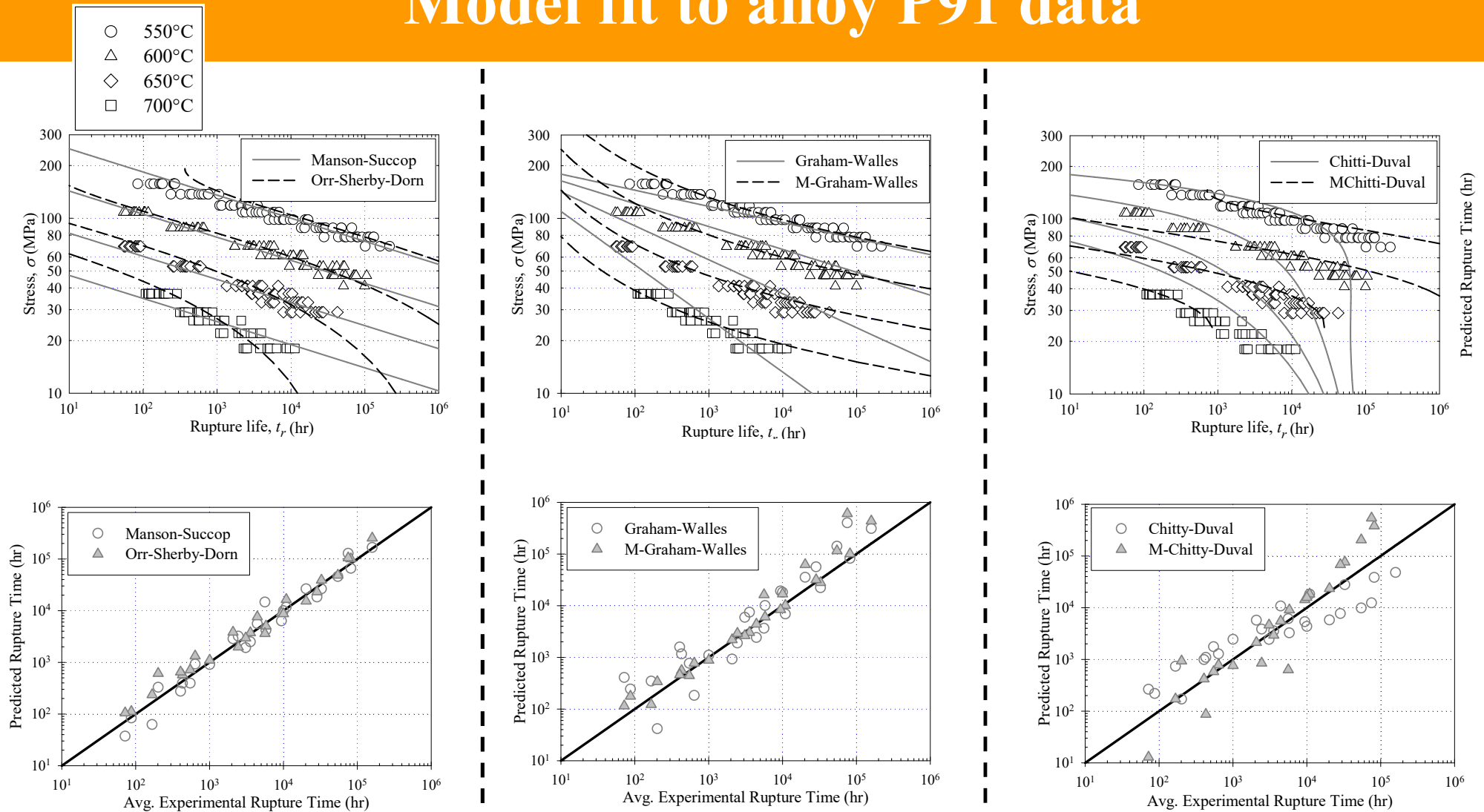
Function type	Equation	Prediction type	Inflection? $\frac{dP_i}{d\sigma} = 0$
Linear	$a + b\sigma$	High Weakening	NO
Logarithm	$a + b\sigma + c \log(\sigma)$	Stable	NO
Ploy-nominal	$a + b\sigma + c\sigma^2$	Weakening	YES
Power	$a + b\sigma^c$	Stable	NO
Exponential	$a + be^{c\sigma}$	Weakening	YES

- Force fit to a particular function may lead to poor prediction.
- A flexible option to choose function may improve prediction accuracy.

Model fit to alloy P91 data



Model fit to alloy P91 data



Calibrated Material constants and Stress-Parameter functions

Table – Material constants, stress-parameter functions, inflection status, and overall NMSE of the models

Model	Material constants	Master curve/Stress-parameter function $P_i = f(\sigma)$ with maximum R^2	Max. R^2	Inflection status	Overall NMSE
MMH	$\alpha_0 = 25.0$	$P_{MMH} = 2.9E - 7\sigma^2 - 1.2E - 4\sigma - 0.03$	0.97	YES	2.46
LM	$\alpha_0 = -19.0$	$P_{LM} = 22000 - 4680 \log(\sigma)$	0.98	NO	3.16
MH	$\alpha_0 = 16.6, \alpha_2 = 210$	$P_{MH} = -1E - 4\sigma - 0.025$	0.98	NO	6.05
GS	$\alpha_0 = 32.0, \alpha_2 = 248$	$P_{GS} = -.05\sigma^2 + 27.1\sigma + 10435$	0.98	YES	1.60
MB	$\alpha_0 = 14.0, \alpha_2 = 25.0$	$P_{MB} = 3E - 10\sigma^2 - 2E - 7\sigma - 2E - 5$	0.99	NO	1.49
MGS	$\alpha_0 = 520, \alpha_2 = 12.4$	$P_{MGS} = 0.31\sigma^2 - 12.8\sigma - 82076$	0.99	YES	1.70
MS	$\alpha_1 = -0.036$ $\alpha_2 = 5.00$	$P_{MS} = -3.29 \ln(\sigma) + 39.22$	0.98	NO	5.49
OSD	$\alpha_1 = 12100$ $\alpha_2 = 5.00$	$P_{OSD} = 2E - 4\sigma^2 - 7E - 2\sigma - 2E - 12.48$	0.99	YES	8.69
GW	$\alpha_2 = 1005$	$P_{GW} = 5E - 8 \ln(\sigma) - 3E - 7$	0.94	NO	66.8
MGW	$\alpha_2 = 900$	$P_{MGW} = 7E8\sigma^{-0.975}$	0.98	NO	64.6
CD	$\alpha_1 = f(\sigma)$	$P_{CD} = -3E - 4\sigma^2 + 0.2\sigma + 5.1;$ $\alpha_1 = -5E - 4\sigma^{0.9}$	0.99	YES	42.8
MCD	$\alpha_1 = f(\sigma)$	$P_{MCD} = -11.1 \ln(\sigma) + 19.2;$ $\alpha_1 = 1E4 \log(\sigma) + 3$	0.99	YES	81.6

- For most model, polynomial function produce the best goodness-of-fit.
- Polynomial function may exhibit inflection during extrapolation.
- Logarithmic and power function does not exhibit inflection during extrapolation.
- Inflection may be avoided by selecting next good-fit function at the cost of accuracy.

