UNIVERSITY OF CAMBRIDGE

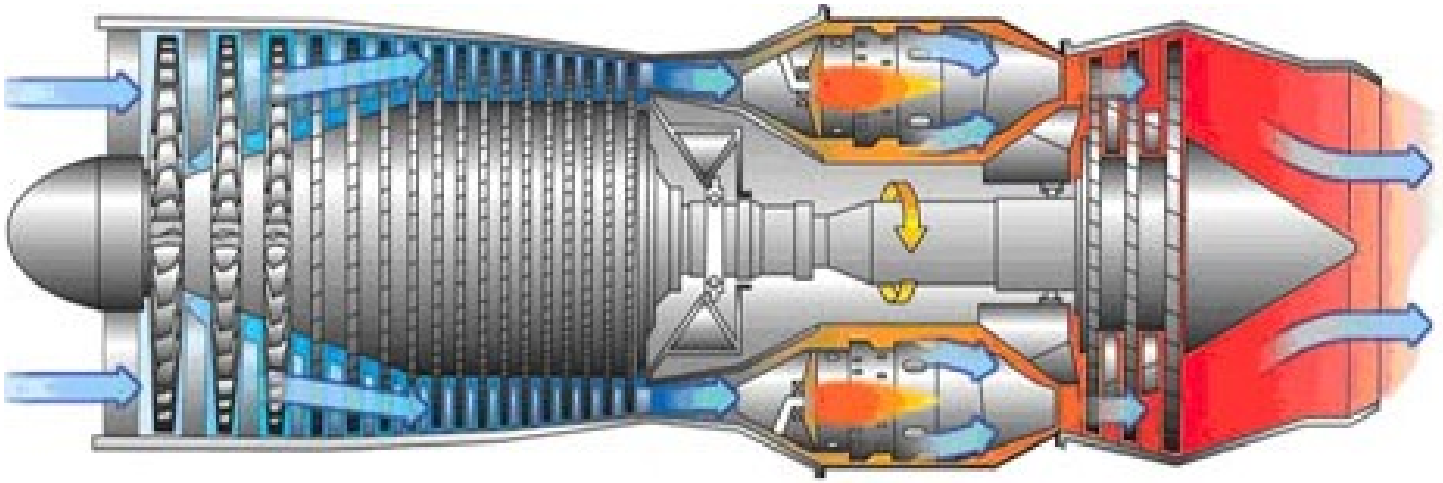
Concurrent materials design

Gareth Conduit

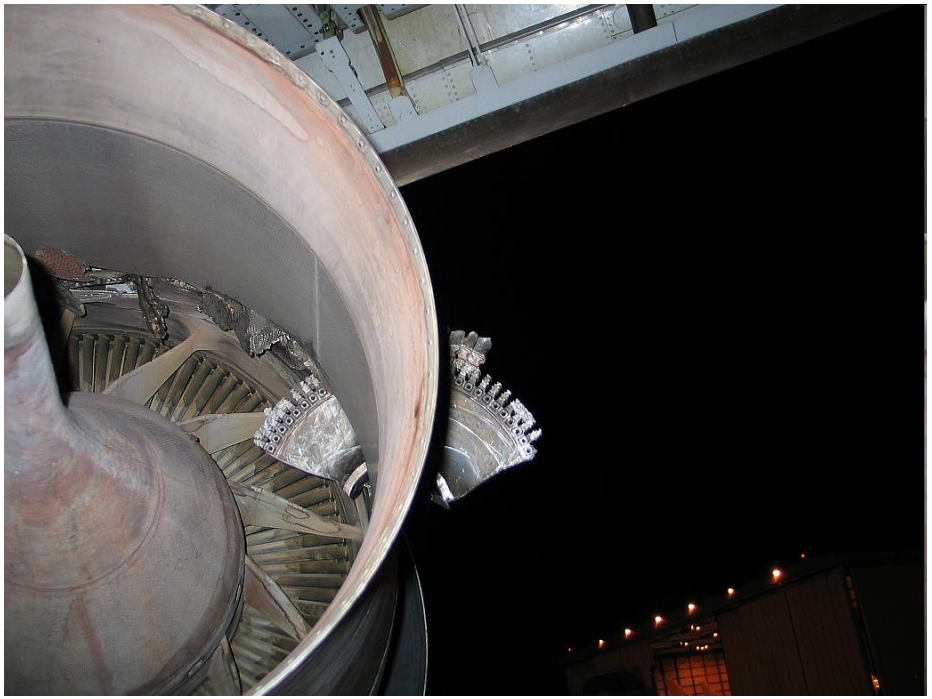
Patent GB1302743.8 (2013)
Patent GB1307533.8 (2013)
Patent GB1307535.3 (2013)
Acta Materialia, **61**, 3378 (2013)
Submitted to Intermetallics (2013)
Rolls-Royce Group plc invention submission NC13006 (2013)
Rolls-Royce Group plc invention submission NC12261 (2012)

TCM Group, Department of Physics

Jet engine

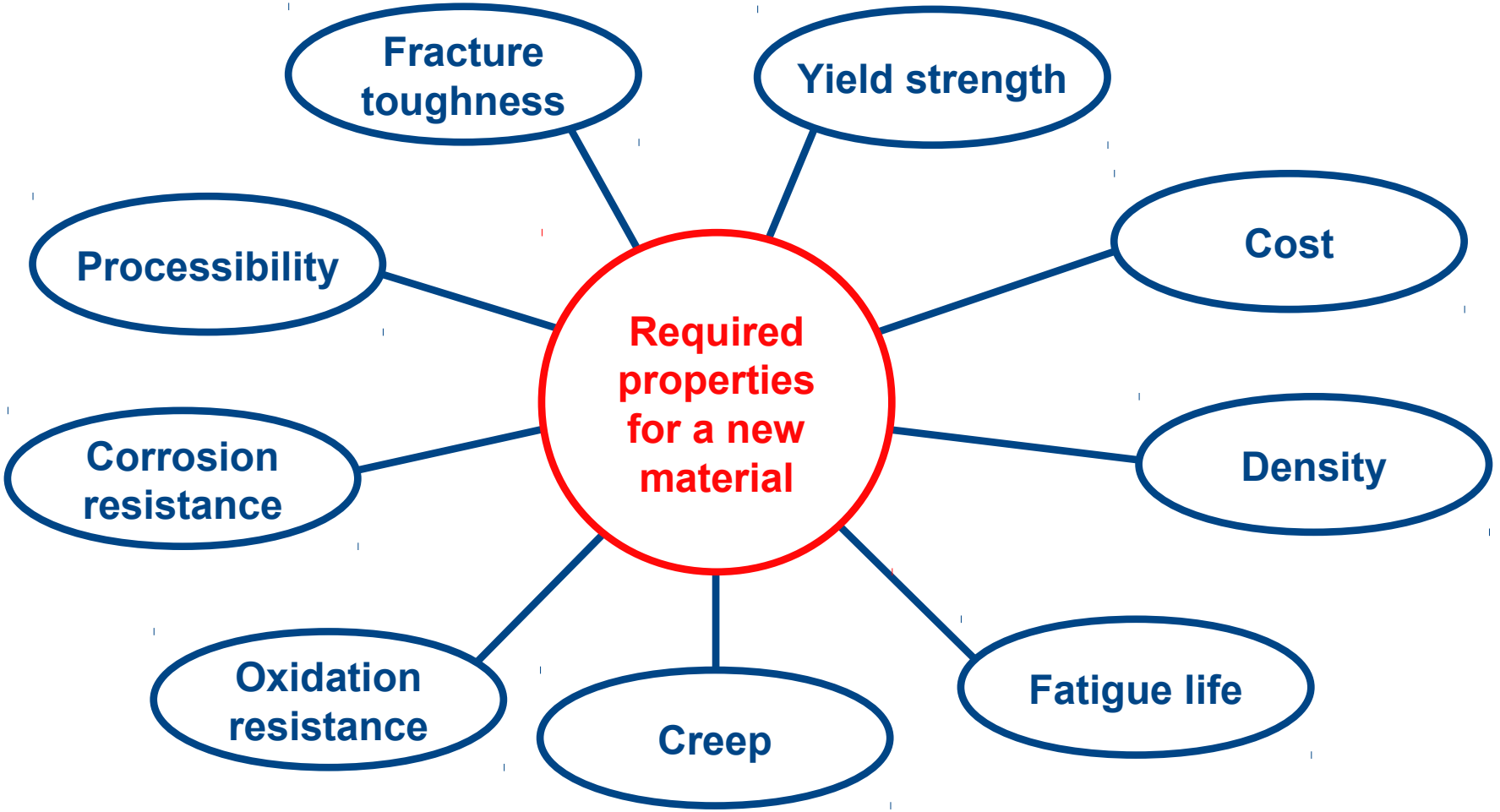


Jet engine: disc failure

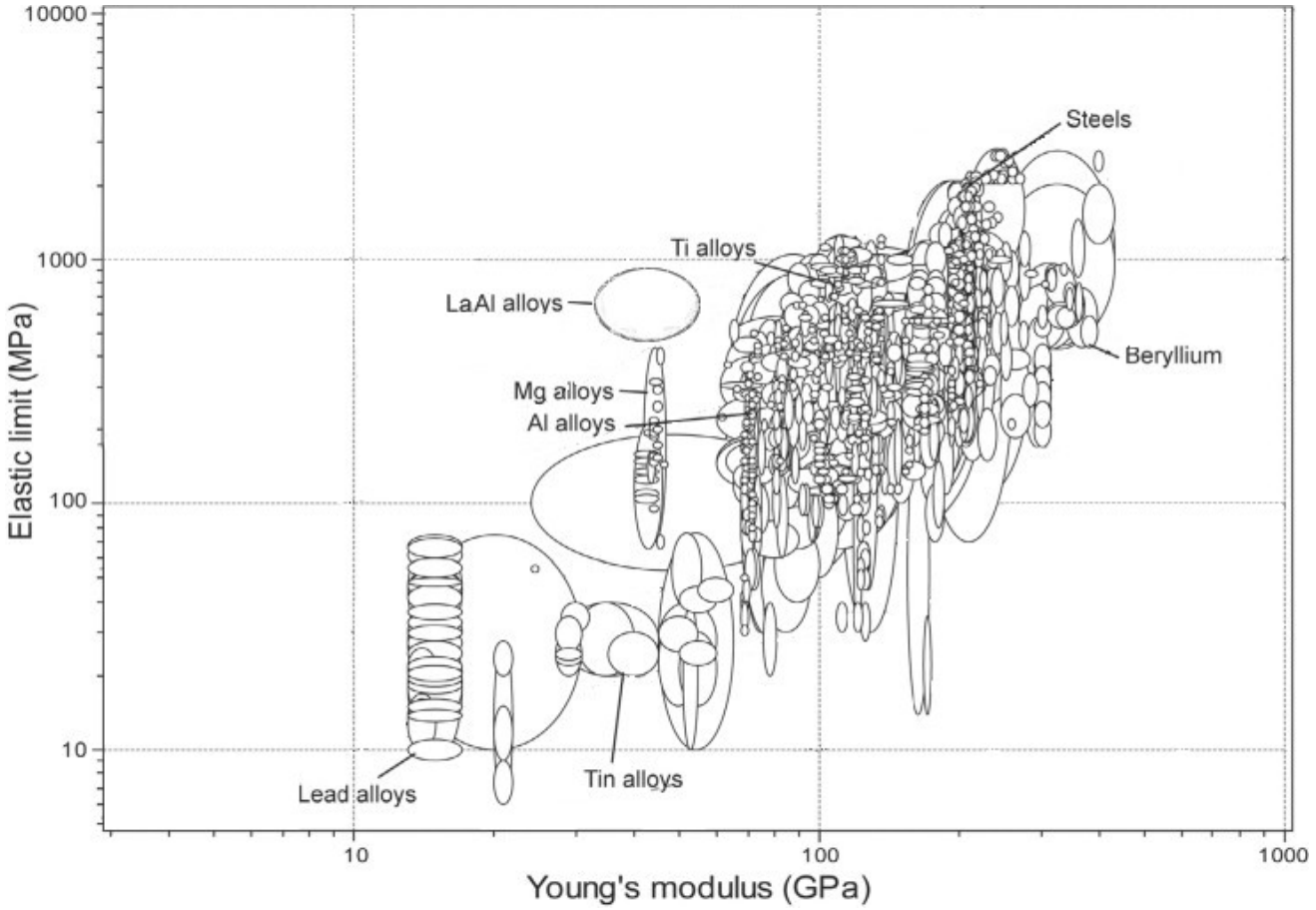


Jet engine: bird strike

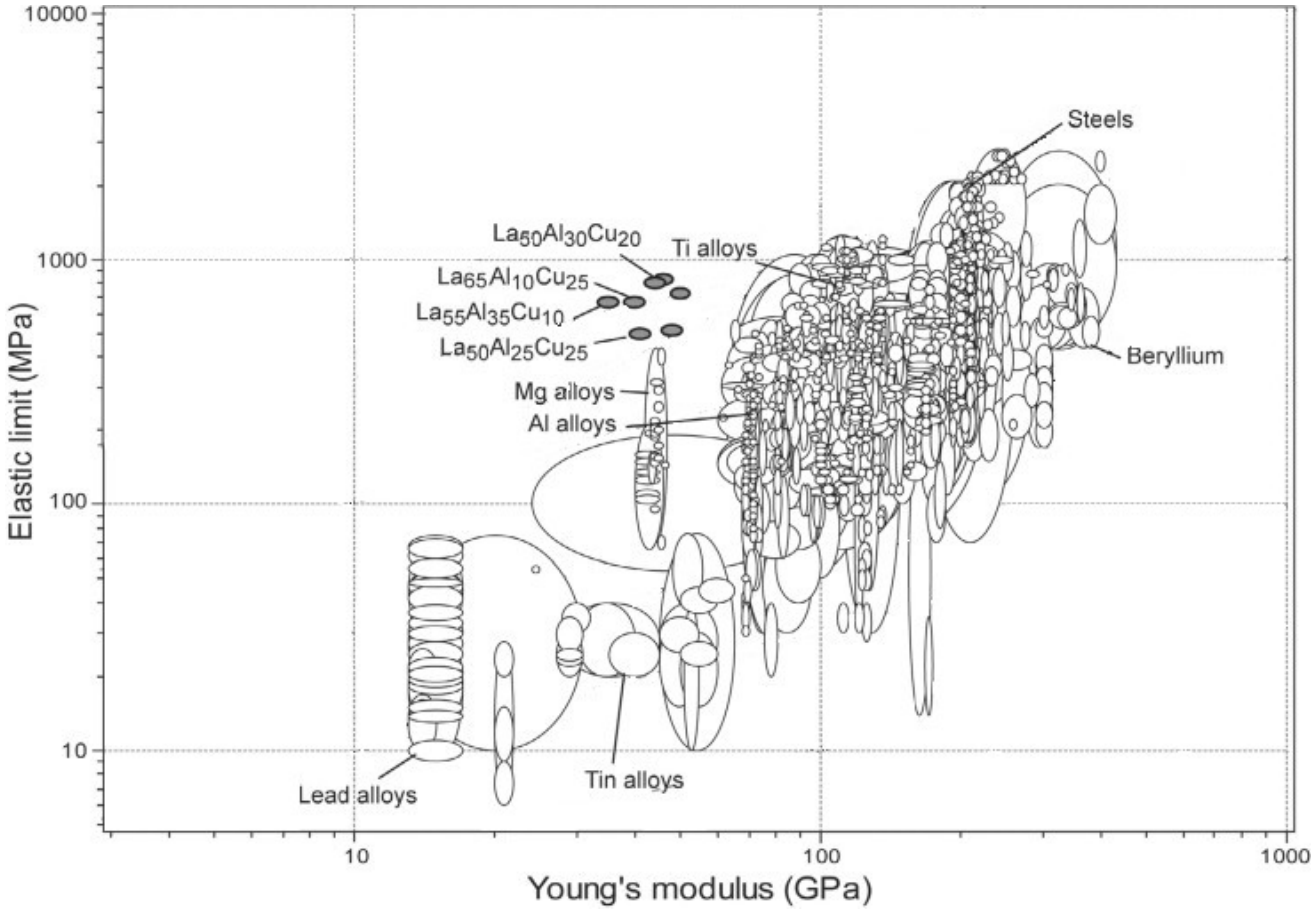
Designing a new material – what is required ?



