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- 2. Interdisciplinary tools



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- 3. Effective optimizer



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## Demands



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1. Low computational cost



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## Demands

- 1. Low computational cost
- 2. Easy model-switching

# A computational framework for designing Ni-based superalloy

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- **\*** Background:  $\gamma \gamma'$  Ni Base Superalloy
- Computational Framework
  - Classical Nucleation, Growth and Coarsening Models
  - Elastic Deformation: Statistical tool, PyMKS
  - Plastic Deformation: Kocks-Mecking & energy conservation models
  - Optimization



 $\gamma - \gamma'$  Ni-Base Superalloy

required

Olson, 1997 Ahmadi et al., 2014 Le Baillif et al., 2014

b) Tensile tests at 850°C •  $\gamma$  and  $\gamma'$  microstructure b С a 700 600 •  $\gamma'$  strengthening Stress (MPa) 500 Ψ≈0<sup>6</sup> • Chemistry & geometry 400 -300°C/min т properties affect performance 300 =120°C/min -30°C/min 200 0.0% 0.2% 0,4% 0,6% 0,8% • Optimum composition &  $120^\circ \le \Psi_c \le 180^\circ$  $0^\circ < \Psi_c \le 120^\circ$  $\Psi_c = 0^\circ$ strain (mm/mm) Weak & shearable Strong & shearable Non-shearable precipitate precipitate precipitate **High Temperature Application** processing conditions are Precipitation hardening Performance Properties Structure Processing Microstructure 11



















#### Homogeneous, NON-Steady-State Nucleation

$$\dot{N}(t) = Z\beta N_0 \exp\left(\frac{-\Delta G^*}{K_B T}\right) \exp\left(\frac{-\tau}{t}\right)$$

- Cluster size is estimated by Boltzmann distribution of total number of clusters, N<sub>0</sub>
- Interface energy  $(E_{INT})$  and  $N_0$  are the remaining parameters
- $\beta$  rate of atom jumps from matrix to precipitate
- Zeldovich factor, Z, is employed to correct the equation
- The parameters are calculated using Thermo-Calc with TCNI6 database

$$\Delta G^* = \frac{16\pi}{3} \frac{E_{INT}^3}{\Delta G_N^2}$$

$$\Delta G_N = \frac{\Delta G_{ch}}{V_m^{\gamma'}} = -\frac{RT}{V_m^{\gamma'}} \sum_i \bar{C}_i^{\gamma'} \ln \frac{a_i}{\bar{a}_i^{\gamma'}}$$

$$Z = \frac{V_M^\beta}{2\pi N_A r_0^2} \sqrt{\frac{E_{INT}}{K_B T}}$$

$$\beta = \frac{4\pi r_0^2}{a^4} \left[ \sum_i \frac{\left(\bar{C}_i^{\gamma'} - \bar{C}_i^{\gamma}\right)^2}{\bar{C}_i^{\alpha} D_i} \right]^{-1}$$

$$r_0 = -\frac{2E_{INT} V_M^\beta}{\Delta G_N}$$

$$\tau = \frac{1}{\theta Z^2 \beta}$$

Rougier et al., 2013 Perez et al., 2008

 $\gamma'$  growth

$$\frac{dR}{dt} = \frac{D_i^{\gamma}}{\left[1 - \lambda_j \sqrt{\pi} \exp(\lambda_j^2) \operatorname{erfc}(\lambda_j)\right] R} \frac{C_i^{\gamma} - \bar{C}_i^{\gamma}}{\left(\bar{C}_i^{\gamma'} - \bar{C}_i^{\gamma}\right)}$$

- $\bar{C}_i^{\gamma'}$  is taken as the equilibrium composition in  $\gamma'$
- $\left[1 \lambda_j \sqrt{\pi} \exp(\lambda_j^2) \operatorname{erfc}(\lambda_j)\right] R$  stands the Effective Diffusion Distance

 $\gamma'$  coarsening

$$\frac{dR}{dt} = \frac{8}{27} \frac{E_{INT} V_M^{\gamma'}}{R^2 N_A k_B T} \frac{D_i^{\gamma} \bar{C}_i^{\gamma}}{\bar{C}_i^{\gamma'} - \bar{C}_i^{\gamma}}$$