Normalized Mean Square Error

- Normalized Mean Square Error: $NMSE = \frac{1}{n} \sum_{i=1}^n [(X_{sim,i} - X_{exp,i}) / X_{exp,i}]^2$

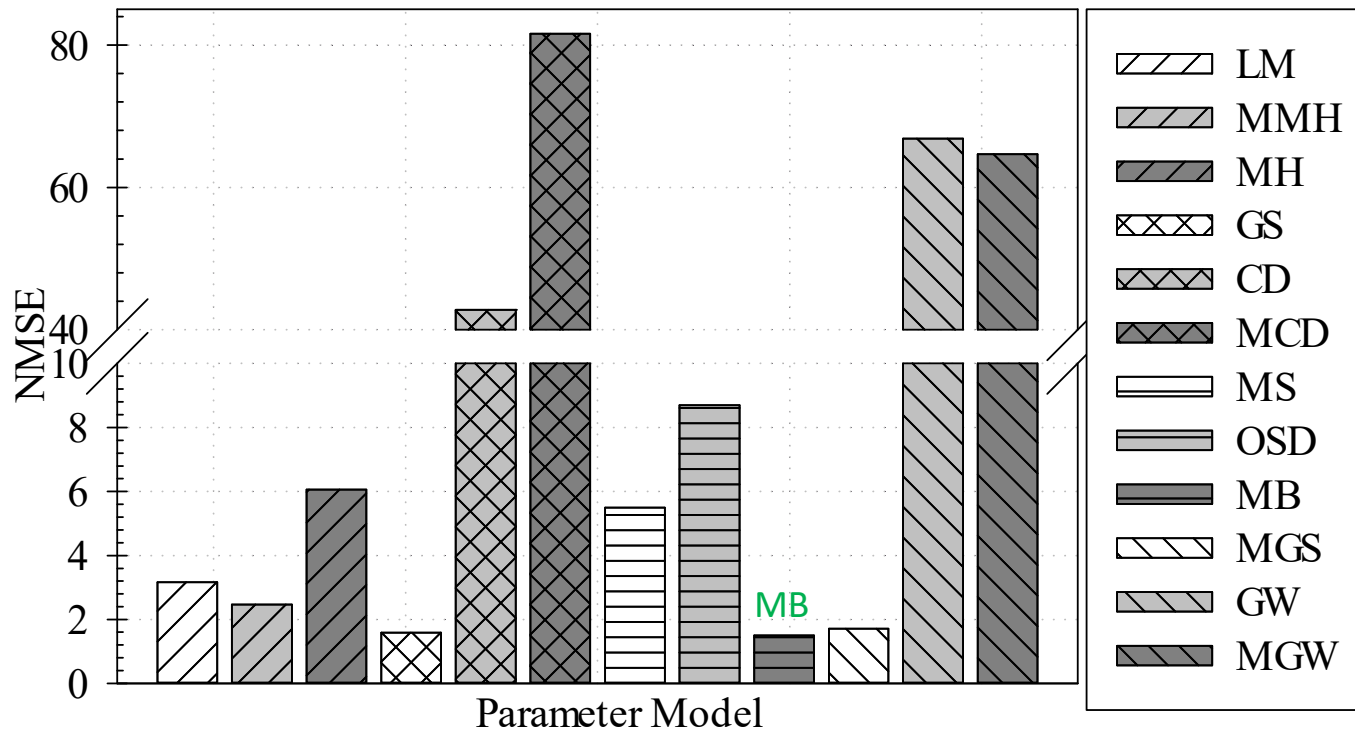
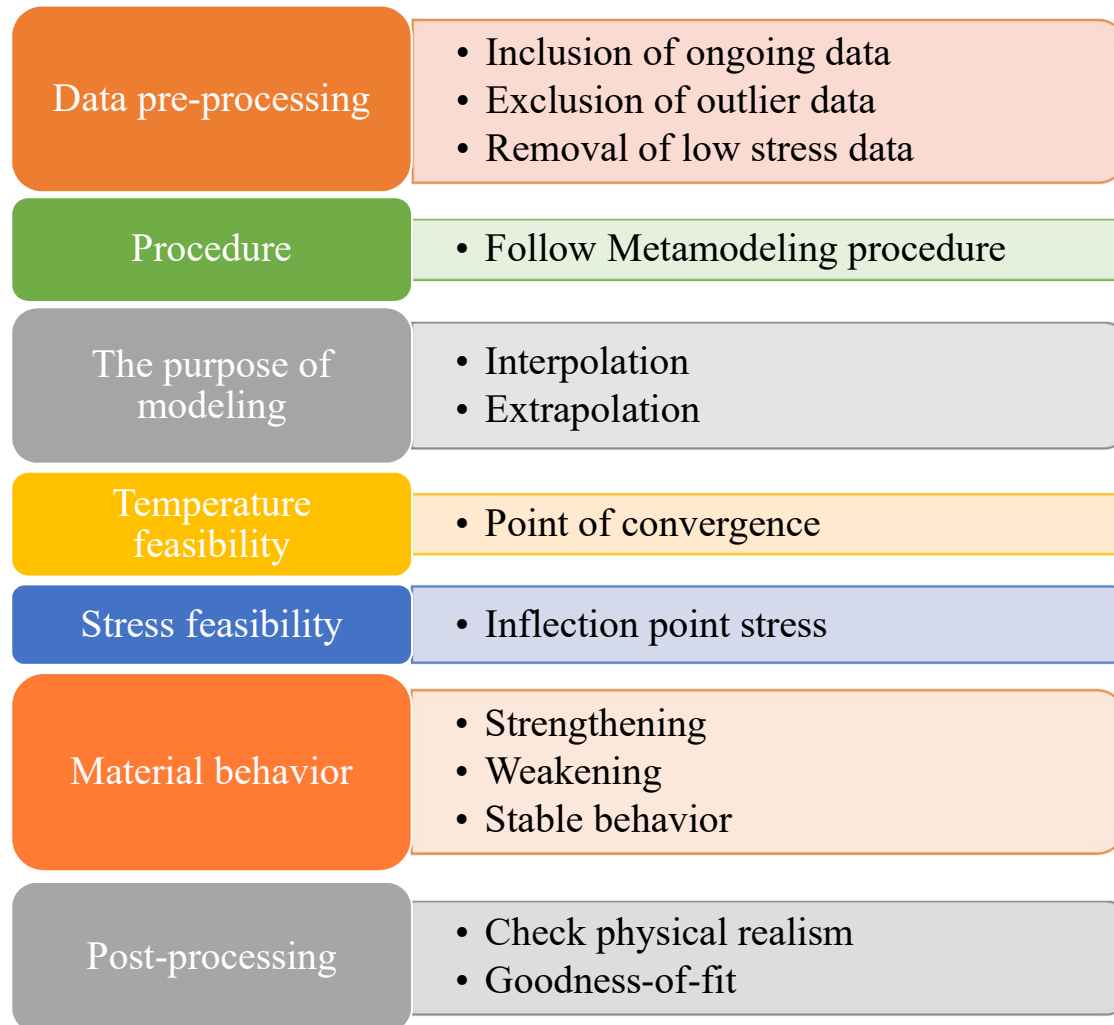


Figure - Comparison of the cumulative creep rupture NMSE of the models

- The Manson-Brown and Goldhoff Sherby produces the lowest NMSE.