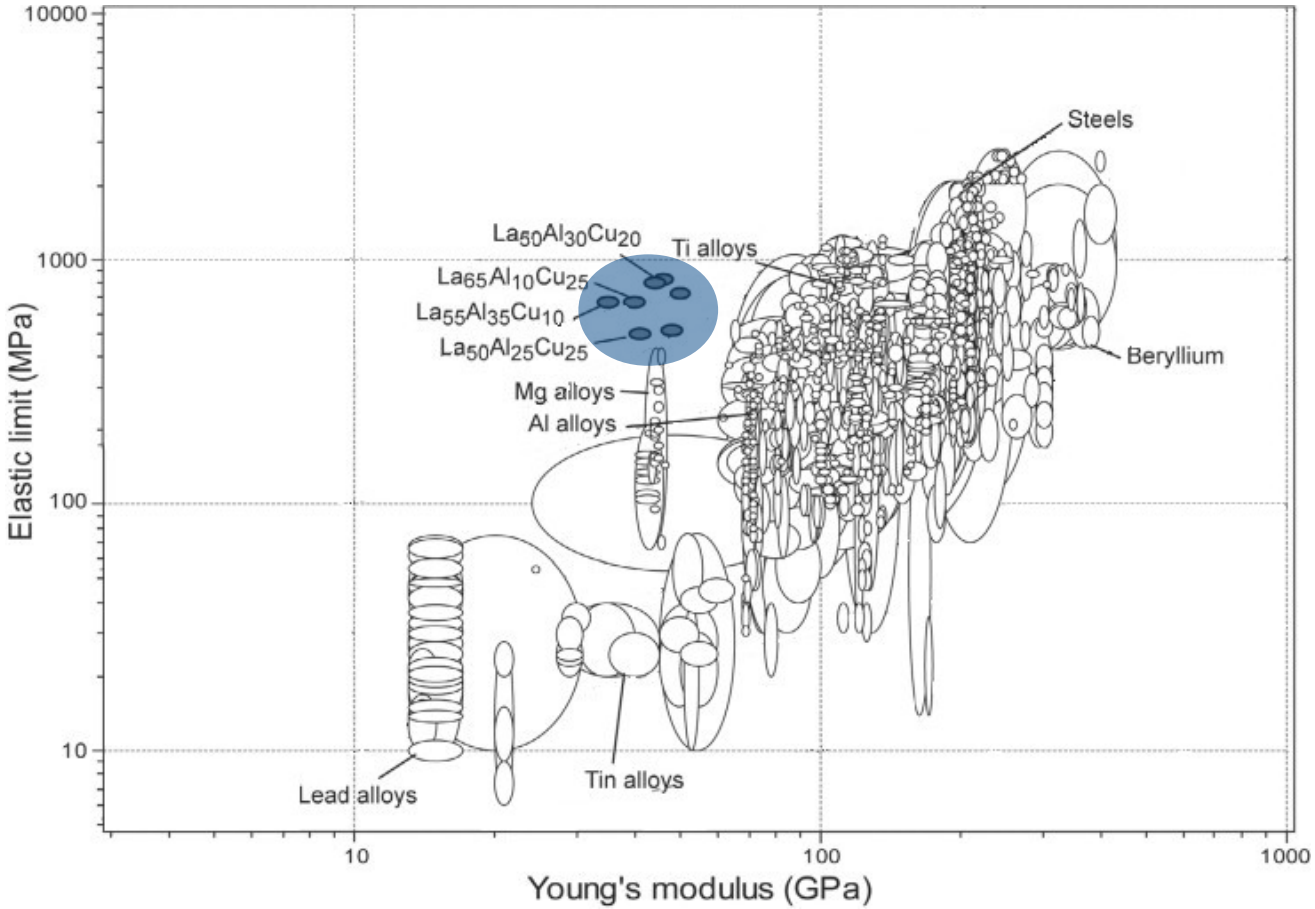
Materials selection



Materials selection



Materials selection



Materials selection

Guide the discovery of new materials

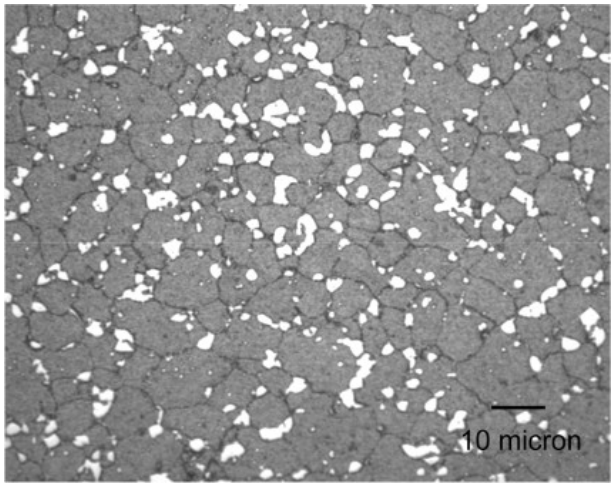
Concurrent materials design

Assess data quality

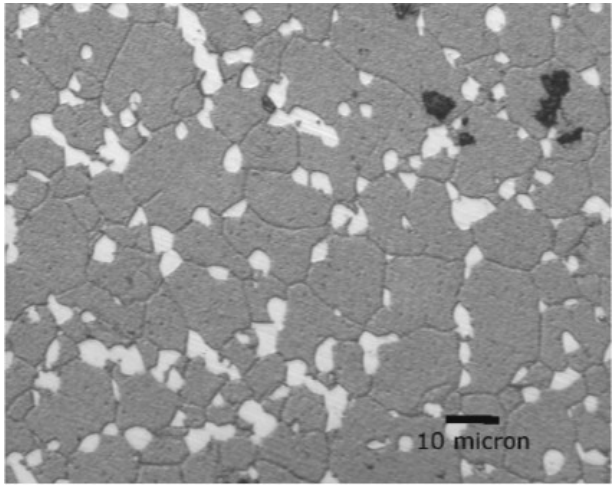
Estimate missing data

Contemporary alloys

RR1000

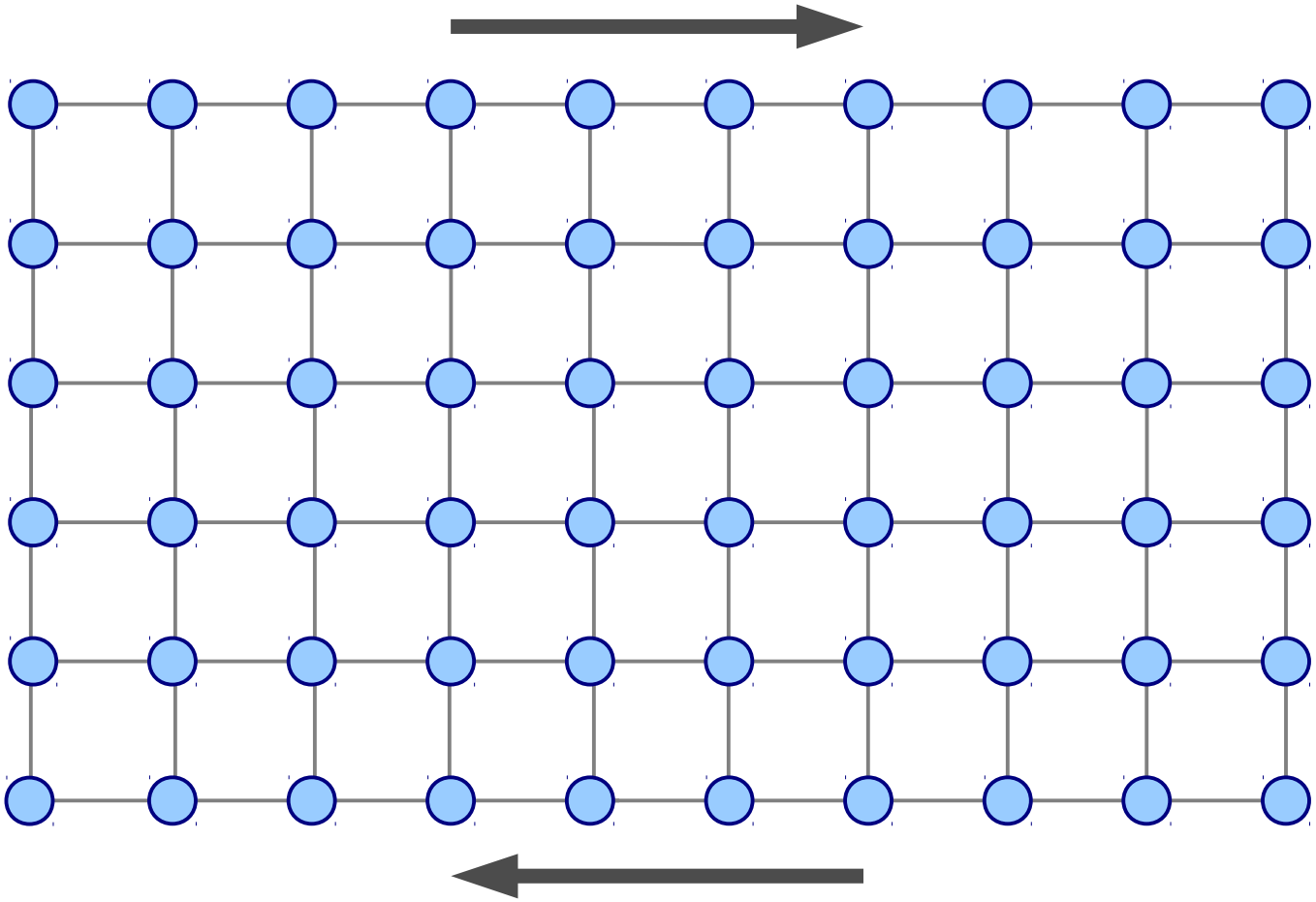


N18



Alloy	Firm	Ni	Cr	Co	Mo	Ti	Al	C	Hf	Ta	W	Nb
Waspaloy	UTC	58	19	13	4	3	1.4					
N18	SNECMA	58	11.1	15.4	6.4	4.3	4.3	0.02	0.5			
Rene 88	General Elec.	56.5	16	13	4	3.7	2.1	0.03			4	0.7
RR1000	Rolls Royce	52.4	15	18.5	5	3.6	3	0.03	0.5	2		

Creep



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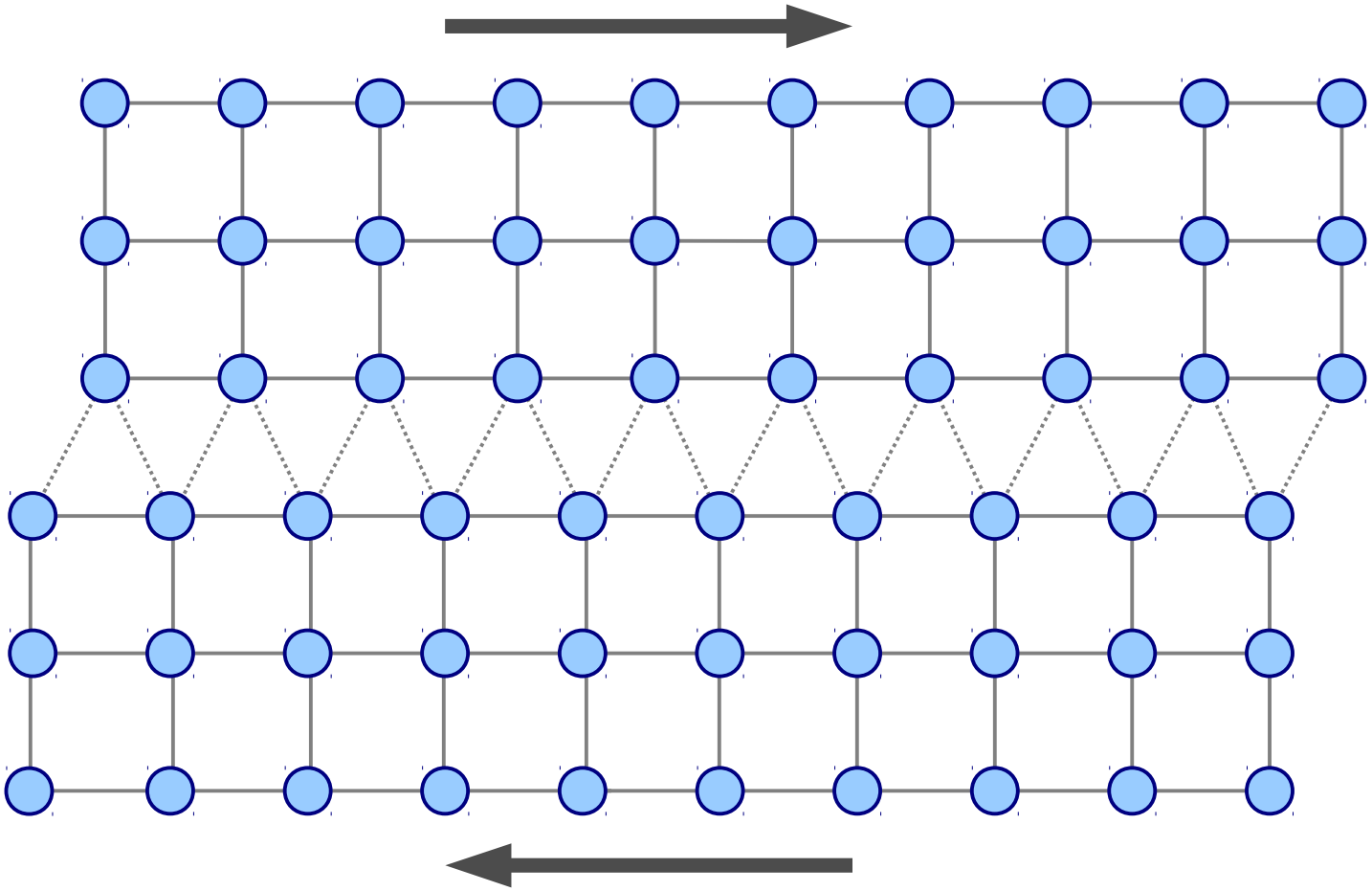

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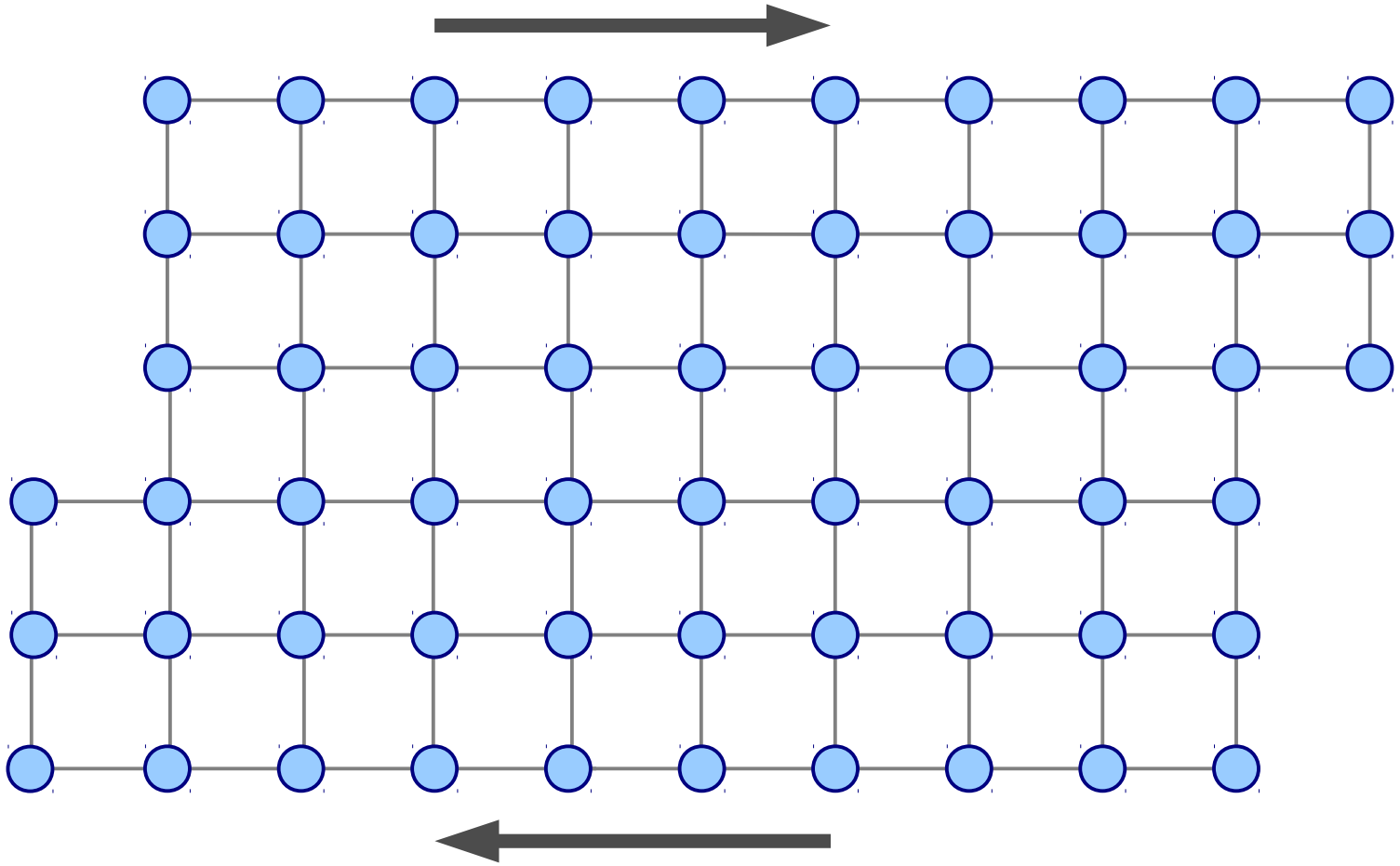