	Experimental		Data Regression		
	Composition, %	<i>Т</i> <sub>р</sub> , К	reference	<b>E</b> <sub>int</sub> , mJ/m <sup>2</sup>	<b>N</b> <sub>0</sub> , 1/m <sup>2</sup>
Kt1	Ni-7.5Al-8.5Cr	873	Booth-Morrison et al., 2008	15	1.5x10 <sup>26</sup>
Kt2	Ni-9.8Al-8.3Cr	1073	Sudbrack et al., 2008	24	5x10 <sup>27</sup>
Kt3	Ni-6.5Al-9.5Cr	873	Booth-Morrison et al., 2010	18	4.0x10 <sup>26</sup>



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Nishizawa et al., 2001 Li et al., 2002

Nishizawa et al., 2001

$$E_{int} = \alpha | \Delta H^{\gamma \to \gamma'}$$

$$E_{int} \& |\Delta H^{\gamma \to \gamma'}|$$

$$\mathbf{a} = f(T_p)$$

$$\mathbf{a} = (3.75 \times 10^{-2}T_p - 2.23) \times 10^{-6}$$

	Experimental		Computation	Parameter		
	Composition %	T <sub>p</sub> Kelvin	<i>ΔΗ<sup>γ→γ′</sup></i>   10⁴ J/mol	<mark>E<sub>int</sub></mark> mJ/m²	α 10 <sup>-6</sup> mol/m²	<b>E<sub>int,cal</sub></b> mJ/m <sup>2</sup>
Kt1	Ni-7.5Al-8.5Cr	873	1.52	15	0.99	15.
Kt2	Ni-9.8Al-8.3Cr	1073	1.34	24	1.80	24
Kt3	Ni-6.5Al-9.5Cr	873	1.64	18	1.10	17.1

$$E_{int} = (3.75 \times 10^{-2} T_p - 2.23) \times 10^{-6} |\Delta H^{\gamma \to \gamma'}|$$



## **Yield Stress**

Thomas et al., 2006 Roth et al., 1997 Reed., 2006 Collins et al., 2014

$$\sigma_{YS} = \sigma_0 + \sigma_{SS} + \sigma_{H-P} + \sqrt{\sigma_{\rho,0}^2 + \sigma_p^2}$$

- Calculations using results from previous steps:  $Vf_{\gamma'}, r_{\gamma'}$ , and composition of matrix
- Lattice friction ( $\sigma_0$ ), solid solution ( $\sigma_{SS}$ ) and Hall-Petch effect ( $\sigma_{H-P}$ ) to yield stress is estimated [Roth97].
- $\sigma_P$  follows the minimum value among  $\sigma_{wc} \sigma_{sc}$  and  $\sigma_{or}$ .
- *E*<sub>APB</sub> is calculated

$$\sigma_{wc} = M \frac{E_{APB}}{2b} \left[ \left( \frac{6\bar{R}^{\gamma'} E_{APB} V_f^{\gamma'}}{\pi L_T} \right)^{0.5} - V_f^{\gamma'} \right]$$
$$\sigma_{sc} = 0.22M \left( \frac{\mu b}{\bar{R}^{\gamma'}} \right) \left( \frac{\pi \bar{R}^{\gamma'} E_{APB} V_f^{\gamma'}}{L_T} - V_f^{\gamma'} \right)^{0.5}$$
$$\sigma_{or} = M \frac{\mu b}{\bar{R}^{\gamma'}} \sqrt{\frac{V_f^{\gamma'}}{\pi}}$$

## APB Energy



 $W_{13} = \frac{3\Delta H^{FCC} + \Delta H^{ORD} \frac{\left(1 - \bar{C}_{s}^{\gamma'}\right)}{\bar{C}_{s}^{\gamma'}}}{24Rc\left(1 - \bar{C}_{s}^{\gamma'}\right)}$ 

 $= W_1 + 2W_3$ 

• Chemical property of <u>Ni-Al-Cr</u> gives

 $E_{APB,[111]} = 0.06 - 0.18 \text{ J/m}^2$ 

Chemical properties are obtained at  $T_s$ 

equilibrium condition at  $T_p$ 

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<i>C<sub>Al</sub></i> , %	13.62
C <sub>cr</sub> , %	17.46
$T_p$ , Kelvin	1230
$V_{f,max}^{\gamma'}$ , %	43.89
$E_{int}$ , mJ/m <sup>2</sup>	25.51
$E_{APB}$ , mJ/m <sup>2</sup>	118