Guideline to model selection

- ASME (American Society of Mechanical Engineers)
- ECCC (European Creep Collaborative Committee)
- ASTM (American Society for Testing and Materials)



Creep Deformation Metamodel

Table 6 – Combinational model summary

No	Base model	Creep Strain Rate	Damage Theta-Omega Identity, $I_{\theta\Omega} = (\dot{\varepsilon}_0 / \theta_3\theta_4)$	Life Prediction
M1	Theta	$\dot{\varepsilon} = (\dot{\varepsilon}_0 / I_{\theta\Omega}) \exp(\theta_4 t)$	$\omega(t) = \frac{t}{t_r}$	$t_r = \frac{1}{\theta_4} \ln \left(\frac{\varepsilon_r \theta_4 I_{\theta\Omega}}{\dot{\varepsilon}_0} + 1 \right)$
M2	Omega	$\dot{\varepsilon} = I_{\theta\Omega} \theta_3 \theta_4 \exp(\Omega \varepsilon)$	$\omega(t) = \frac{t}{t_r}$	$t_r = \left(\frac{1}{I_{\theta\Omega} \theta_3 \theta_4 \Omega} \right)$
Omega-Sinh Identity, $I_{\Omega S} = \dot{\varepsilon}_0 / A \sinh(\sigma / \sigma_s)$				
M3	Omega	$\dot{\varepsilon} = I_{\Omega S} A \sinh(\sigma / \sigma_s) \exp(\Omega \varepsilon)$	$\omega(t) = \frac{t}{t_r}$	$t_r = \left(\frac{1}{I_{\Omega S} A \sinh(\sigma / \sigma_s) \Omega} \right)$
M4	<u>Sinh</u>	$\dot{\varepsilon} = (\dot{\varepsilon}_0 / I_{\Omega S}) \exp(\lambda \omega^{3/2})$	$\omega(t) = -\frac{1}{\phi} \ln \left[1 - \left[1 - \exp(-\phi) \right] \frac{t}{t_r} \right]$	$t_r = \left(\frac{1}{\dot{\varepsilon}_0 \cdot \Omega} \right)$
Sinh-Theta Identity, $I_{S\theta} = \theta_3 \theta_4 / A \sinh(\sigma / \sigma_s)$				
M5	<u>Sinh</u>	$\dot{\varepsilon} = (\theta_3 \theta_4 / I_{S\theta}) \exp(\lambda \omega^{3/2})$	$\omega(t) = -\frac{1}{\phi} \ln \left[1 - \left[1 - \exp(-\phi) \right] \frac{t}{t_r} \right]$	$t_r = \frac{1}{\theta_4} \ln \left(\frac{\varepsilon_r}{\theta_3} + 1 \right)$
M6	Theta	$\dot{\varepsilon} = I_{S\theta} A \sinh(\sigma / \sigma_s) \exp(\theta_4 t)$	$\omega(t) = \frac{t}{t_r}$	$t_r = \frac{1}{\theta_4} \ln \left(\frac{\varepsilon_r \theta_4}{I_{S\theta} A \sinh(\sigma / \sigma_s)} + 1 \right)$

CDM Metamodel

<p>KR-LM metamodel</p> $I_{t_r, KR-LM} = \ln\left(\frac{M_K \sigma^z (\phi_K + 1)}{M_L \sigma^q}\right) - \ln\left(\frac{1 - \exp(-\phi_L)}{\phi_L}\right) - \ln(\phi_K + 1)$ $\omega_L = (-\phi_K / \phi_L) \ln(1 - \omega_K) + I_{t_r, KR-LM} / \phi_L$
<p>LM-Sinh metamodel</p> $I_{t_r, LM-Sinh} = \ln\left(\frac{M_L \sigma^P}{M_S \text{Sinh}(\sigma / \sigma_t)}\right) - \ln\left(\frac{1 - \exp(-\phi_S)}{\phi_S}\right) + \ln\left(\frac{1 - \exp(-\phi_L)}{\phi_L}\right)$ $\omega_S = (\phi_L / \phi_S) \omega_L + I_{t_r, LM-Sinh} / \phi_S$
<p>Sinh-KR metamodel</p> $I_{t_r, Sinh-KR} = \ln\left(\frac{M_K \sigma^z (\phi_K + 1)}{M_S \text{Sinh}(\sigma / \sigma_t)}\right) - \ln\left(\frac{1 - \exp(-\phi_S)}{\phi_S}\right) - \ln(\phi_K + 1)$ $\omega_S = (-\phi_K / \phi_S) \ln(1 - \omega_K) + I_{t_r, Sinh-KR} / \phi_S$

Creep Deformation Metamodeling process

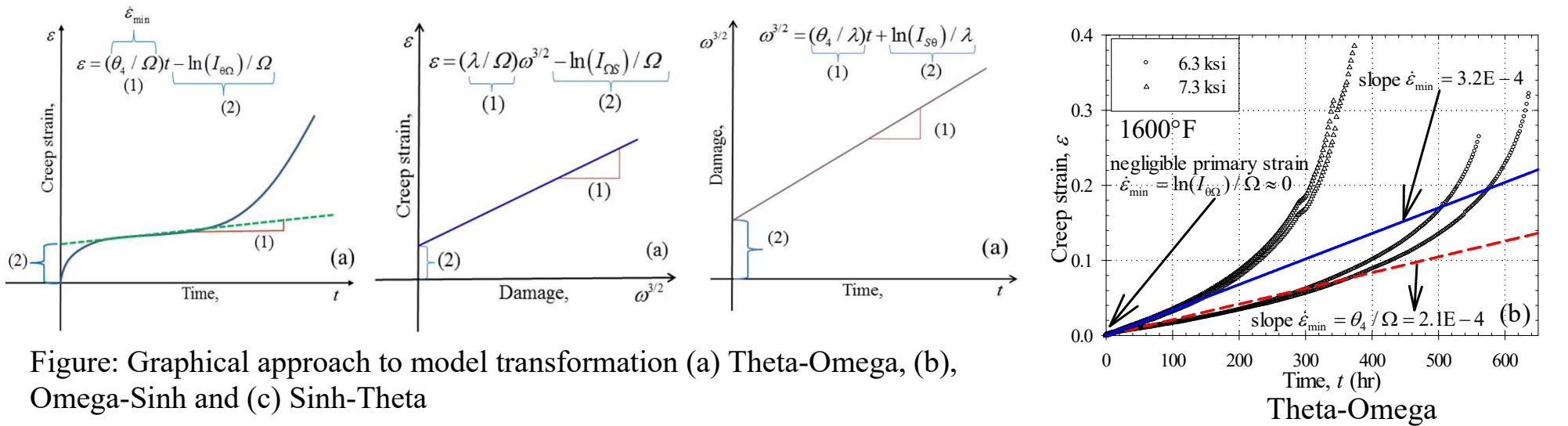
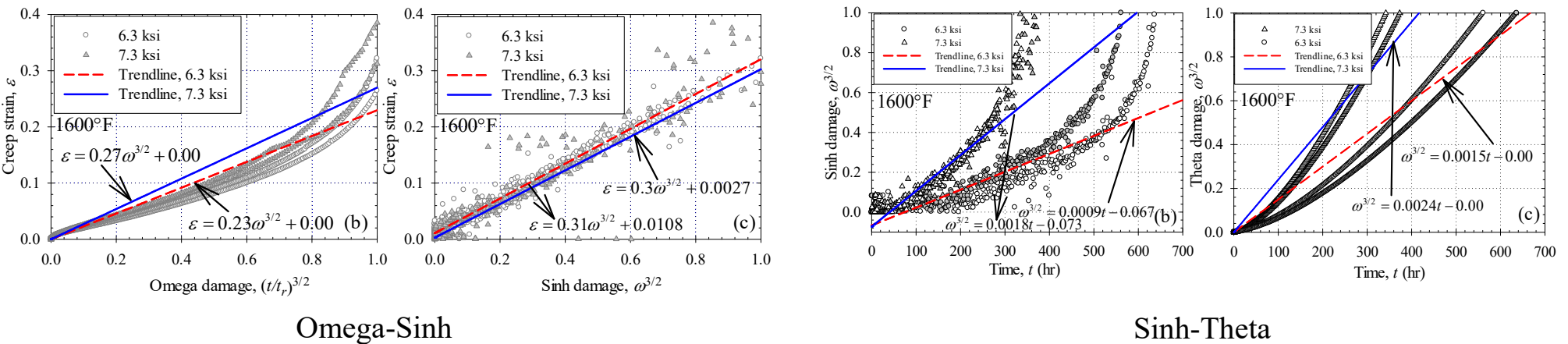