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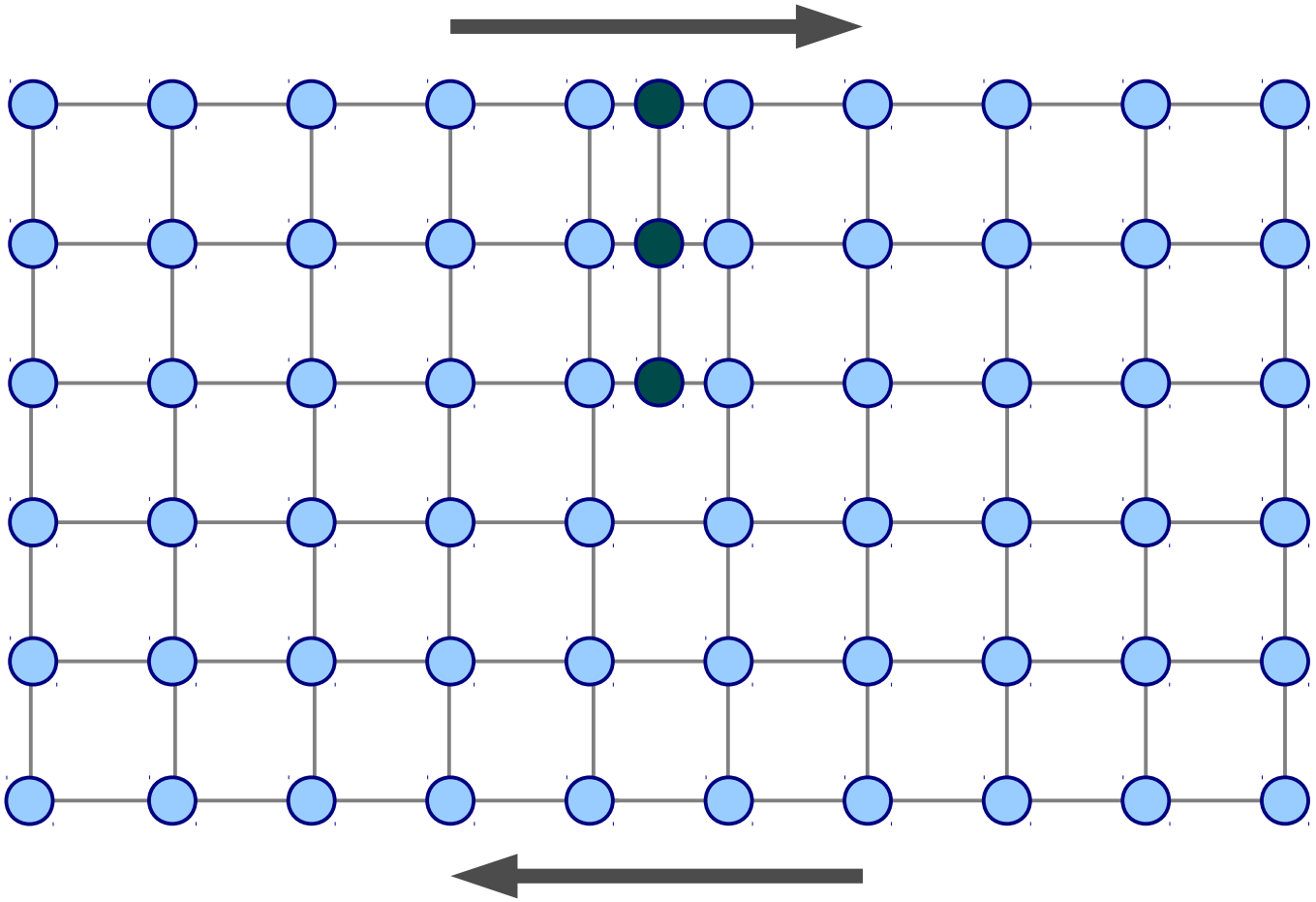

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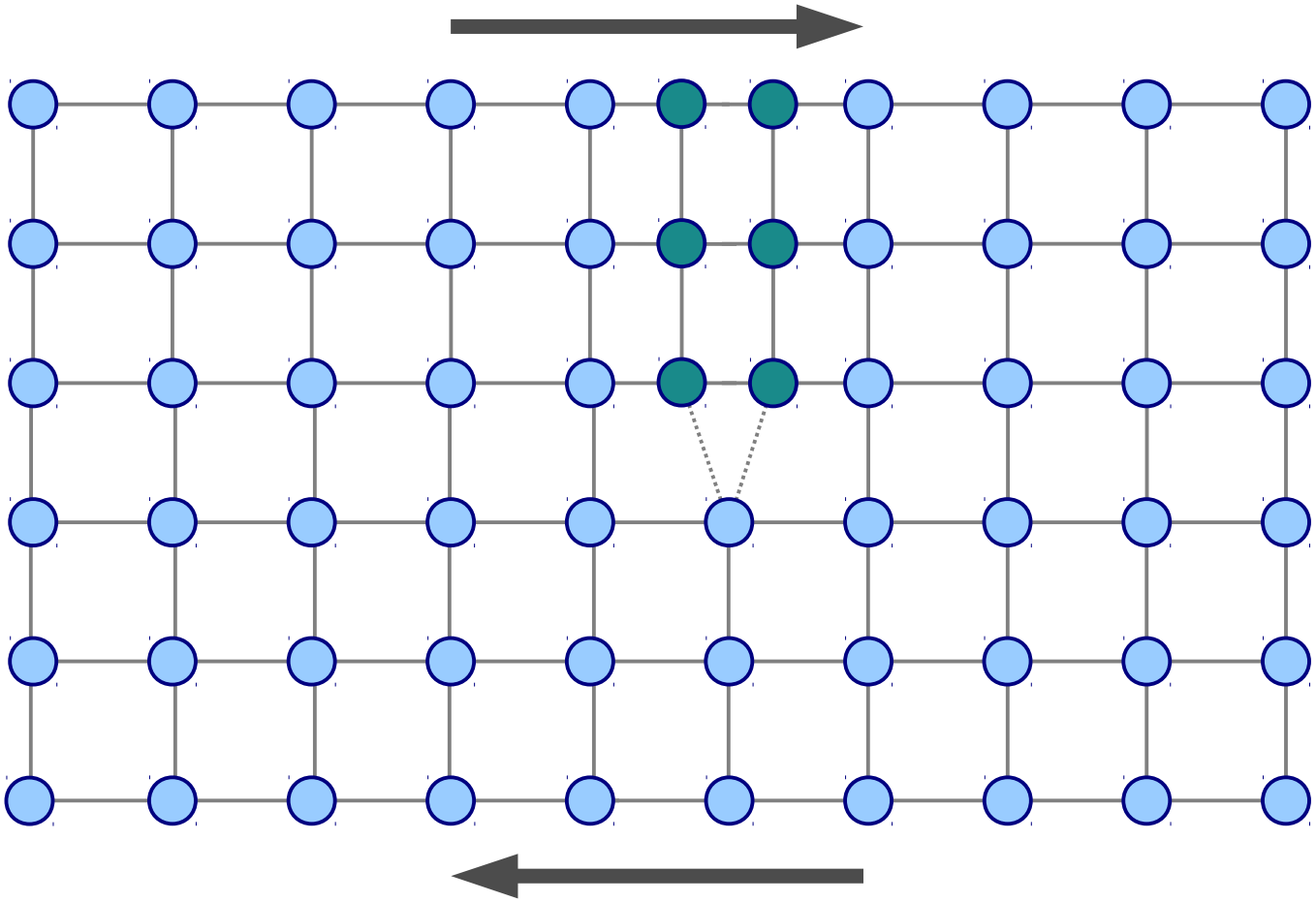

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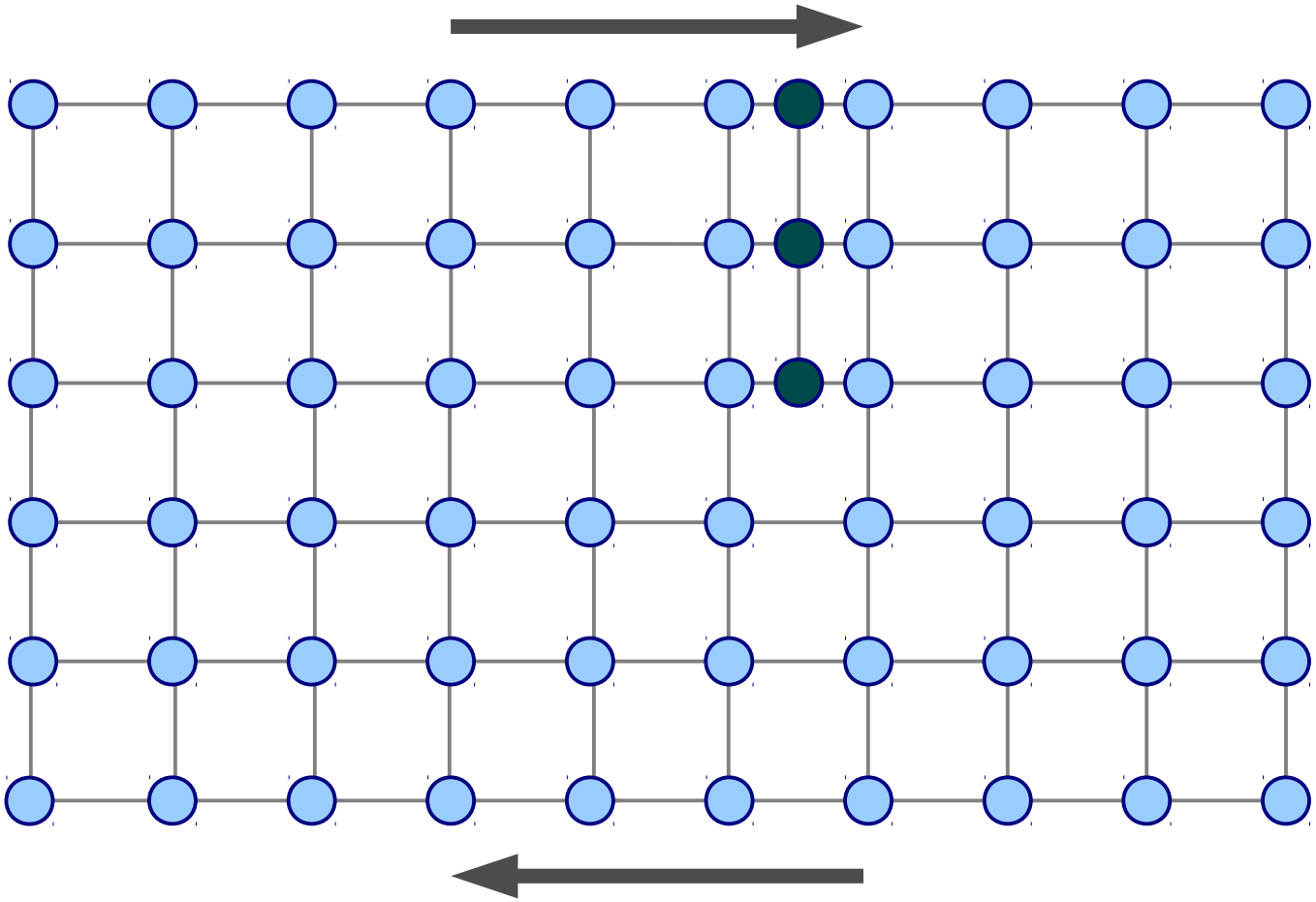

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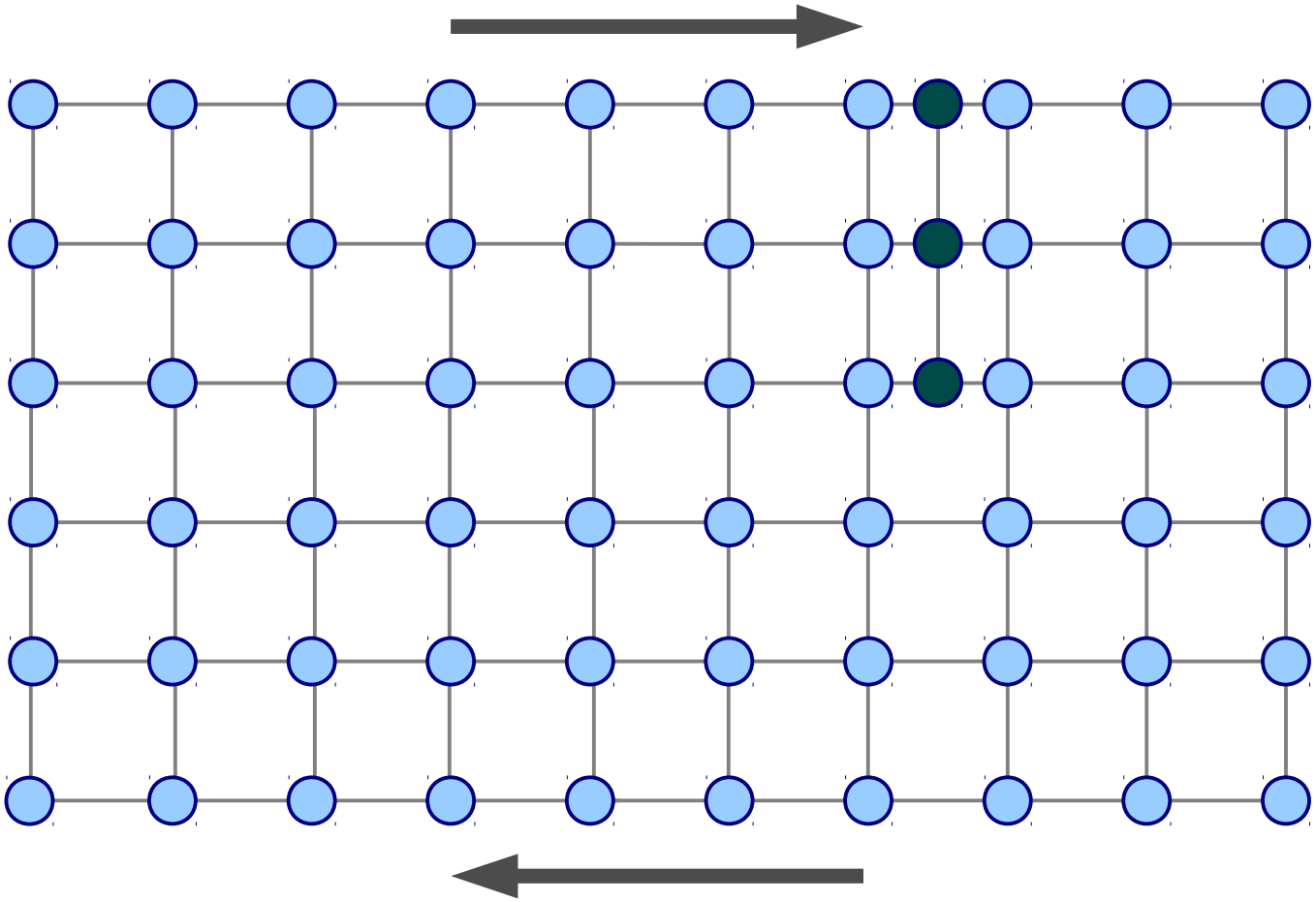