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Average radius

$$\bar{R}^{\gamma'} = \frac{\sum N_c R_c^{\gamma'}}{\sum N_c}$$



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Average radius

$$\bar{R}^{\gamma'} = \frac{\sum N_c R_c^{\gamma'}}{\sum N_c}$$



<i>C<sub>Al</sub></i> , %	14.05
C <sub>cr</sub> , %	10.16
$T_p$ , Kelvin	1328
$V_{f,max}^{\gamma'}$ , %	26.31
$E_{int}$ , mJ/m <sup>2</sup>	27.47
$E_{APB}$ , mJ/m <sup>2</sup>	157





Statistical/machine learning tool correlating microstructure with properties by linear function

- By employing machine learning technique, • PyMKS can reproduce the FEM calculations in a more efficient way
- For the microstructure with high volume ٠ fraction of  $\gamma'$ , the representative volume element is created for PyMKS calculation





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#### Machine Learning Tech for Elastic Deformation – PyMKS

Fast et al., 2011 Kalidindi, 2012

The evolution of the <u>dislocation density</u> ( $\rho$ ) is calculated as increasing strain based on non-equilibrium thermodynamics and Kocks-Mecking model using <u>shear modulus</u> ( $\mu$ ), <u>burger's vector</u> (b), <u>mean free path</u> of dislocation (l), <u>vibration frequency</u> ( $v_0$ ), <u>energy barrier</u> of dislocation annihilation ( $\Delta G_{\rho}$ ) and <u>model constant</u> (C,  $\alpha$ )

$$\rho_{in,\varepsilon + \Delta\varepsilon} = \rho_{in,\varepsilon} + \frac{(\mu b^2 + \tau bl)\frac{\nu_0}{\dot{\varepsilon}}exp\left(-\frac{\Delta G_{\rho}}{kT}\right)\rho_{in,\varepsilon} - \tau_{\rho}}{\frac{1}{2}C\alpha\mu b^2 - (\mu b^2 + \tau bl)}\Delta\varepsilon$$

$$\Delta\varepsilon$$

$$\Delta\varepsilon$$

$$M\tau = \sigma_0 + \sigma_{SS} + \sigma_{H-P} + \sqrt{\sigma_{\rho}^2 + \sigma_{p}^2}$$

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#### **Plastic Deformation Model**

	$V_f^{\gamma'}$	$\sigma_{ys}$ , MPa	ε <sub>ys</sub> , %
PD1	0.323	450	0.29
PD2	0.296	560	0.33
PD3	0.278	570	0.36



• 
$$T_s = 1123K$$

• PD1 is used for calibration

• 
$$\Delta G_{\rho} = 3.08 \text{ eV}$$
 and  
 $C = -180$  provide the  
best agreement to PD1

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- $T_s = 1123K$
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#### Input conditions

	<i>C<sub>Al</sub></i> ,%	C <sub>Cr</sub> ,%	$T_p$ , К
Min	10	10	1123
Max	25	20	1473

- Objective: high work to necking  $(E_{WTN})$  at  $T_s = 1123K$
- Pass the calculations while  $\bar{V}_{f}^{\gamma'} < 0.4$
- 6 bits of memory is utilized for 1 variable
- 6 samples are selected in 1 generation and 10 generations are calculated

#### Optimization



#### Optimization



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#### **Optimization - WTN**



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<i>C<sub>Al</sub></i> , %	0.231
<i>C<sub>Cr</sub></i> , %	0.195
$T_p$ , Kelvin	1450
$\sigma_{UTS}$ , MPa	394
$\varepsilon_{UTS}$ , %	12.98



- The optimization of Ni-Al-Cr ternary system is demonstrated.
- With proper  $E_{INT}$  and  $N_0$ , nucleation-growth-coarsening models successfully approach  $\gamma'$  precipitation in binary and ternary system.
- The yield stress and young's modulus are calculated by empirical formulas and PyMKS package, respectively.
- The IRT model is implemented to simulating the plastic deformation.
- To optimize the chemical composition, Genetic Algorithm is used as the close loop of process-structure-properties.

#### Optimization