Figure: Graphical approach to model transformation (a) Theta-Omega, (b), Omega-Sinh and (c) Sinh-Theta



Omega-Sinh

Sinh-Theta

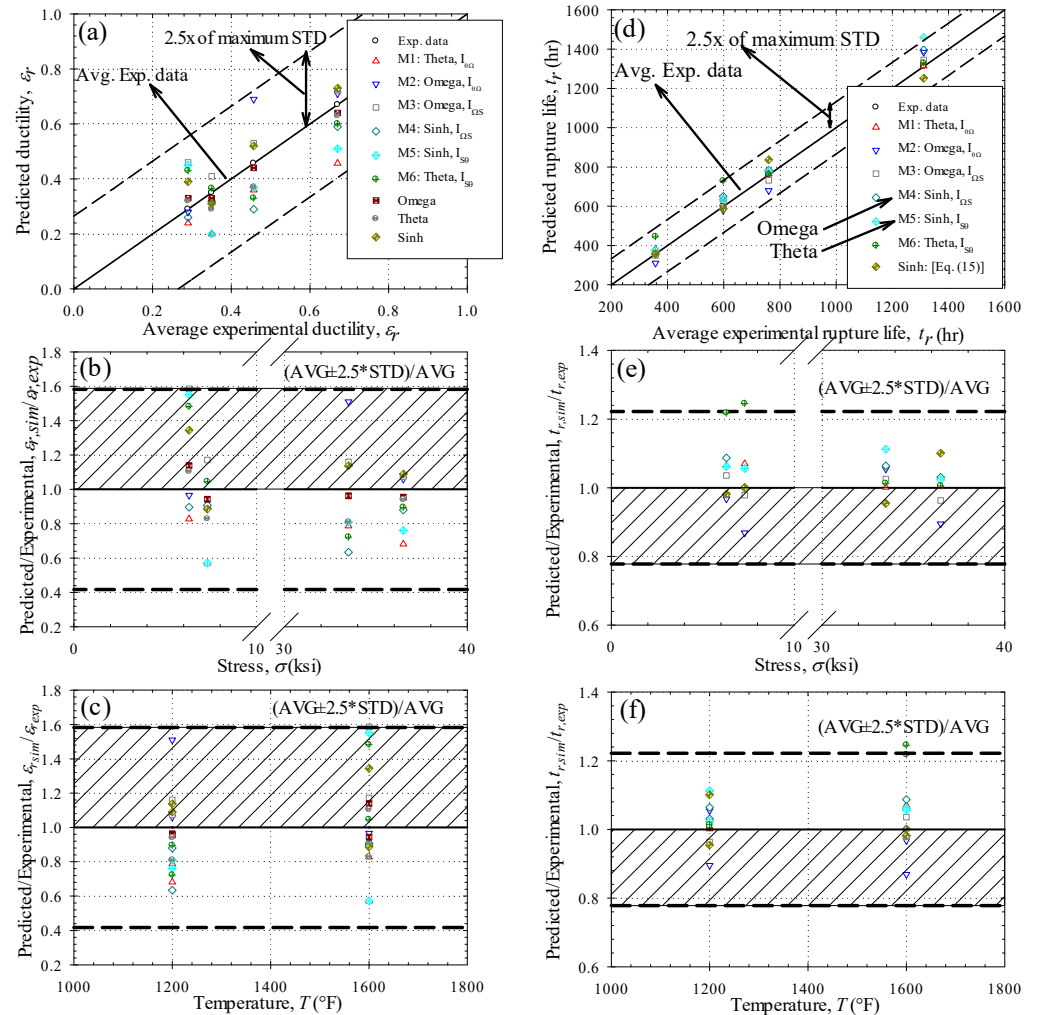
Assessment

- Normalized Mean Squared Error

$$NMSE = \frac{1}{n} \sum_{i=1}^n [(X_{sim,i} - X_{exp,i}) / X_{max}]^2$$

- Physical realism and numerical stability
- Conservatism

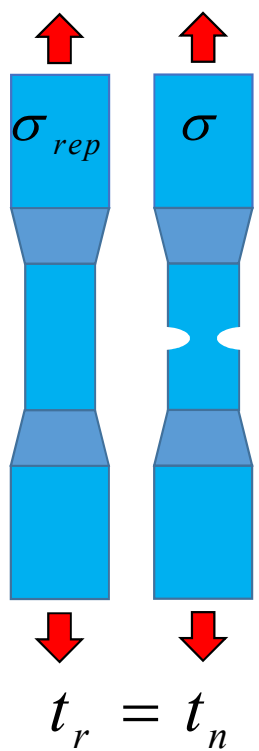
Figure: Model performance in predicting (a-c) creep ductility and (d-f) rupture life where the dotted lines indicate 2.5x of the maximum standard deviation in repeated tests. In the ratio plots, the dotted line indicates the ratio of 2.5x of the maximum standard deviation divided by the average obtained in repeated tests. The highlight areas indicate conservative predictions.



Task 8: Multiaxial Representative Stress Function

Representative stress (σ_{rep})

The stress applied to a plain bar that results in the same effective strain accumulation or rupture life as that obtained in a notched bar tested at the same temperature.