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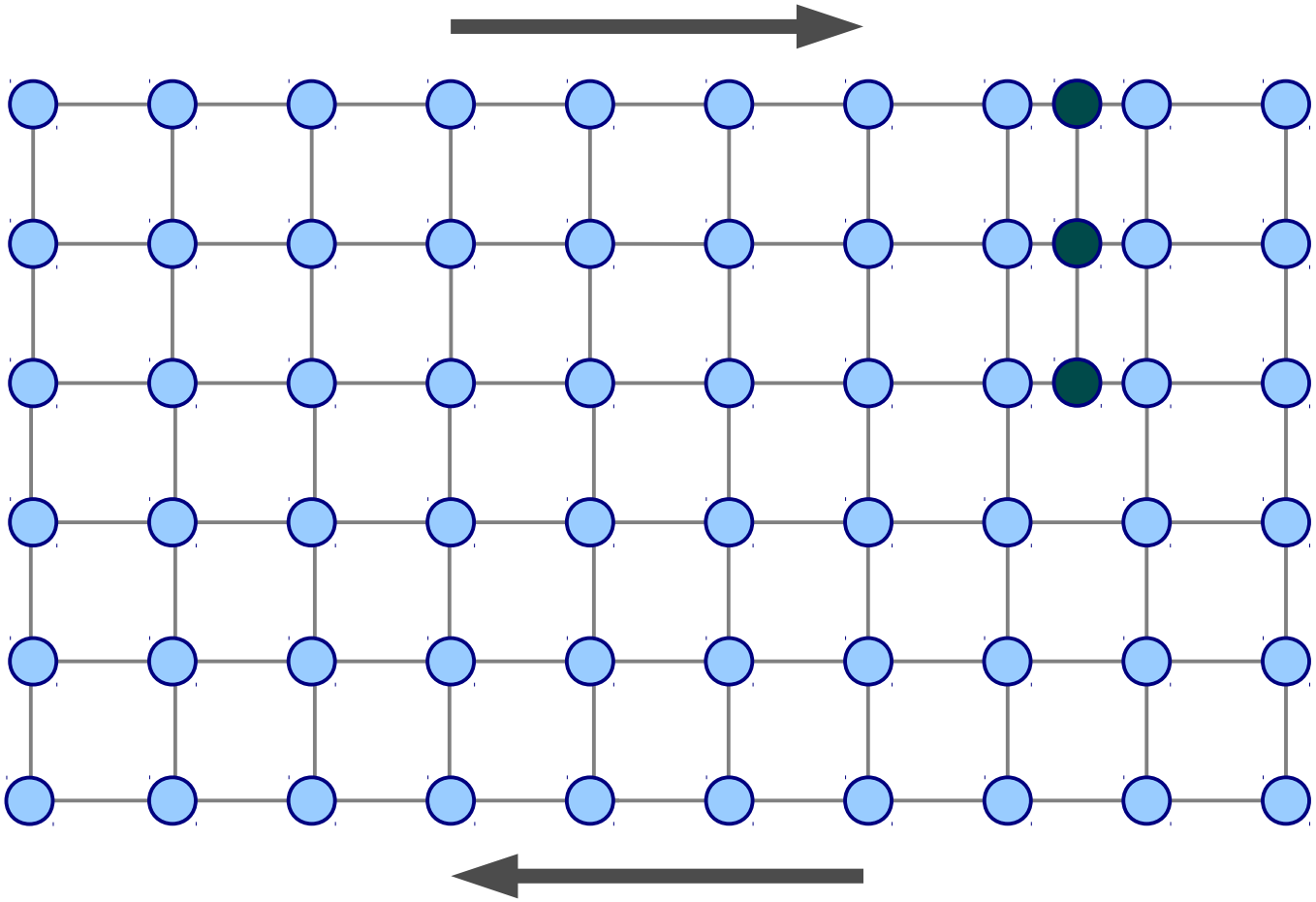

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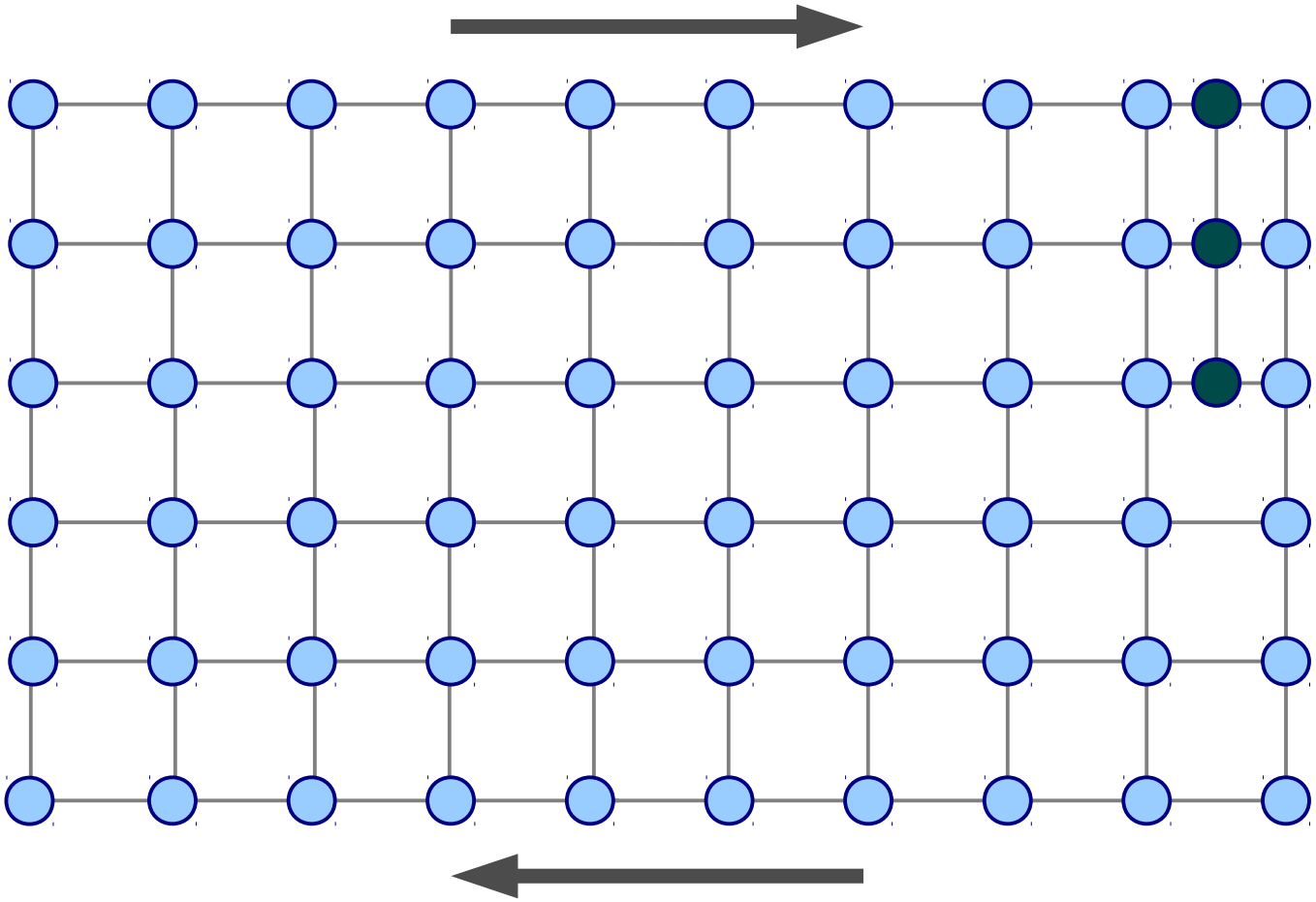

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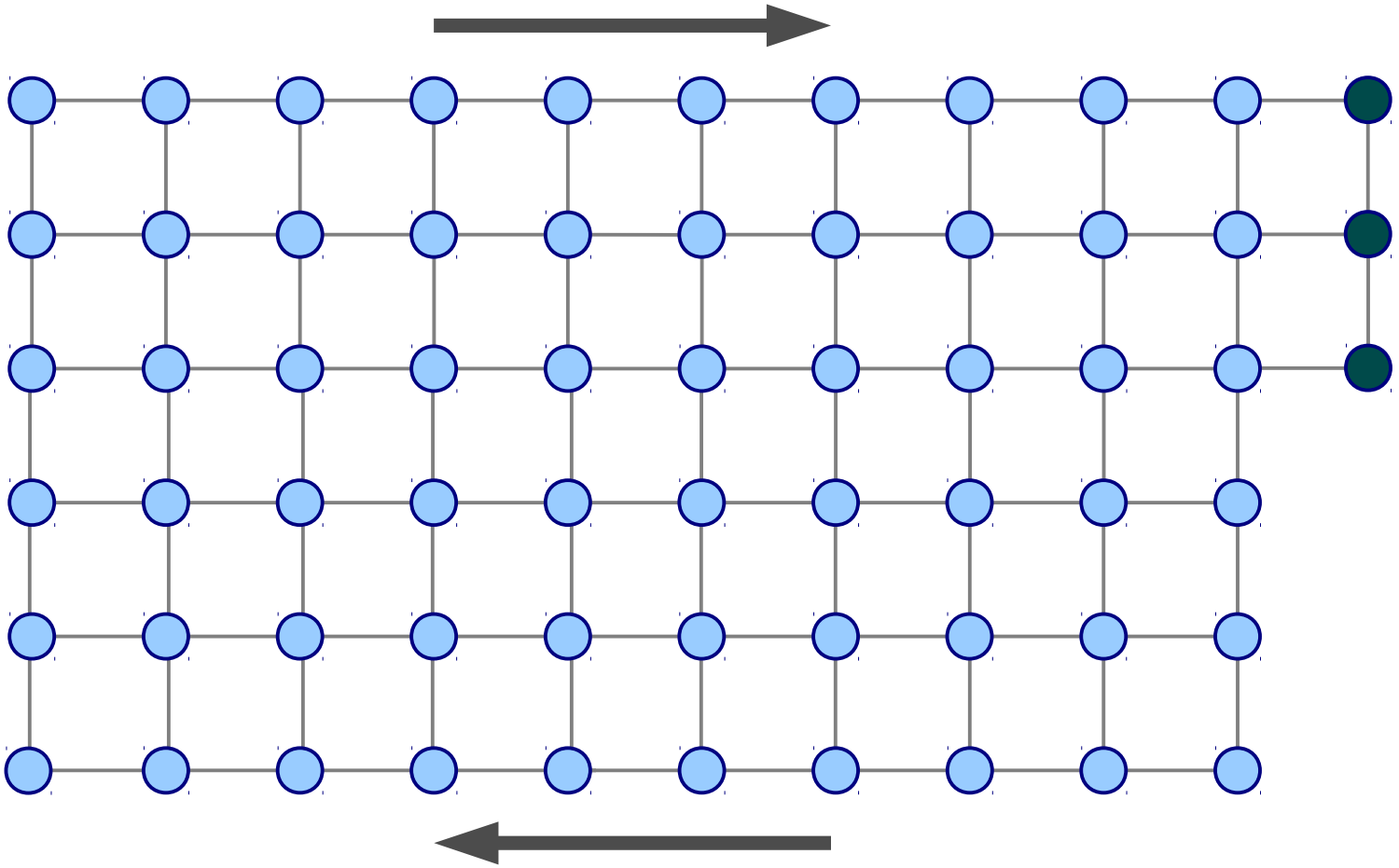
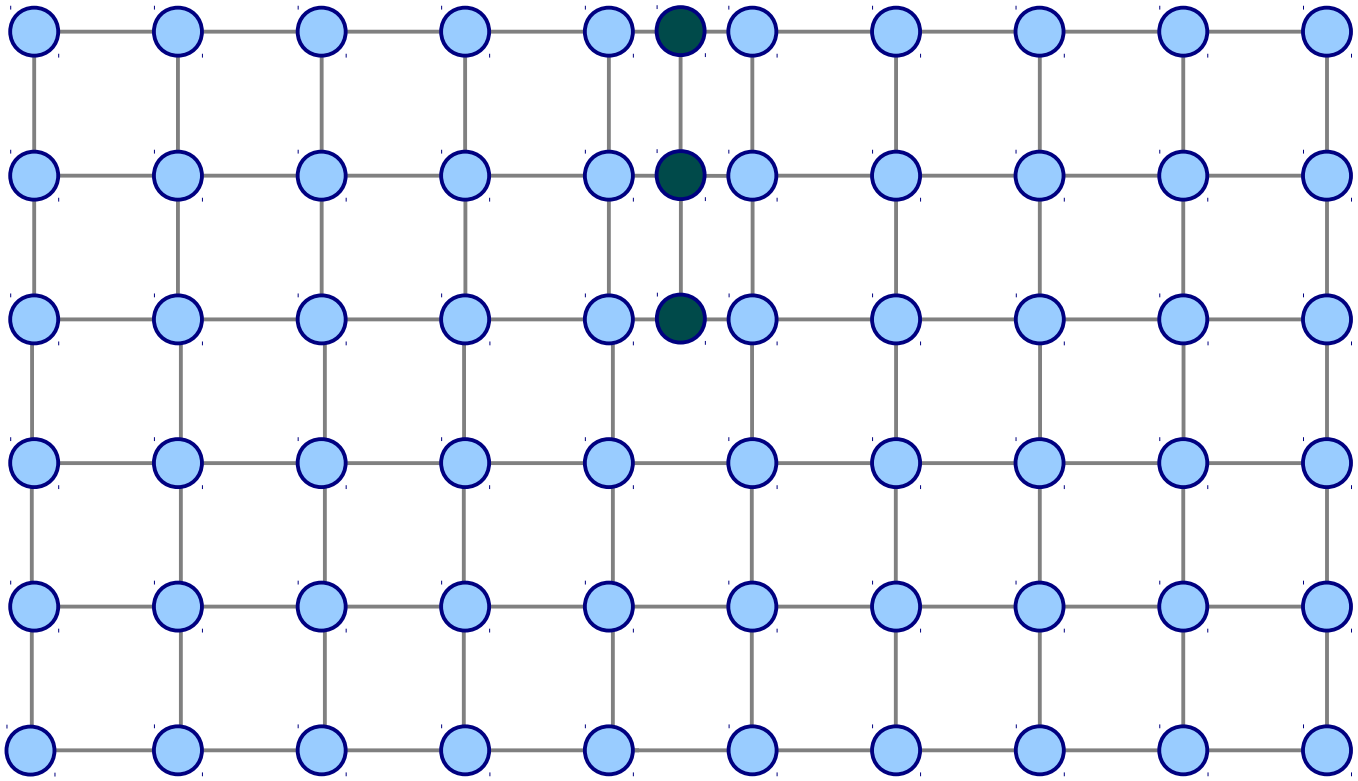


Solution hardening



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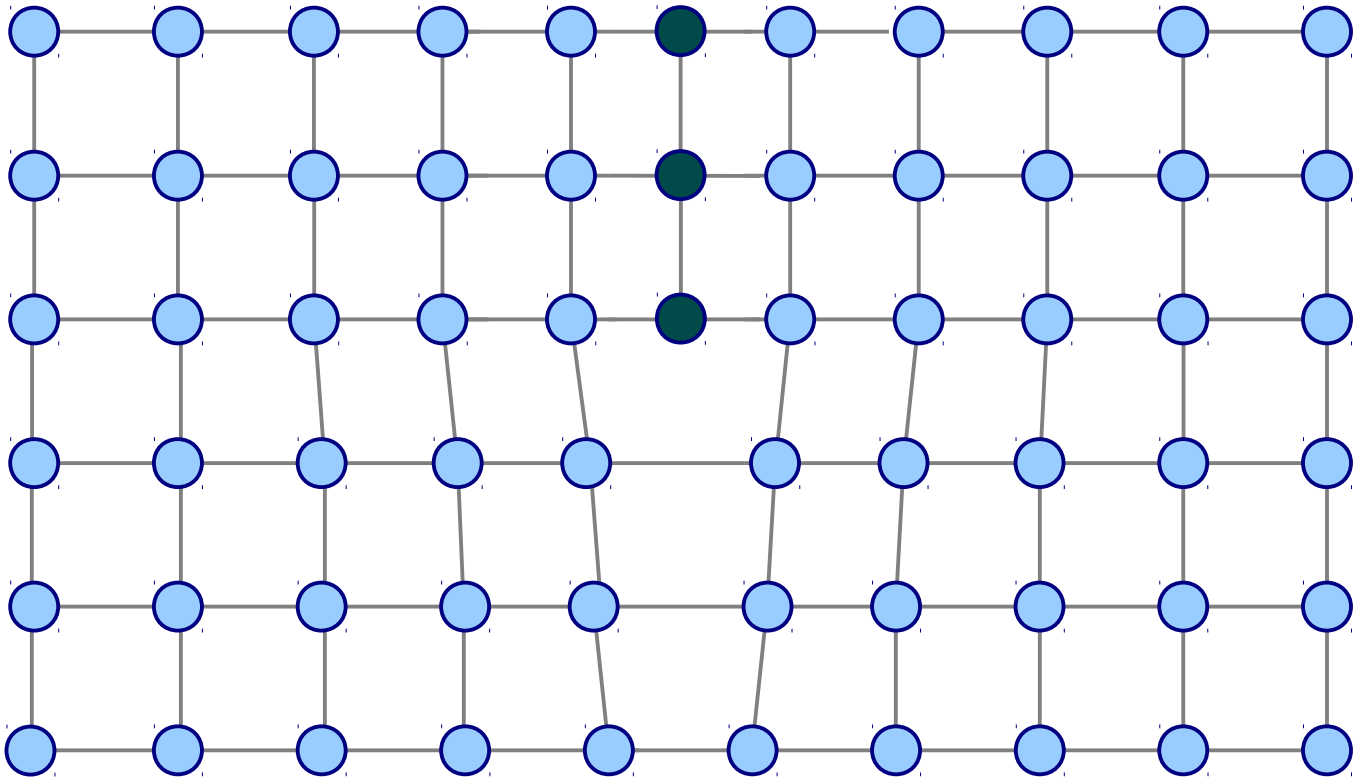

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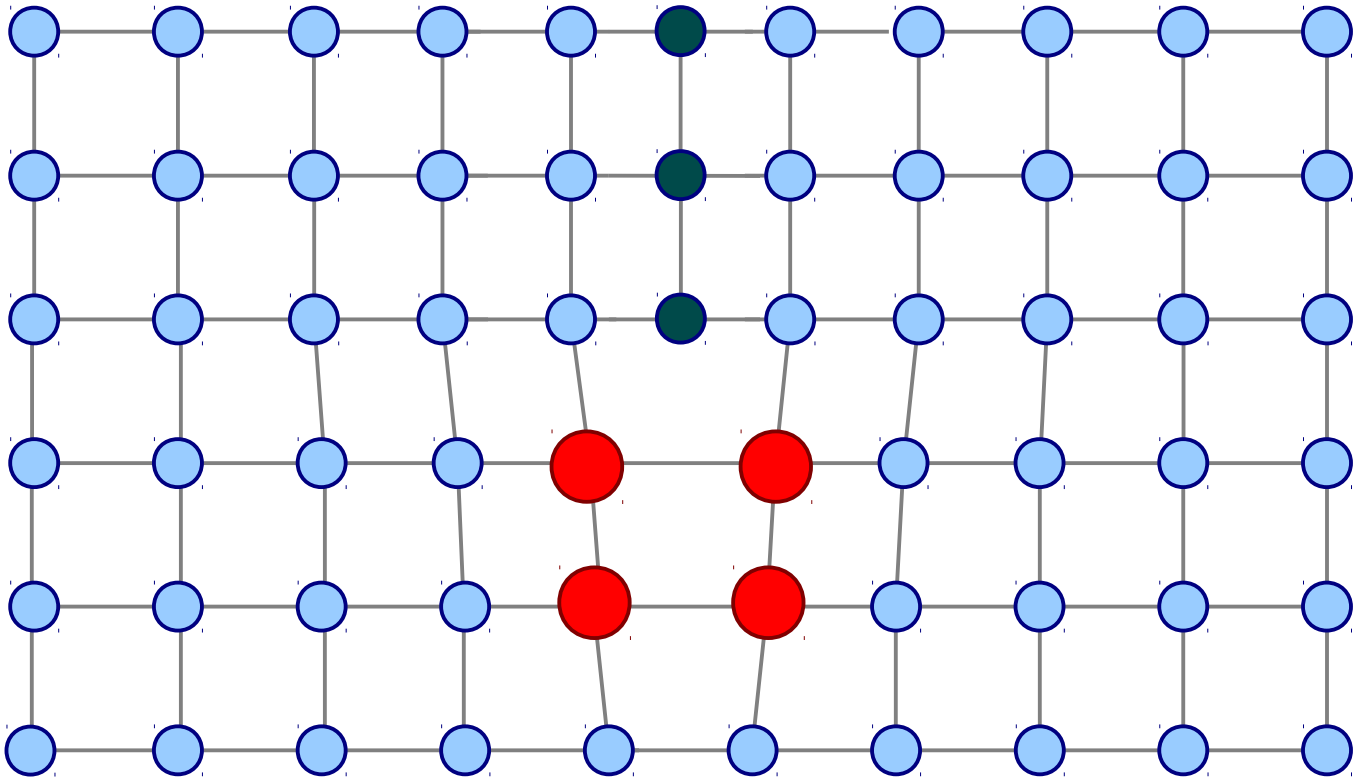

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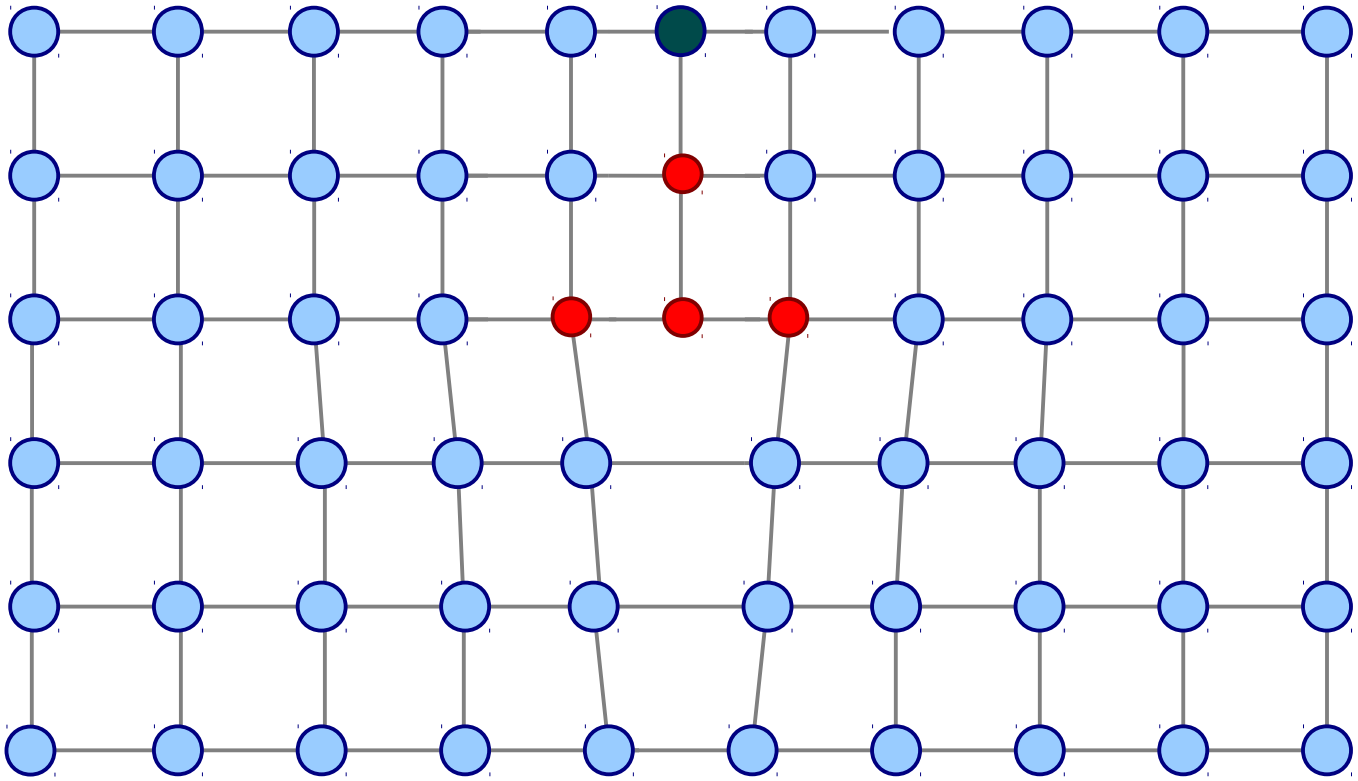
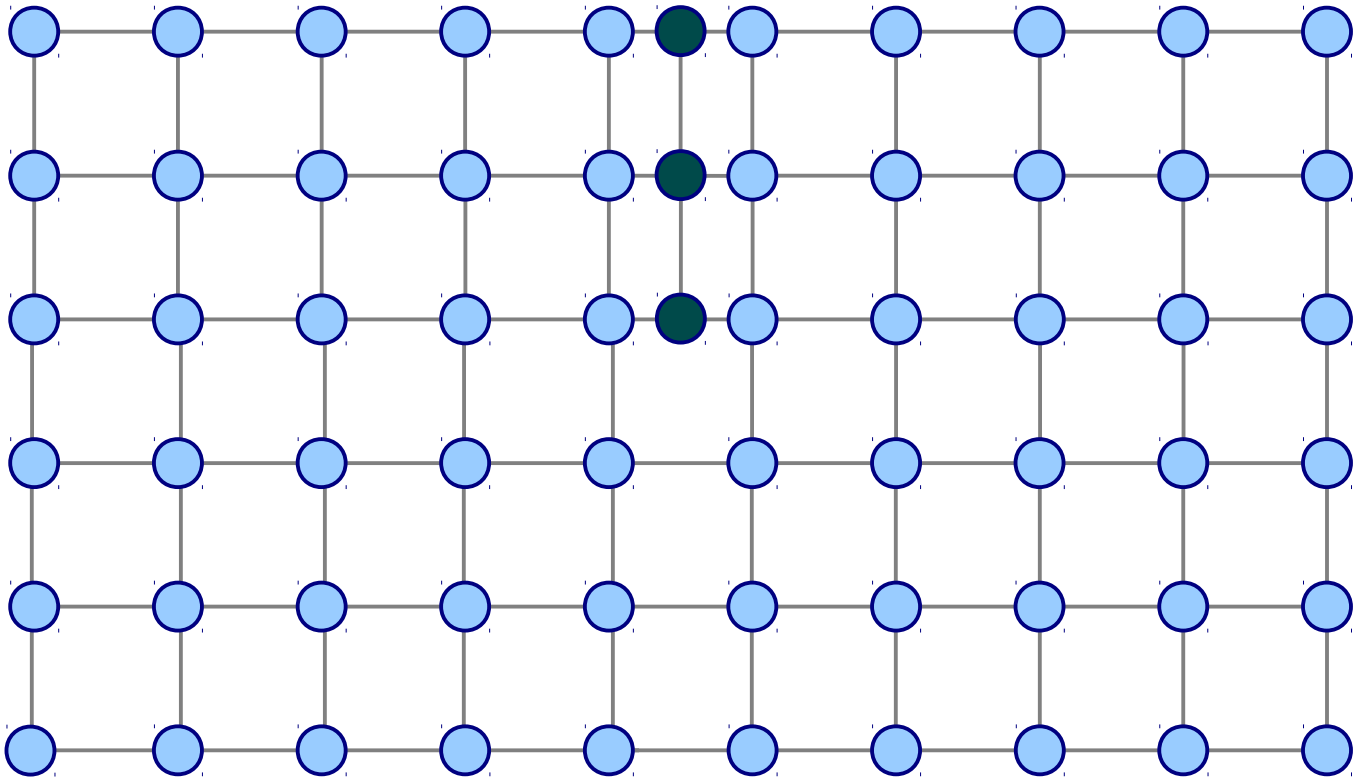


Precipitate hardening



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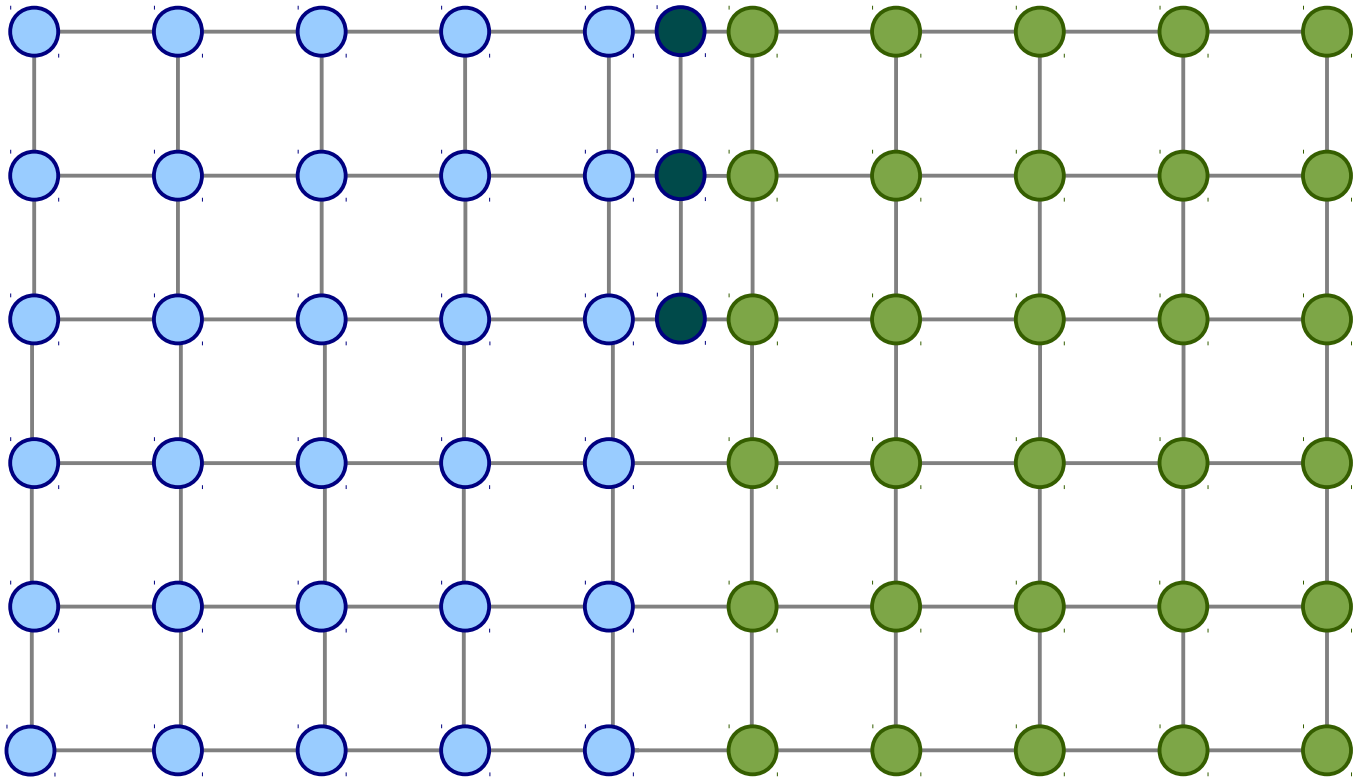
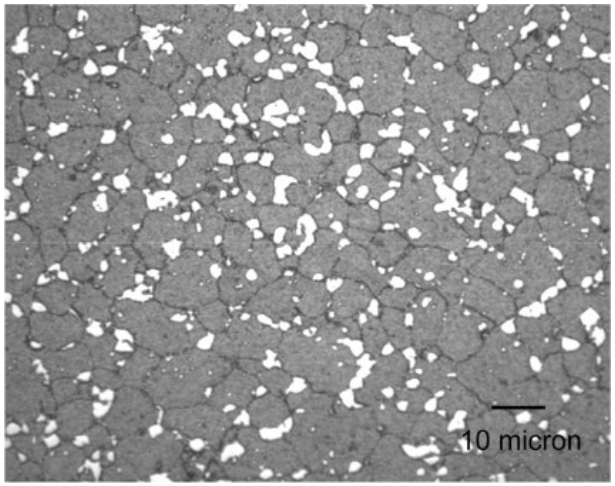


Contemporary alloys

RR1000



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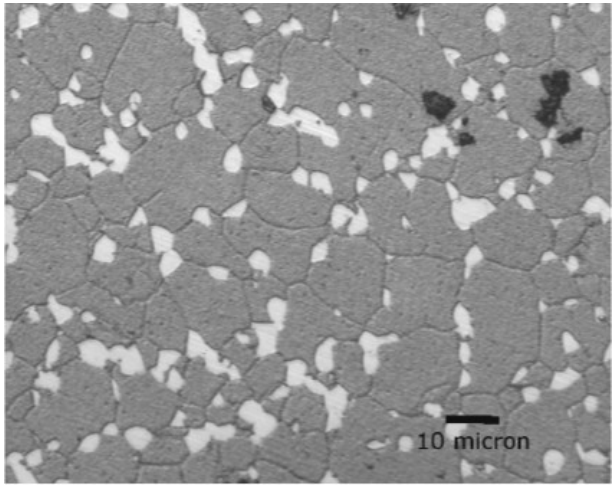

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Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

High cycle fatigue

Weldability

Creep model

Properties

Cost $lb^{-1}
γ' fraction
Stability
Density gcm^{-3}
Yield stress MPa
UTS MPa
Oxidation index
Stress rupture MPa
Resistivity μΩcm
Entropy Jmol^{-1}K^{-1}
Low cycle fatigue
High cycle fatigue
Weldability
Creep model

$$\text{Cost}[\$/\text{lb}] = 9.59n_{ni} + 0.94n_{Al} + 6.77n_{Cr} + 16.5n_{Co} + 19.6n_{Mo} + 5.44n_{Ti}$$

Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

High cycle fatigue

Weldability

Creep model

Collect data for yield stress from 2248 alloys

Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

High cycle fatigue

Weldability

Creep model

Collect data for yield stress from 2248 alloys

Generate neural network model

$$YS[\mathrm{MPa}] = F(n_{ni}, n_{Al}, n_{Cr}, n_{Co}, n_{Mo}, n_{Ti}, T_{HT}, t_{HT})$$

Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

High cycle fatigue

Weldability

Creep model

Collect data for yield stress from 2248 alloys

Generate neural network model

$$YS[\mathrm{MPa}] = F(n_{ni}, n_{Al}, n_{Cr}, n_{Co}, n_{Mo}, n_{Ti}, T_{HT}, t_{HT})$$

Calculate uncertainty in neural network model

Properties

Cost $lb^{-1}
γ' fraction
Stability
Density gcm^{-3}
Yield stress MPa
UTS MPa
Oxidation index
Stress rupture MPa
Resistivity μΩcm
Entropy Jmol^{-1}K^{-1}
Low cycle fatigue
High cycle fatigue
Weldability
Creep model