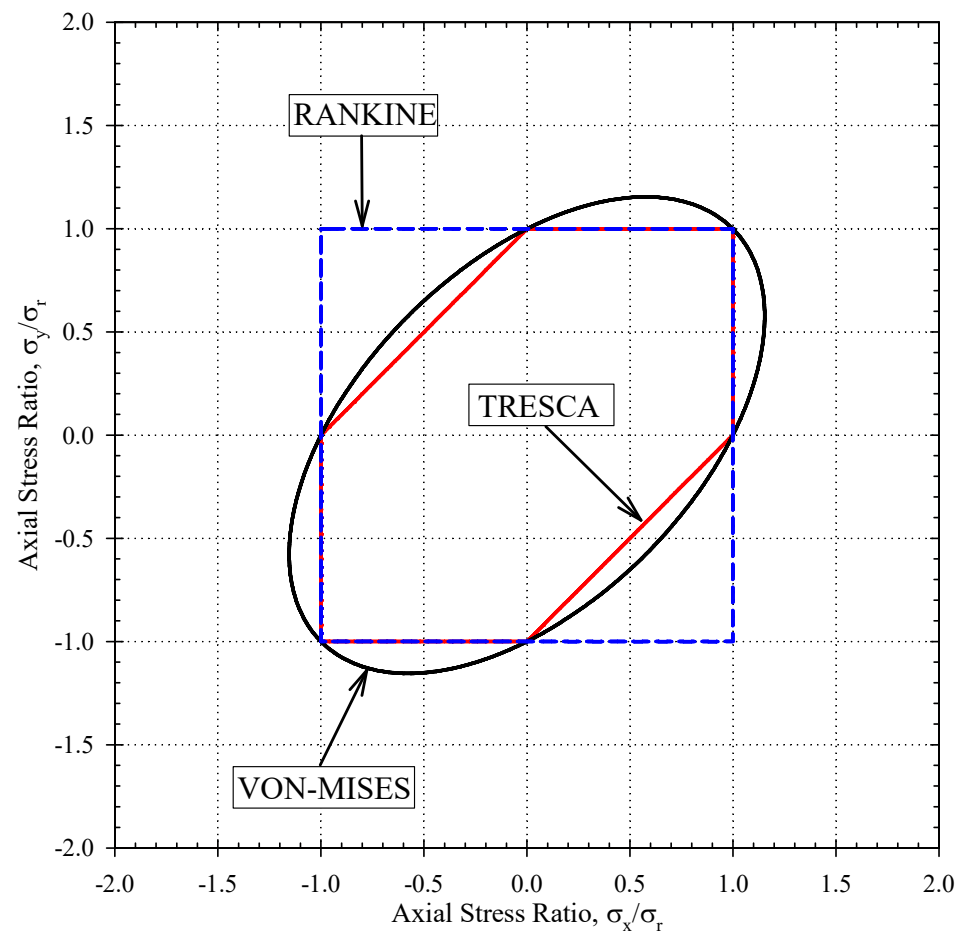
Classic approaches

- Rankine criteria, $\sigma_{p,max} > \sigma_{uts}$
- Tresca criteria, $\tau_{max} > \sigma_{uts}$
- Von-Mises criteria,

$$\sigma_{vm} = \frac{1}{\sqrt{2}} \sqrt{(\sigma_1 - \sigma_2)^2 + (\sigma_2 - \sigma_3)^2 + (\sigma_3 - \sigma_1)^2}$$

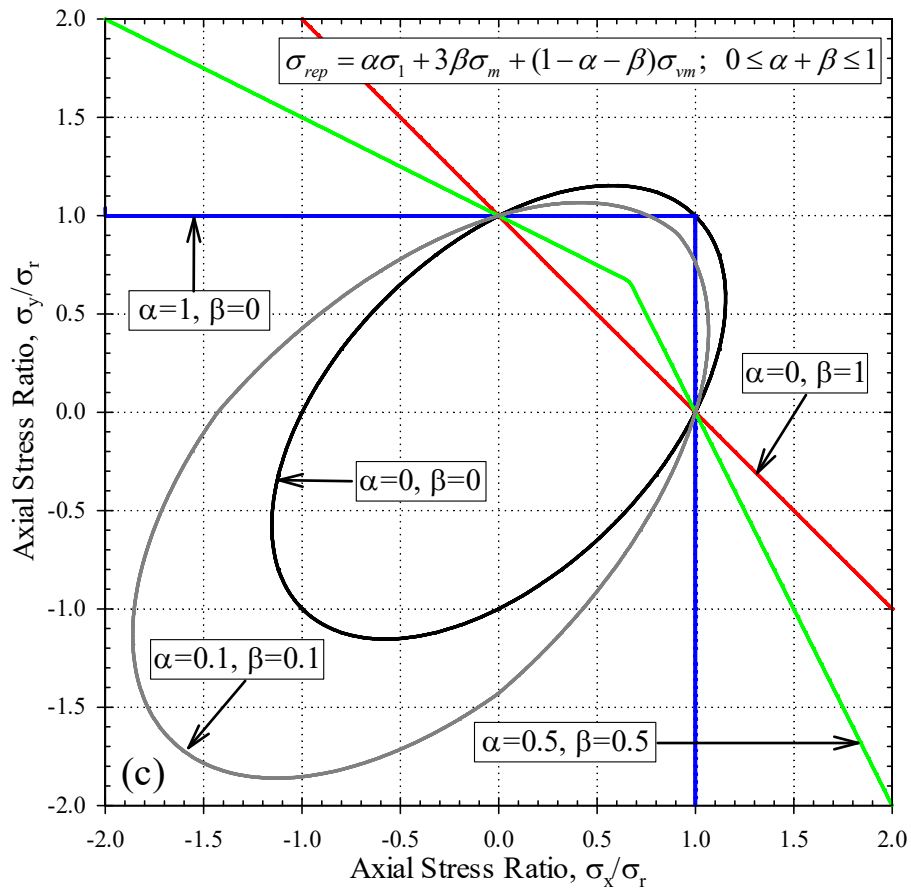
Limitations of Classic approaches

- Under equi-biaxial tension, $\sigma_x / \sigma_y = 1$ the rupture stress is non-conservative and identical to the uniaxial case.

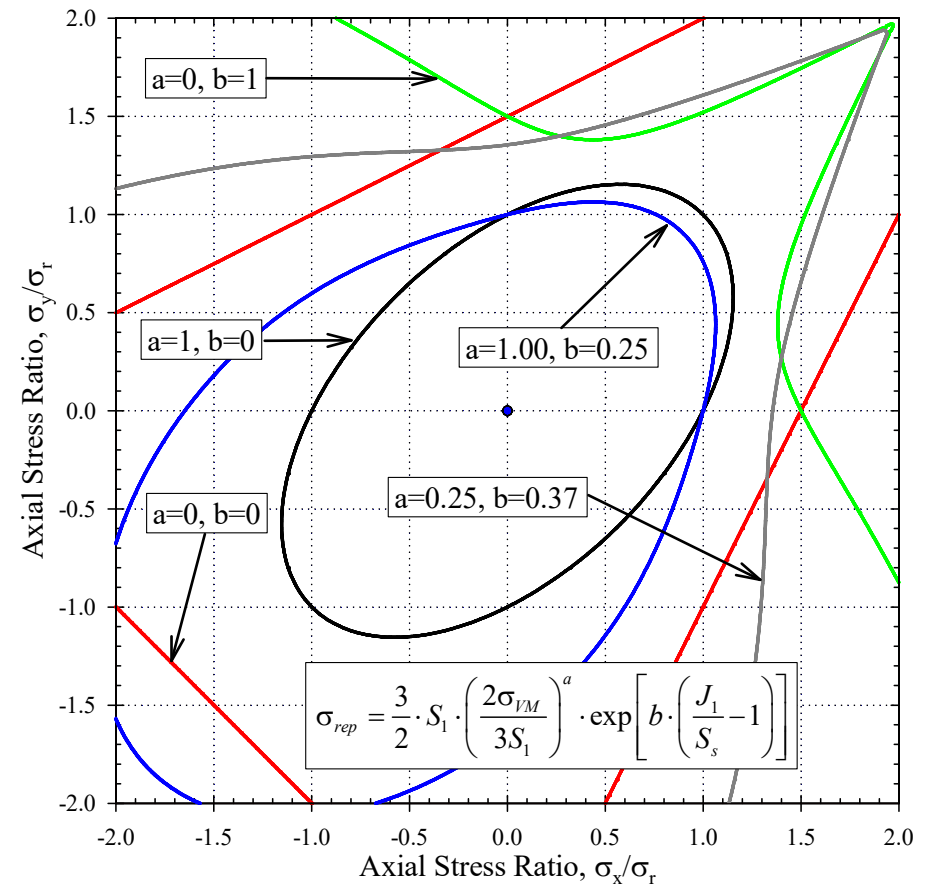


Classic function rupture surface

Representative Stress Functions



Hayhurst function rupture surface



Huddleston function rupture surface

Guideline to representative function selection

- ASME B&PV III (ASME Boiler and Pressure Vessel Code)

Stress function

- Either the Hayhurst or Huddleston function be used.
- Hayhurst function is simpler.
- Huddleston function is more flexible.