Collect data for yield stress from 2248 alloys

Generate neural network model

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Properties

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γ' fraction
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Yield stress MPa
UTS MPa
Oxidation index
Stress rupture MPa
Resistivity μΩcm
Entropy Jmol^{-1}K^{-1}
Low cycle fatigue
High cycle fatigue
Weldability
Creep model

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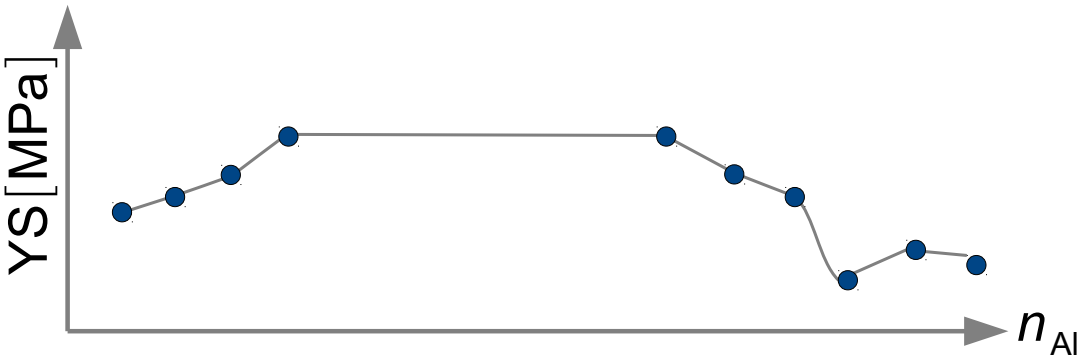

Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

High cycle fatigue

Weldability

Creep model

Collect data for yield stress from 2248 alloys

Generate neural network model

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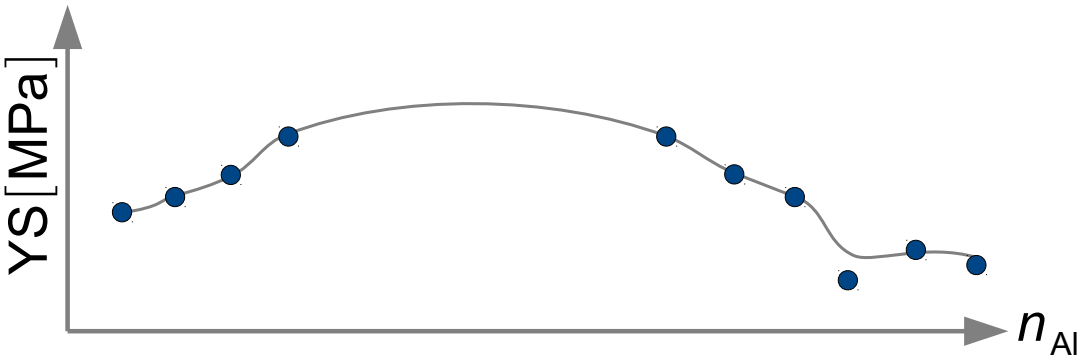
Calculate uncertainty in neural network model



Properties

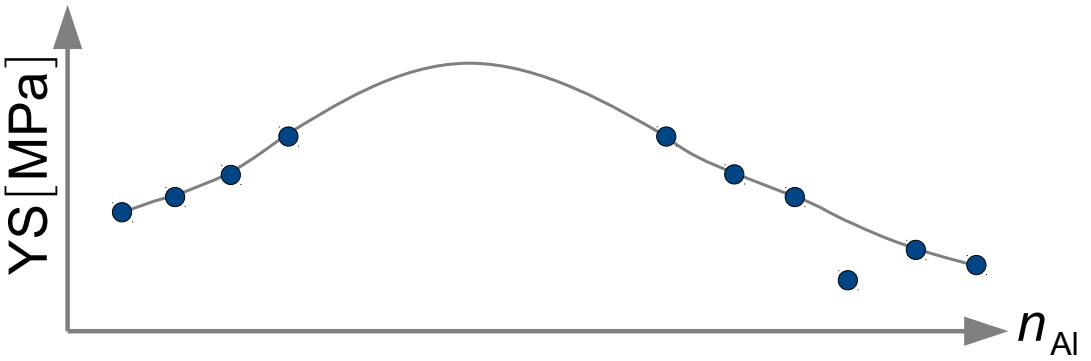
Cost $lb^{-1}
γ' fraction
Stability
Density gcm^{-3}
Yield stress MPa
UTS MPa
Oxidation index
Stress rupture MPa
Resistivity μΩcm
Entropy Jmol^{-1}K^{-1}
Low cycle fatigue
High cycle fatigue
Weldability
Creep model

Collect data for yield stress from 2248 alloys

Generate neural network model

$$YS[\mathrm{MPa}] = F(n_{ni}, n_{Al}, n_{Cr}, n_{Co}, n_{Mo}, n_{Ti}, T_{HT}, t_{HT})$$

Calculate uncertainty in neural network model



Properties

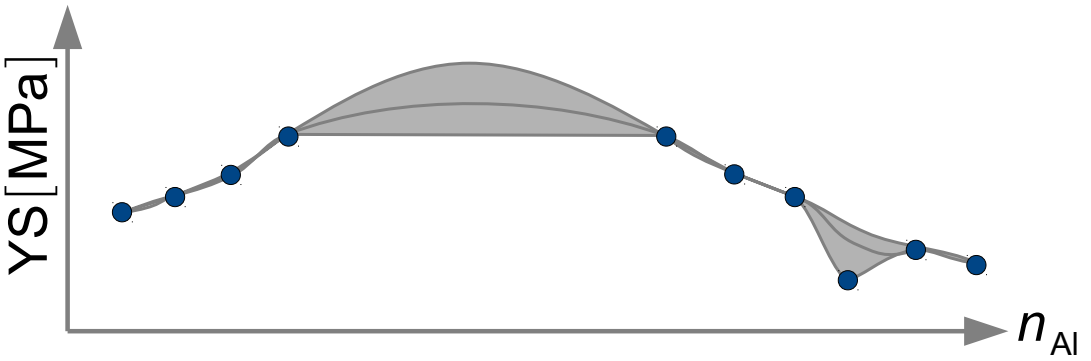
Cost $lb^{-1}
γ' fraction
Stability
Density gcm^{-3}
Yield stress MPa
UTS MPa
Oxidation index
Stress rupture MPa
Resistivity μΩcm
Entropy Jmol^{-1}K^{-1}
Low cycle fatigue
High cycle fatigue
Weldability
Creep model

Collect data for yield stress from 2248 alloys

Generate neural network model

$$YS[\mathrm{MPa}] = F(n_{ni}, n_{Al}, n_{Cr}, n_{Co}, n_{Mo}, n_{Ti}, T_{HT}, t_{HT})$$

Calculate uncertainty in neural network model



Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

High cycle fatigue

Weldability

Creep model

Calculate grid of

$$F_{(\gamma,\gamma')}(n_{ni}, n_{Al}, n_{Cr}, n_{Co}, n_{Mo}, n_{Ti})$$

Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

High cycle fatigue

Weldability

Creep model

Calculate grid of

$$F_{(\gamma,\gamma')}(n_{ni}, n_{Al}, n_{Cr}, n_{Co}, n_{Mo}, n_{Ti})$$

Generate neural network models

Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

High cycle fatigue

Weldability

Creep model

Calculate grid of

$$F_{(\gamma,\gamma')}(n_{ni}, n_{Al}, n_{Cr}, n_{Co}, n_{Mo}, n_{Ti})$$

Generate neural network models

Calculate phase equilibrium and uncertainty

Properties

Cost $lb^{-1}

γ' fraction

Stability

Density gcm^{-3}

Yield stress MPa

UTS MPa

Oxidation index

Stress rupture MPa

Resistivity μΩcm

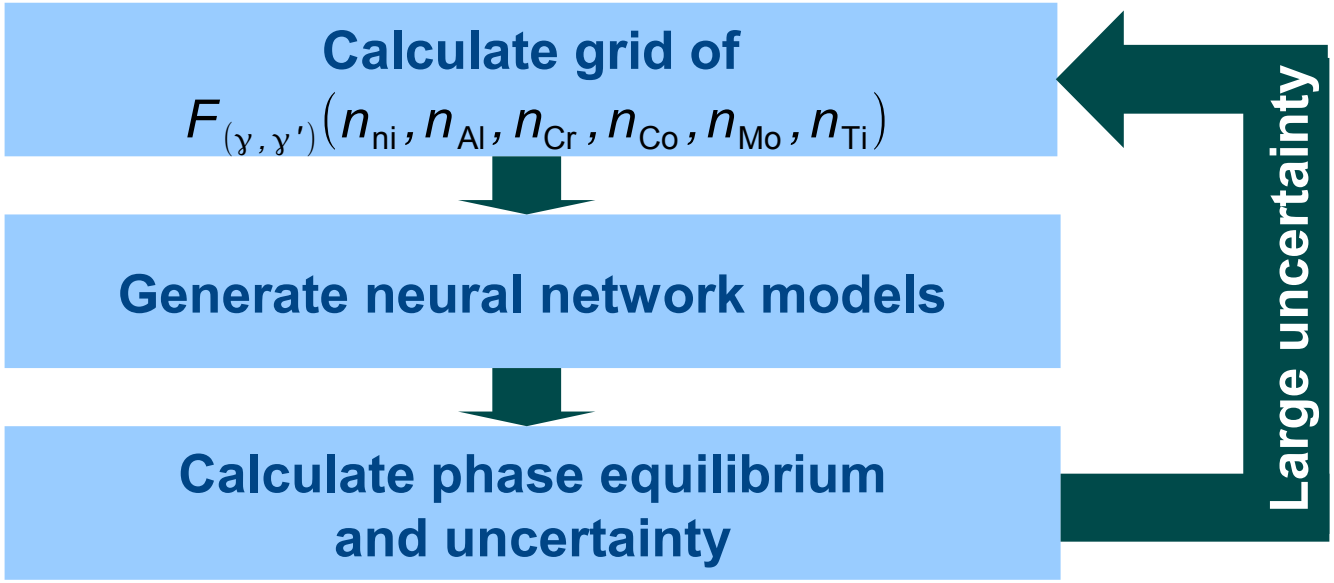
Entropy Jmol^{-1}K^{-1}

Low cycle fatigue

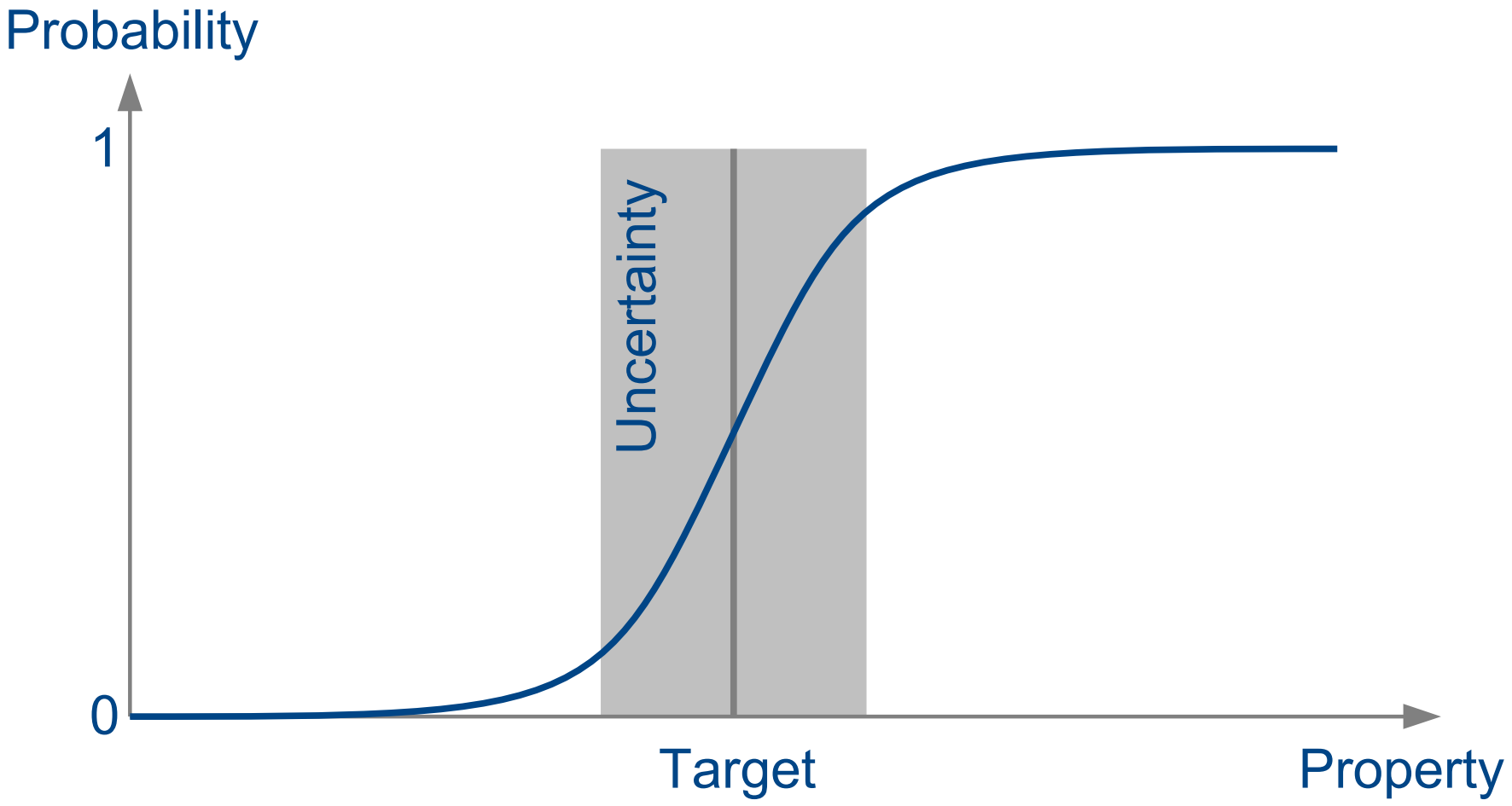
High cycle fatigue

Weldability

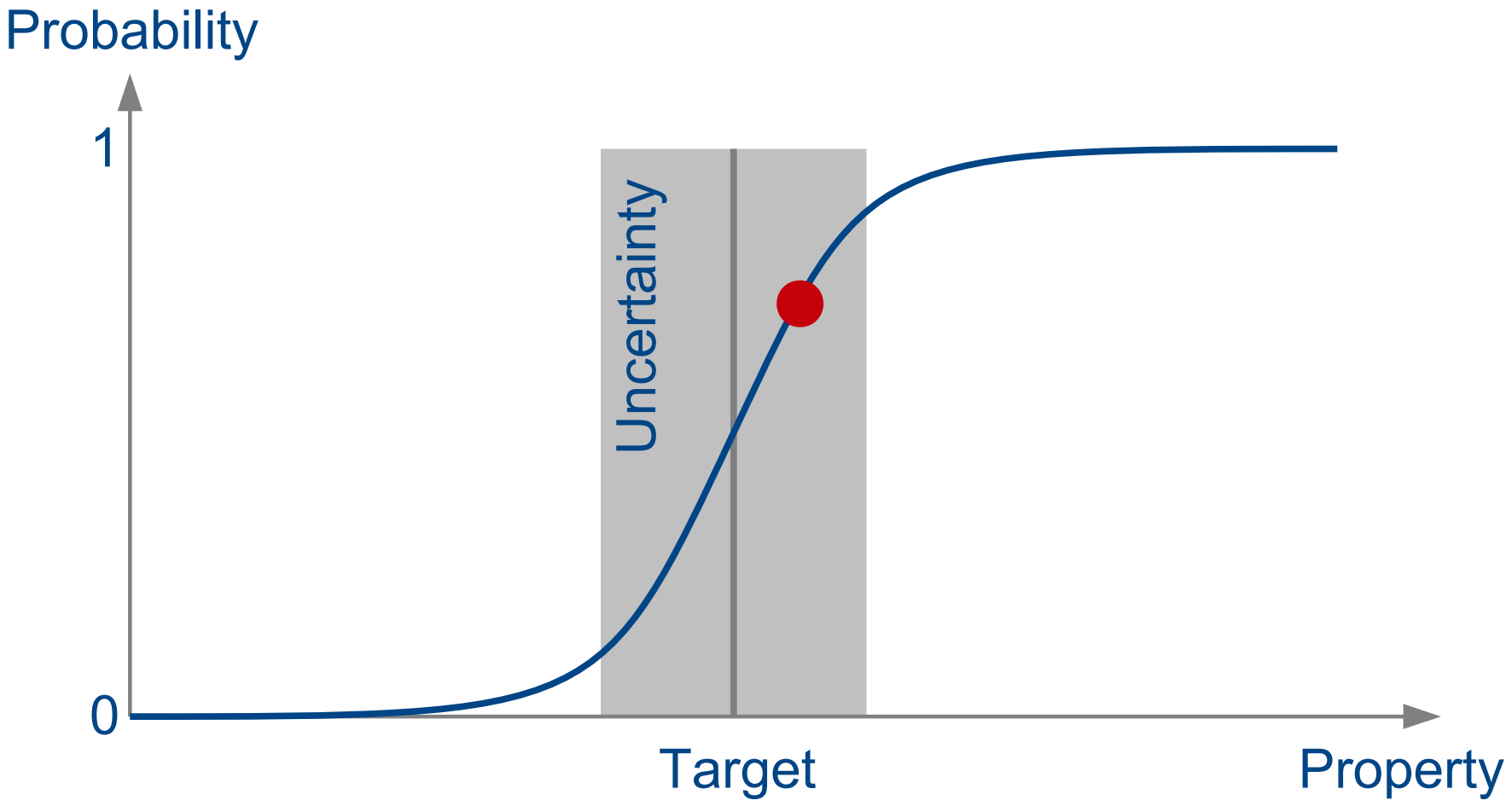
Creep model



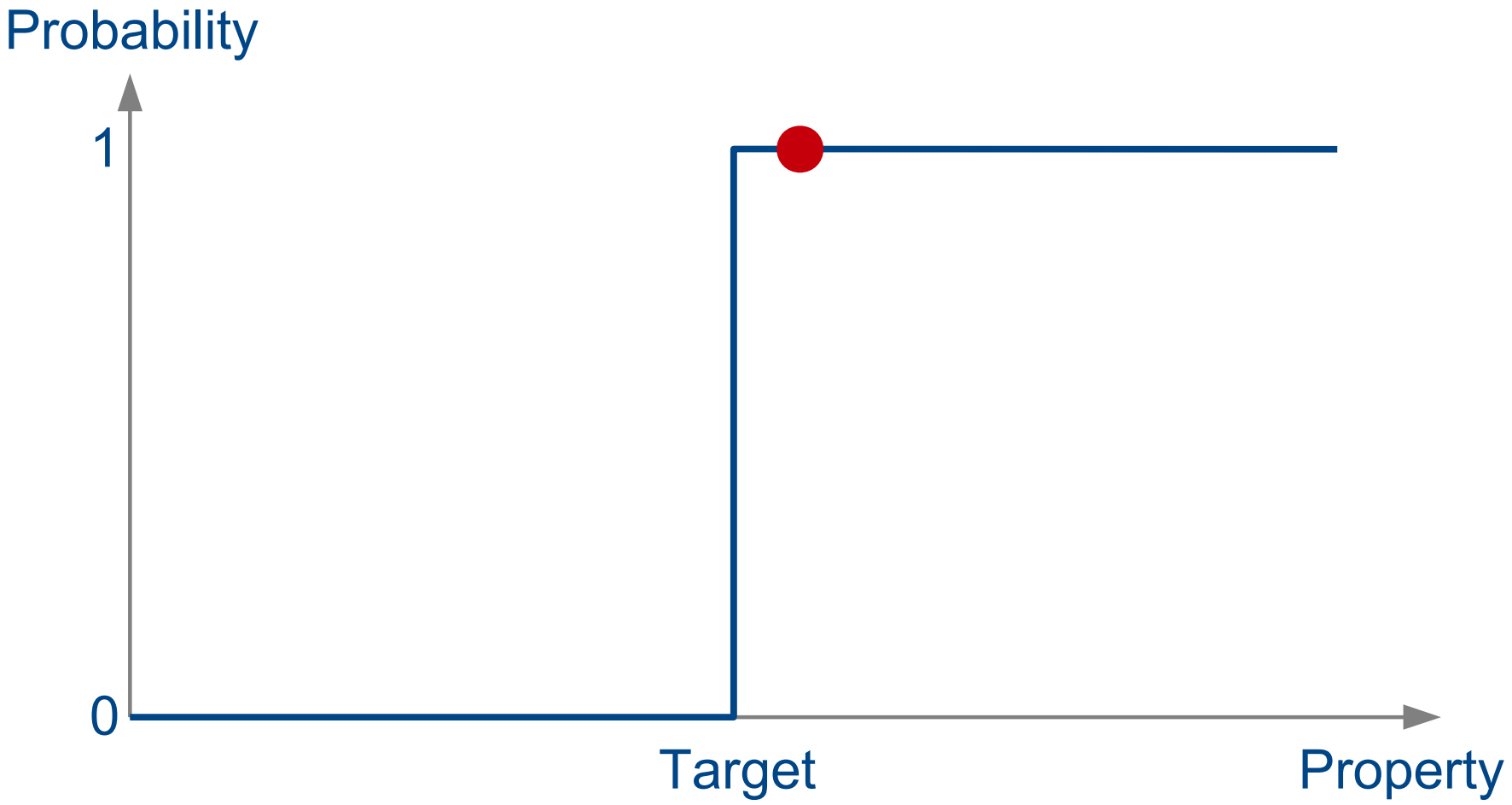
Probability



Probability

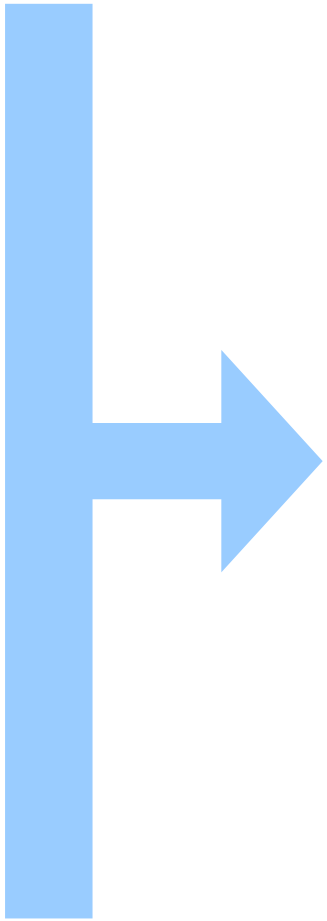


Probability



Probability

Cost $lb^{-1}	$P_{cost}(\mathbf{C})$
γ' fraction	$P_{\gamma'}(\mathbf{C})$
Stability	$P_{stable}(\mathbf{C})$
Density gcm^{-3}	$P_{density}(\mathbf{C})$
Yield stress MPa	$P_{YS}(\mathbf{C})$
UTS MPa	$P_{UTS}(\mathbf{C})$
Oxidation index	$P_{oxidize}(\mathbf{C})$
Stress rupture MPa	$P_{SR}(\mathbf{C})$
Resistivity μΩcm	$P_{resis}(\mathbf{C})$
Entropy Jmol^{-1}K^{-1}	$P_{entropy}(\mathbf{C})$
Low cycle fatigue	$P_{LCF}(\mathbf{C})$
High cycle fatigue	$P_{HCF}(\mathbf{C})$
Weldability	$P_{weld}(\mathbf{C})$
Creep model	$P_{creep}(\mathbf{C})$



$P_{spec}(\mathbf{C})$

Optimization – probability

10 specified properties,
each with probability of 0.5

$0.5^{10} = 0.001$ chance of success

Multidimensional design space

Cr 

Co 

Mo 

W 

Ta 

Nb 

Al 

Ti 

Fe 

Mn 

Si 

C 

B 

Zr 

Cu 

N 

P 

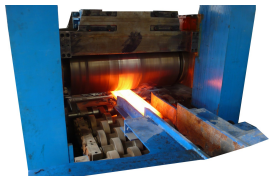
V 

Hf 

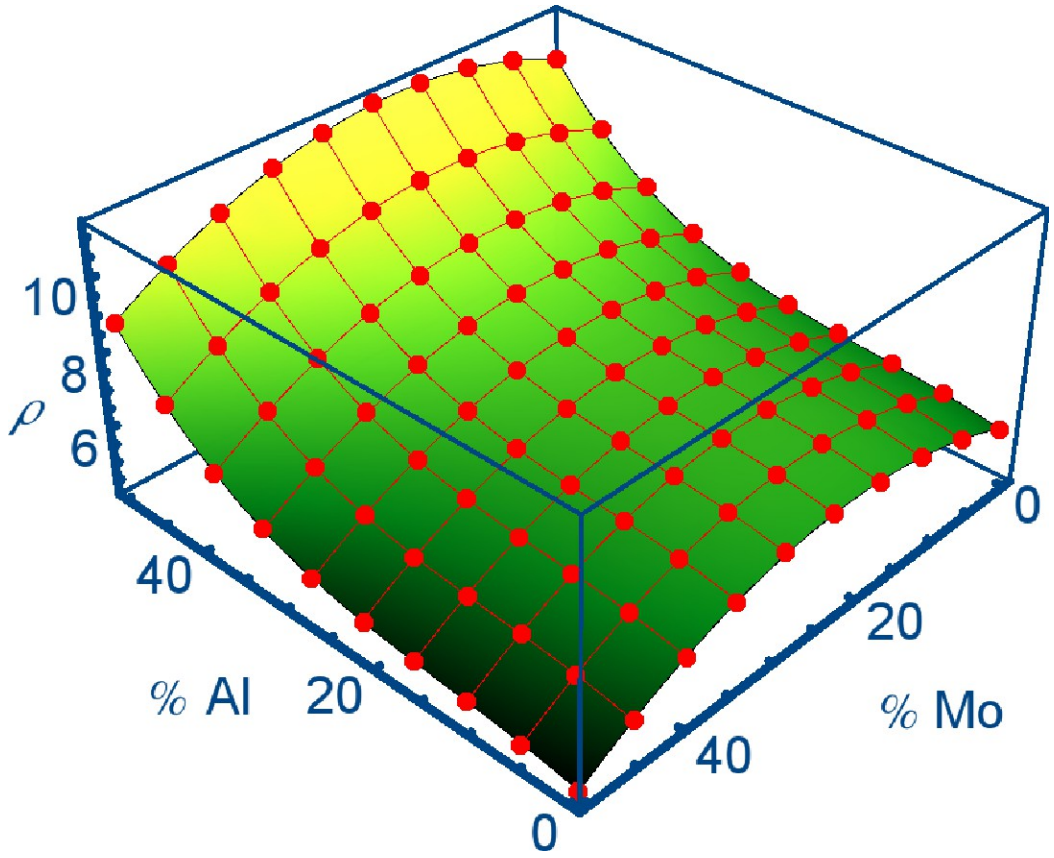
Mg 

Ni 

Heat treatment

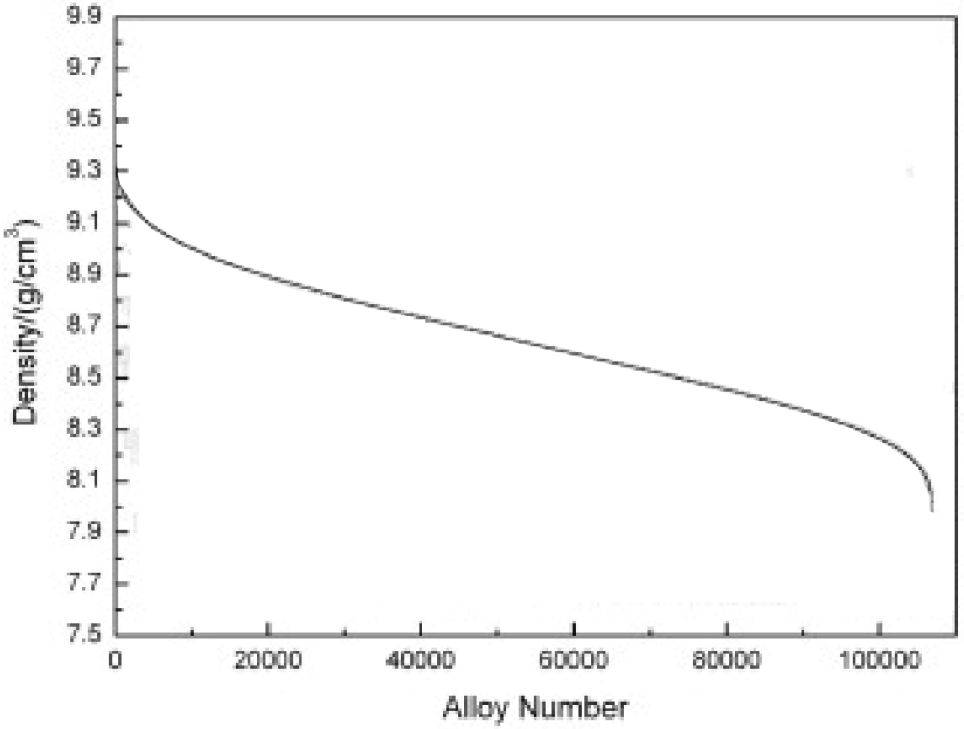


Optimization – tradeoff diagrams



R.C. Reed, T. Tao & N. Warnken, Acta Materialia **57**, 5898 (2009)

Optimization – tradeoff diagrams



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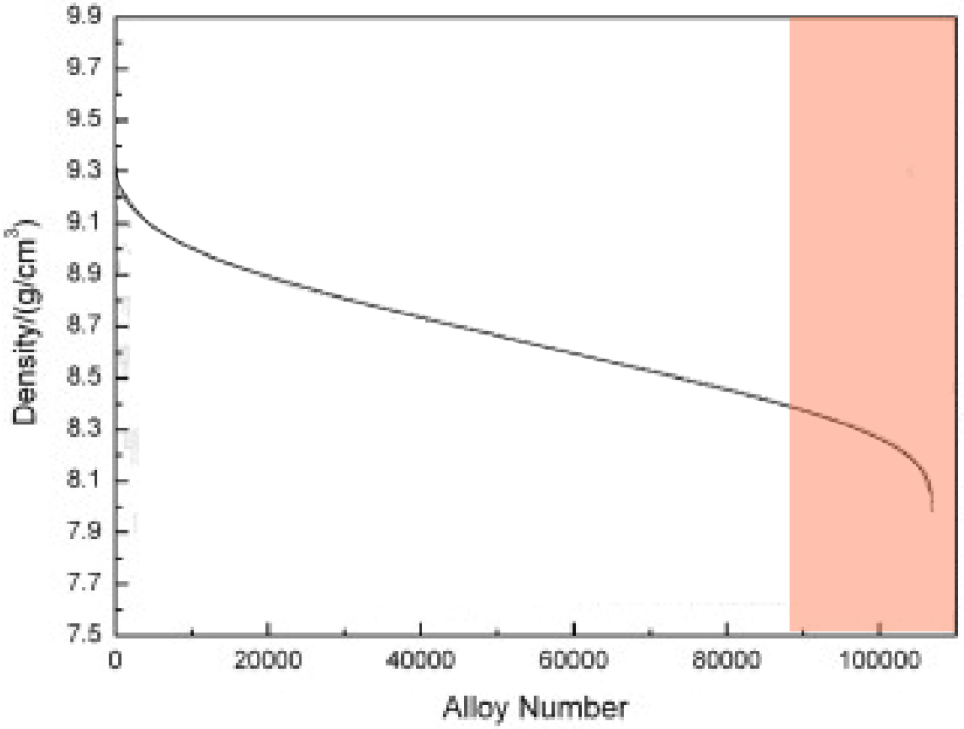
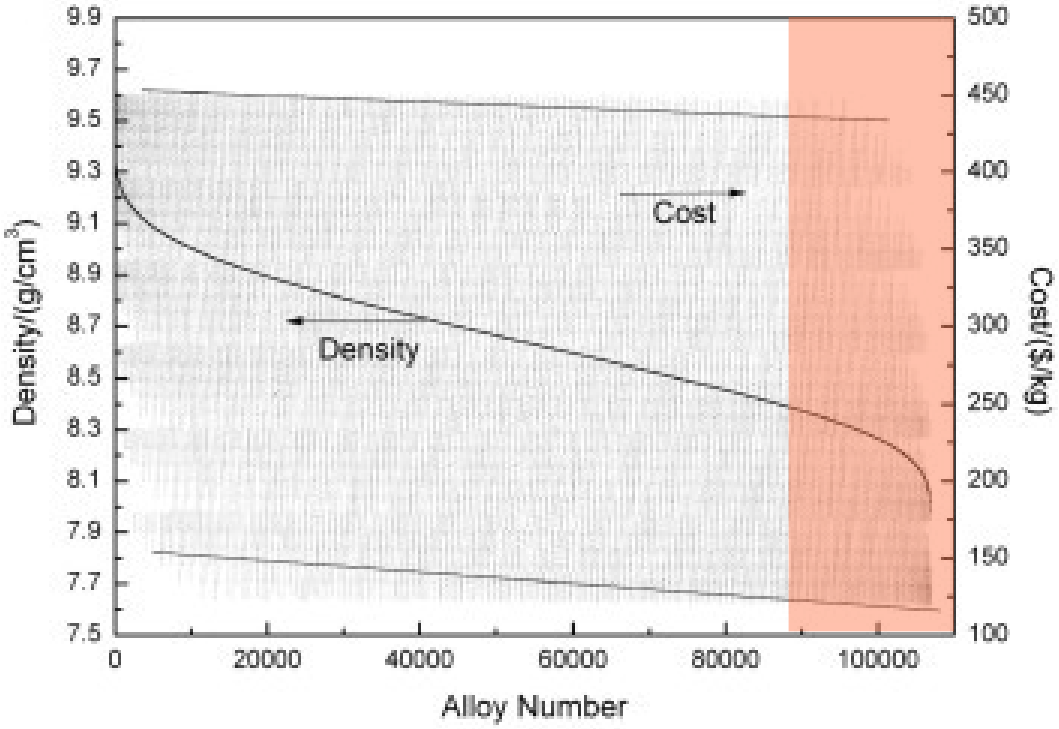
Optimization – tradeoff diagrams