- RCC-MR (Design and Construction Rules for Power Generating Stations)

Numerical stability

- Check for infinite or imaginary condition leading to numerical instability.
- Hayhurst function is completely stable.
- Huddleston function is unstable at unstressed condition.

- ESIS (European Structural Integrity Society)

Material parameter calibration

- Uniaxial and Equi-biaxial tension creep tests.
- Regression analysis.

Damage mechanism

- Von-Mises stress: deformation.
- Hydrostatic stress: void growth.
- Principal stress: intergranular damage.

Incompressibility

- Constraints should be applied to enforce incompressibility.

Results and recommendations

- Performance of the individual models
- Performance of the Metamodels
- Recommendations
 - Guideline to TTP model selection
 - Guideline to representative function selection.
 - Guideline to minimum creep strain rate model selection.
 - Guideline to CDM model selection.
- A master guideline to adaptive approach towards creep modeling.

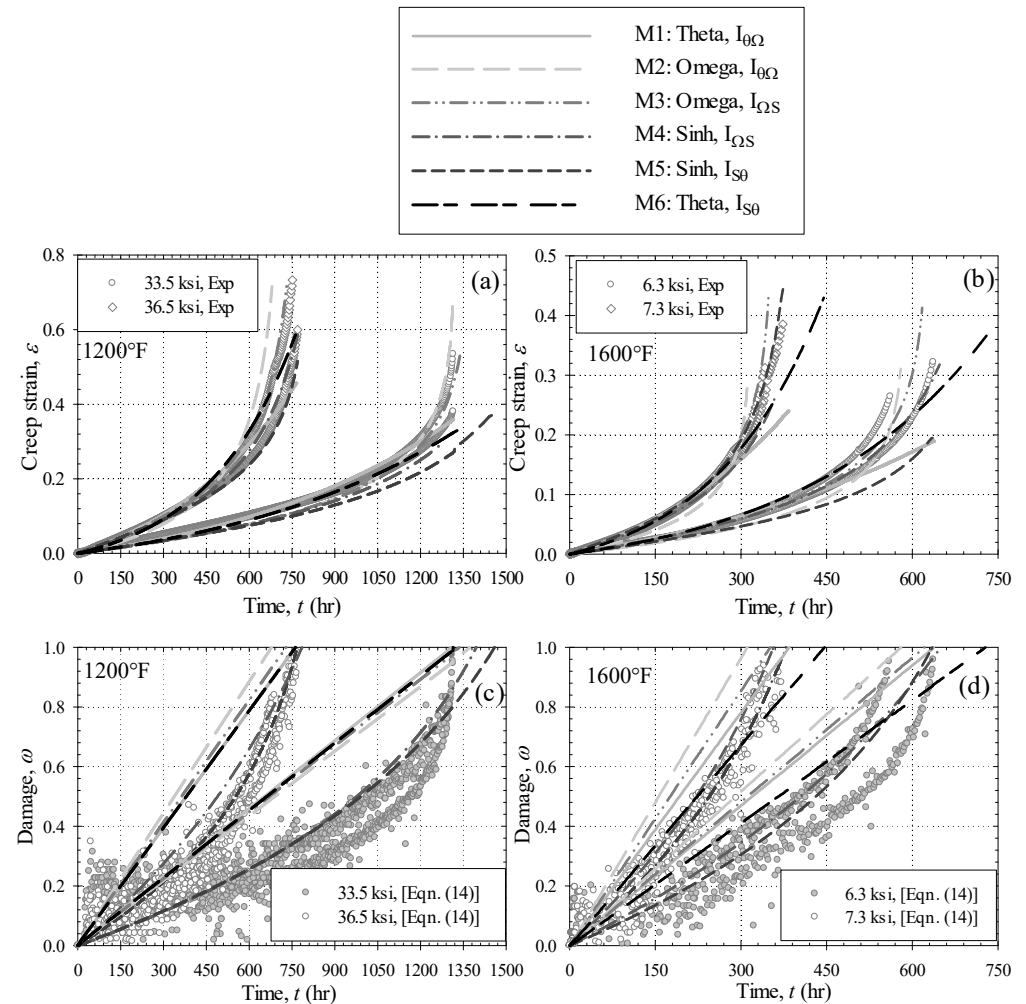
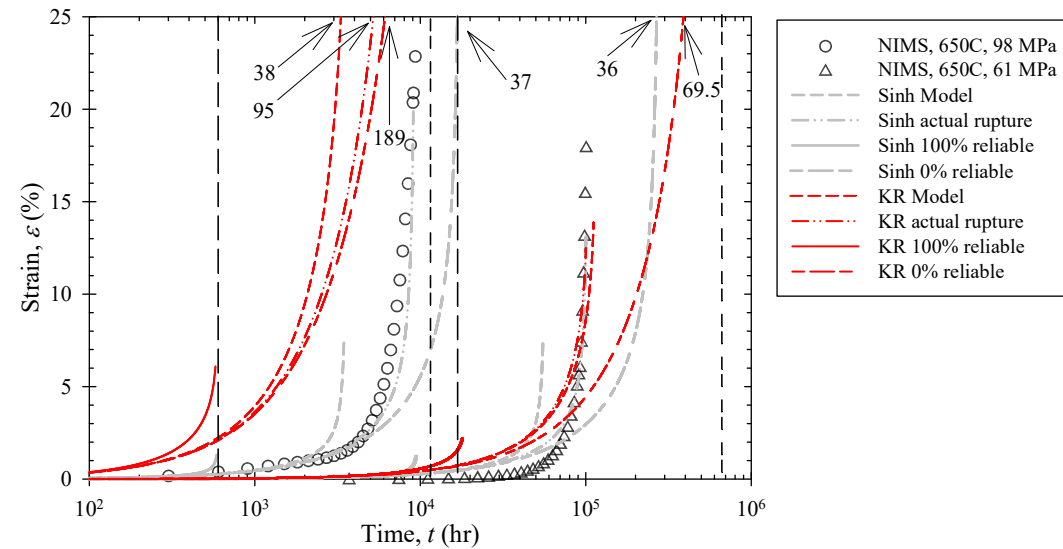


Figure: Comparison of combinational models using Siemens data (a-b) deformation, (c-d) damage prediction for Hastelloy X at 1200°F and 1600°F

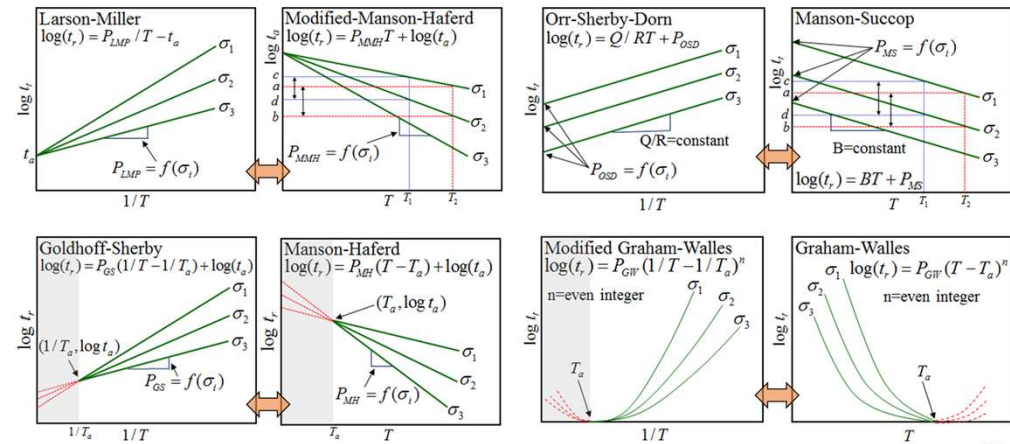


Summary



- Inclusion of **disparate data** into calibration process can significantly improve the prediction accuracy.

- A **metamodel** derived by combining and exploiting creep models where the submodels (existing and new models that are exploited to develop the metamodel) are special case can facilitate instantaneous and efficient evaluation of the submodels leading to the selection of “best” model”.



Publications

1. Haque, M. S., and Stewart, C. M. (2017). The Stress-Sensitivity, Mesh-Dependence, and Convergence of Continuum Damage Mechanics Models for Creep. **Journal of Pressure Vessel Technology**, 139(4), 041403.
2. Haque, M. S., Ramirez, C. and Stewart, C. M. (2017). Novel Metamodeling Approach For Time-temperature Parameter Models. **ASME** Pressure Vessel and Piping Conference. Paper No. PVP2017-65297. (**Award: Honorable mention**)
3. Haque, M. S., and Stewart, C. M. (2017). Selection Of Representative Stress Function Under Multiaxial Stress State Condition For Creep. **ASME** Pressure Vessel and Piping Conference. Paper No. PVP2017-65296.
4. Ramirez, C., Haque, M. S., and Stewart, C. M. (2017). Guidelines To The Assessment Of Creep Rupture Reliability For 316SS Using The Larson-miller Time-temperature Parameter Model. **ASME** Pressure Vessel and Piping Conference. Paper No. PVP2017-65816.
5. Haque, M. S., Ramirez, C., Chessa, J. F., Stewart, C. M., "A Guideline for the Assessment of Uniaxial Creep and Creep-Fatigue Data and Models", 2017 Project Review Meeting for Crosscutting Research, Gasification Systems and Rare Earth Elements Research Portfolios, NETL, Department of Energy.
6. Haynes, A. C., Haque, M. S., Stewart, C. M., "The Numerical Analysis of Equivalent Stress Functions for Multiaxial Creep Deformation, Damage, and Rupture", UTEP COURI Project, May 2017.

In progress:

1. Haque, M. S., and Stewart, C. M. (2018). The Disparate Data Problem: The Calibration of Creep Laws Across Test Type And Stress, Temperature, and Time Scales. (ready for submission)
2. Haque, M. S., and Stewart, C. M. (2018). Metamodeling Time-Temperature Parameter Models for Creep Rupture. (90% complete)
3. Perez, J., Vega, R., Haque, M., Zamorano, D., and Stewart, C. (2018). Calibration and Validation of Phenomenological Creep Deformation Laws. (90% complete)

Future Work

Software: Secondary Creep model

Material Constant Custom Input

Source file:	Select Model	Constant Input
<input type="button" value="File"/>	<input type="checkbox"/> Norton	<input type="checkbox"/> A <input type="text"/>
<input type="button" value="Apply"/>	<input type="checkbox"/> Soderberg	<input type="checkbox"/> n <input type="text"/>
	<input type="checkbox"/> Dorn	<input type="checkbox"/> σ_0 <input type="text"/>
	<input type="checkbox"/> McVetty	<input type="checkbox"/> σ_1 <input type="text"/>
	<input type="checkbox"/> Garofalo	<input type="checkbox"/> σ_2 <input type="text"/>
	<input type="checkbox"/> JHK	



Ricardo Vega

Future Work

Software: Creep Deformation model

Material Constant Custom Input

Source file:	Select Model	Constant Input
<input type="button" value="File"/>	<input type="checkbox"/> Omega	σ_0 <input type="text"/>
<input type="button" value="Apply"/>	<input type="checkbox"/> Theta projection	σ_1 <input type="text"/>
	<input type="checkbox"/> Sine-Hyperbolic	σ_2 <input type="text"/>
	<input type="checkbox"/> Kachanov-Rabotnov	σ_3 <input type="text"/>
		σ_4 <input type="text"/>
	<input type="button" value="Use Data base"/>	<input type="button" value="Execute"/>



Jimmy Perez

Future Work

Software: Extrapolation and Design Maps

Help!!

Source File

Temperature Range

Stress Range

Parameters

Select model

- Stress-rupture
- Secondary creep
- Deformation

Apply

Material Constant Custom Input

Temperature Function

- Linear
- Logarithmic
- Exponential
- Power
- Polynomial

Check Inflection

Stress Function

- Linear
- Logarithmic
- Exponential
- Power
- Polynomial

Check Inflection

Output

- Extrapolation
- Design map
- Detail Analysis

Execute

Thank you

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National Energy Technology Laboratory (NETL)
Award Number(s): [DE-FE0027581](#)



- Omer R. Bakshi
– Federal Project Manager, Crosscutting Research, NETL, U.S. DOE



UTEP MERG Research Team

Time-Temperature Parameter (TTP) Model

① **TTP model**

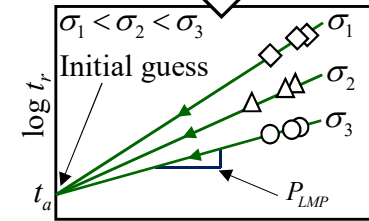
② Find initial guess constant, t_a
(graphical approach)

③ Calculate P_{LMP} for each data point

④ Simultaneous calibration of constant, t_a and stress-parameter function.

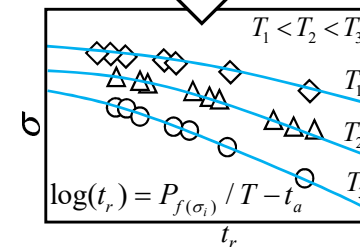
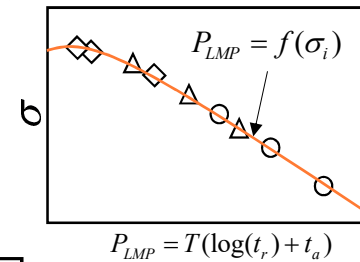
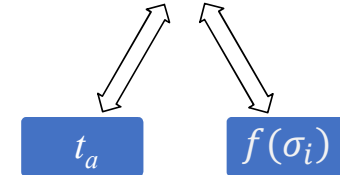
⑤ Plot model with experimental data

$$\log(t_r) = P_{LMP} / T - t_a$$



$$P_{LMP} = T(\log(t_r) + t_a)$$

calibration to P_{LMP}



Metamodel

Existing approach

- Holdsworth, 1999 (LM, MH, Mod-MH, MRM, and Mod-MRM)

$$\log(t_r) = \left\{ \sum_{k=0}^n \beta_k (\log[\sigma]^k) \right\} (T - T_a) + \beta_5; \quad n = 2, 3, 4$$

- Eno, 2008 (LM, MH, OSD, MS, and MRM)

$$\log(t_r) = \beta_0 + \beta_1 (T) + \beta_2 \log(\sigma) + \beta_3 \log(\sigma) (T)$$

$$\log(t_r) = \beta_0 + \beta_1 \left(\frac{1}{T} \right) + \beta_2 \log(\sigma) + \beta_3 \log(\sigma) \left(\frac{1}{T} \right)$$

- Seruga, 2011 (LM, MH, OSD, and MB)

$$\log(t_r) = (T - T_a \cdot \langle q \rangle)^q (a_0 + a_1 \log(\sigma) + a_2 \log^2(\sigma)) + \log(t_a) T^{q-1}$$



Suitable for verified material behavior.

- New approach:** (Includes 12 TTP models)

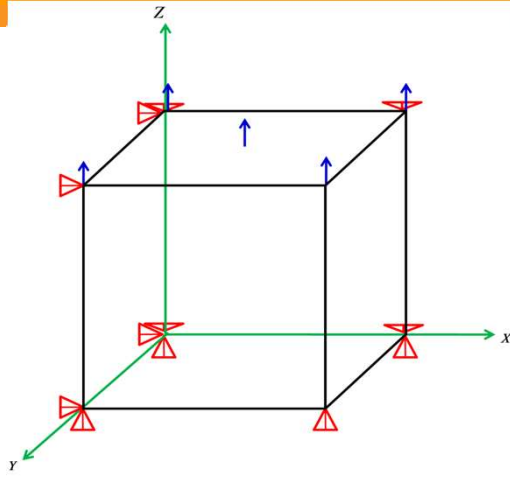
$$P_{HS}(\sigma) = \frac{\log(t_r) - \alpha_0 - \alpha_1 T^r}{(T^r - \alpha_2^r)^q}$$



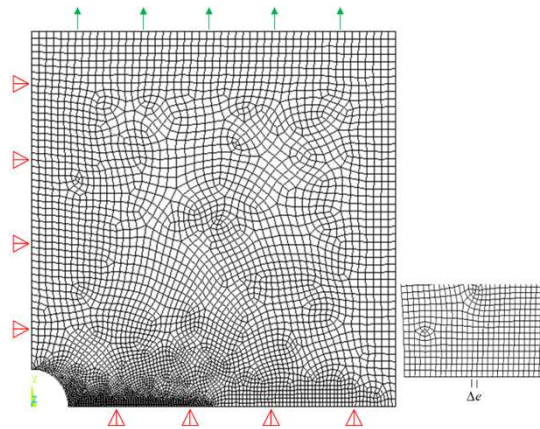
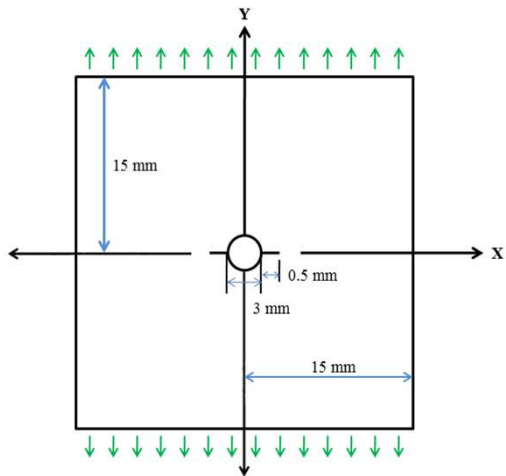
Flexible to fit any material behavior.

Task 8: Computational Analysis (2D)

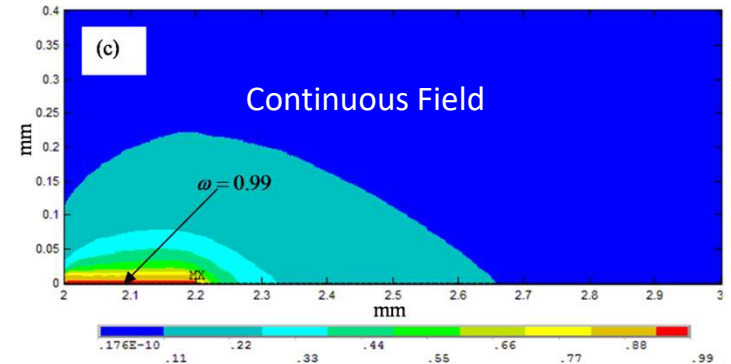
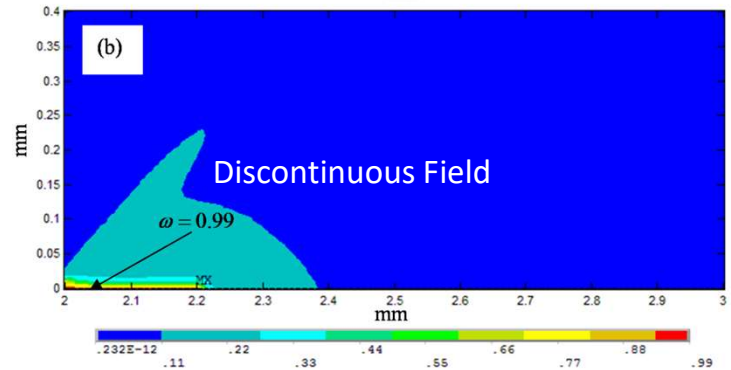
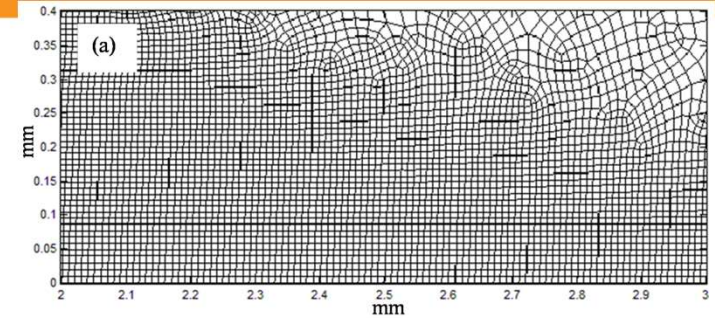
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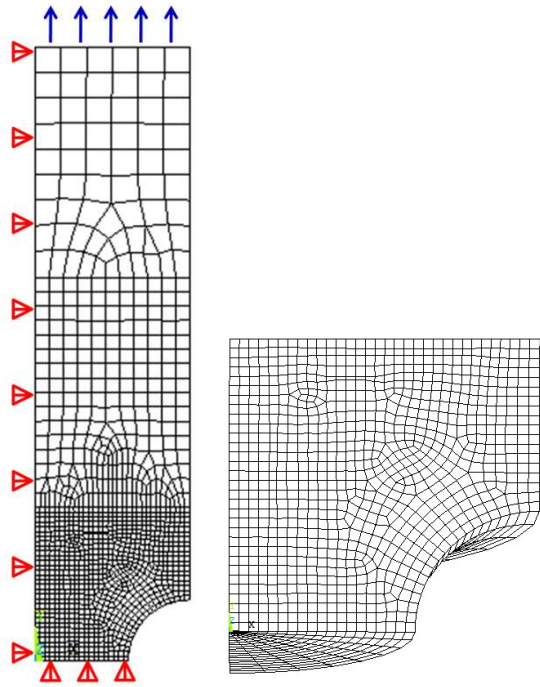
Eight node single element with boundary constraint



2D center-hole plate (a) dimensions, (b) ANSYS mesh ($\Delta e = 0.05$ mm)



Task 8: Computational Analysis (3D)



ANSYS FEM mesh of Bridgeman notch specimen

Sinh damage at 250 MPa and 700°C (a) 50 hr, (b) 200 hr, (c) 400 hr, (d) 600 hr, (e) 700 hr, and (f) 832 hr