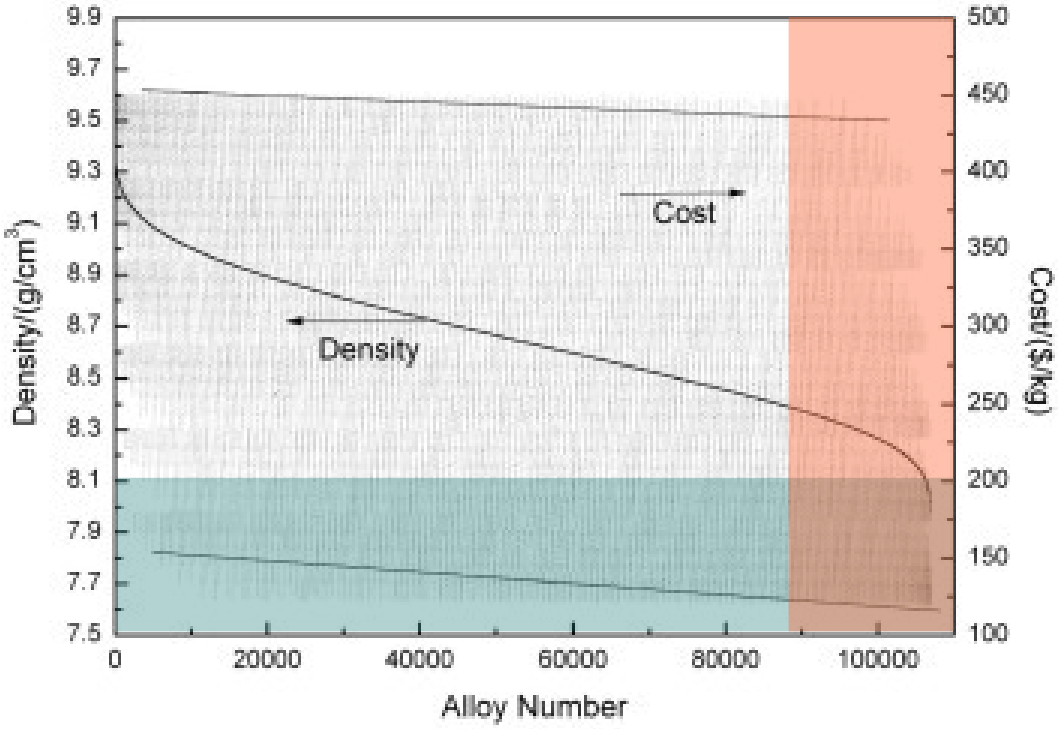
R.C. Reed, T. Tao & N. Warnken, Acta Materialia **57**, 5898 (2009)

Optimization – tradeoff diagrams



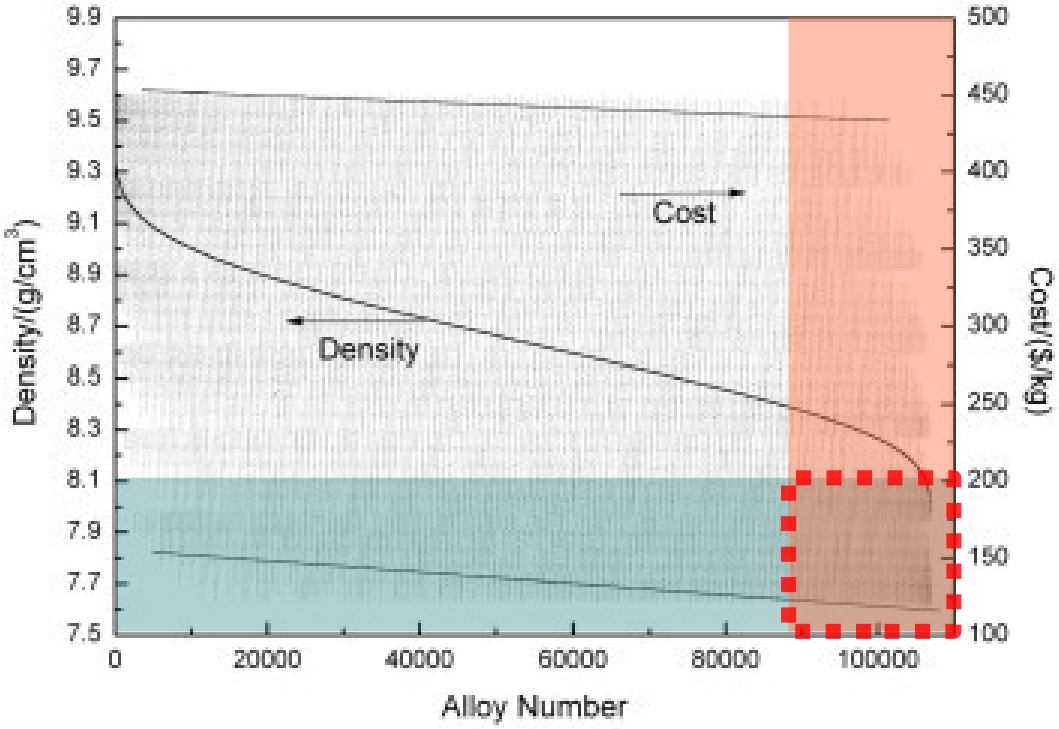
R.C. Reed, T. Tao & N. Warnken, Acta Materialia **57**, 5898 (2009)

Optimization – tradeoff diagrams



R.C. Reed, T. Tao & N. Warnken, Acta Materialia **57**, 5898 (2009)

Optimization – tradeoff diagrams



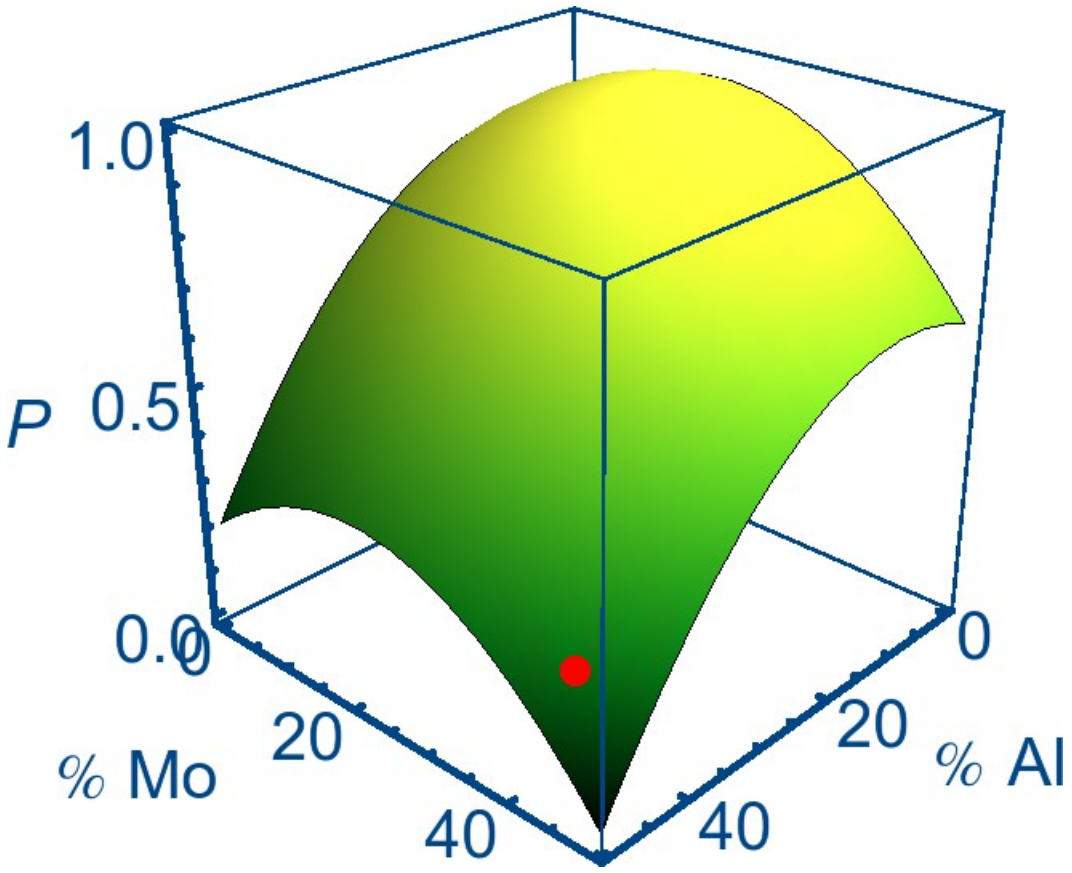
R.C. Reed, T. Tao & N. Warnken, Acta Materialia **57**, 5898 (2009)

Optimization – tradeoff diagrams

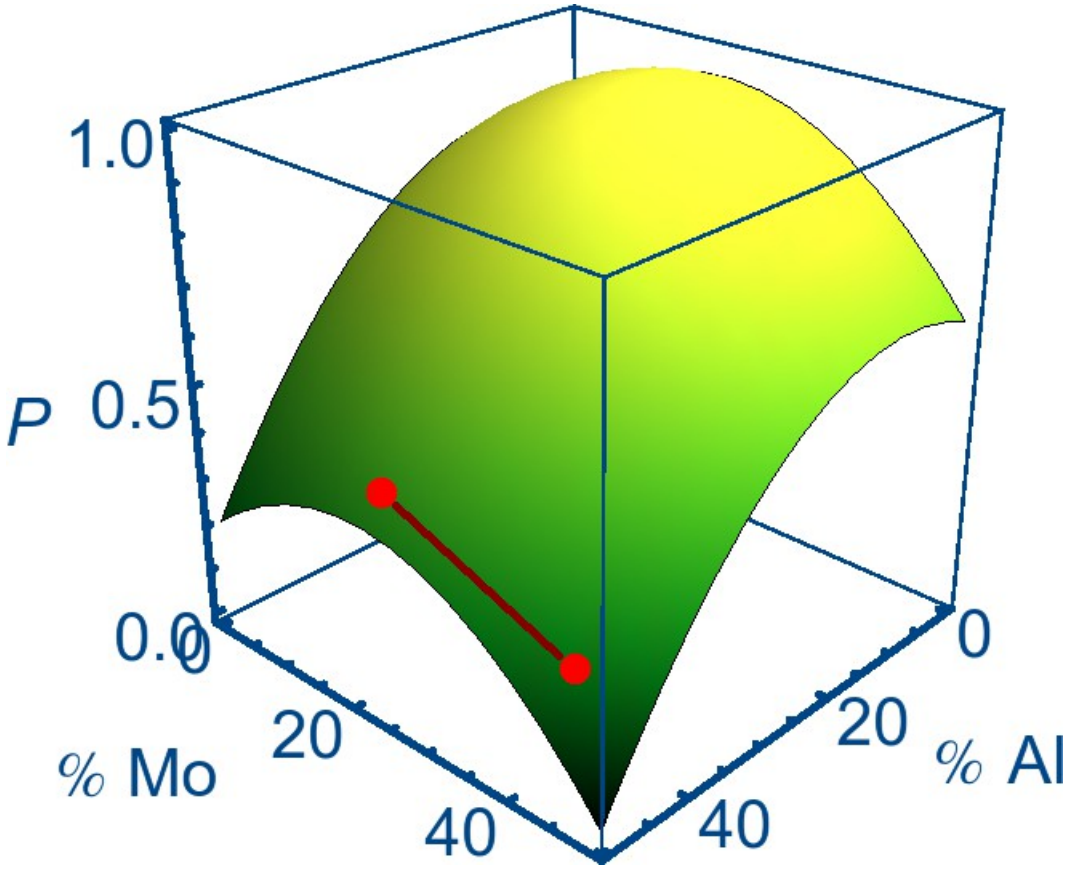
30 design variables, at 0.1% accuracy, and evaluation time of 1ms

$1000^{30} = 3 \times 10^{79}$ years

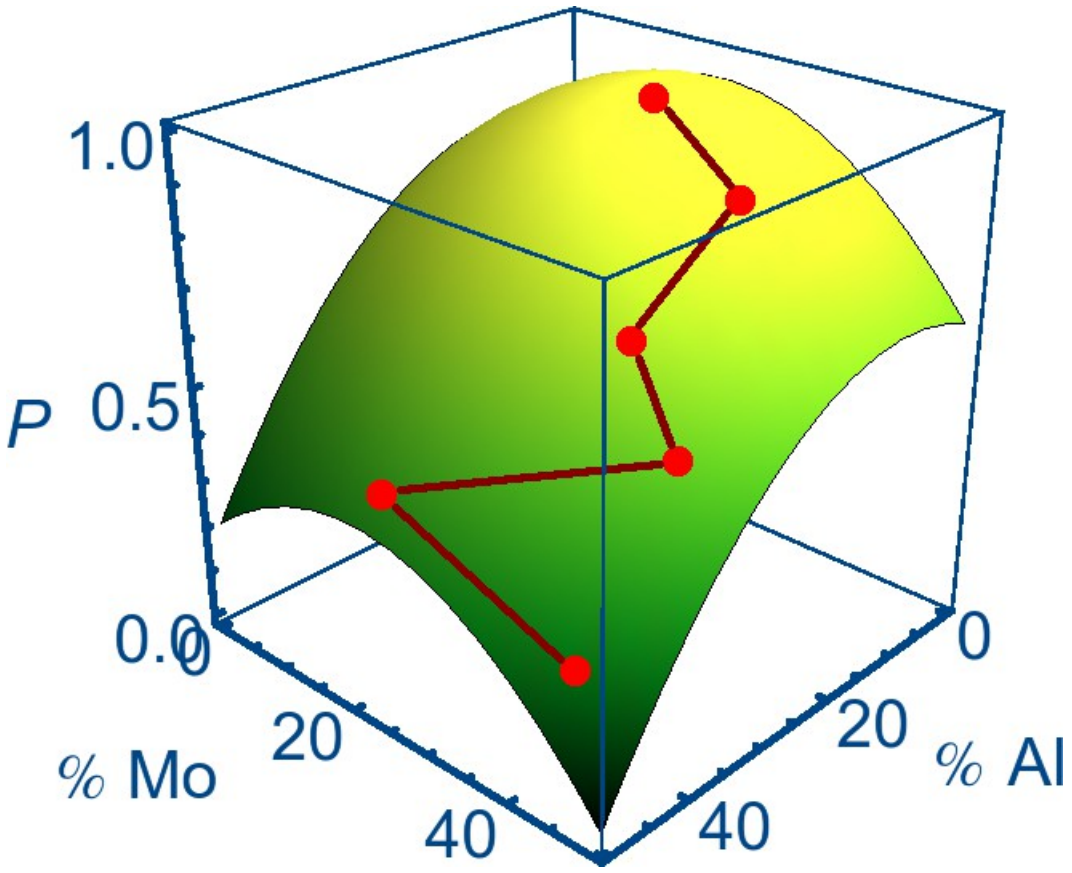
Optimization – replica exchange sampling



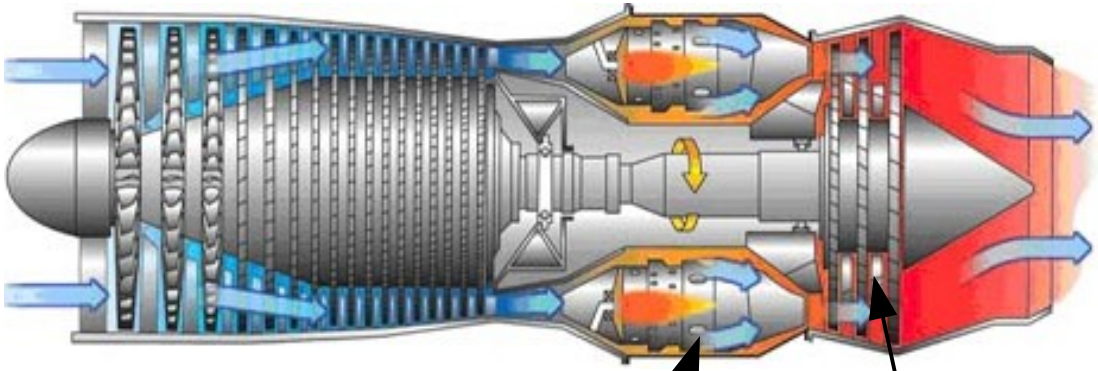
Optimization – replica exchange sampling



Optimization – replica exchange sampling



Predicted alloys

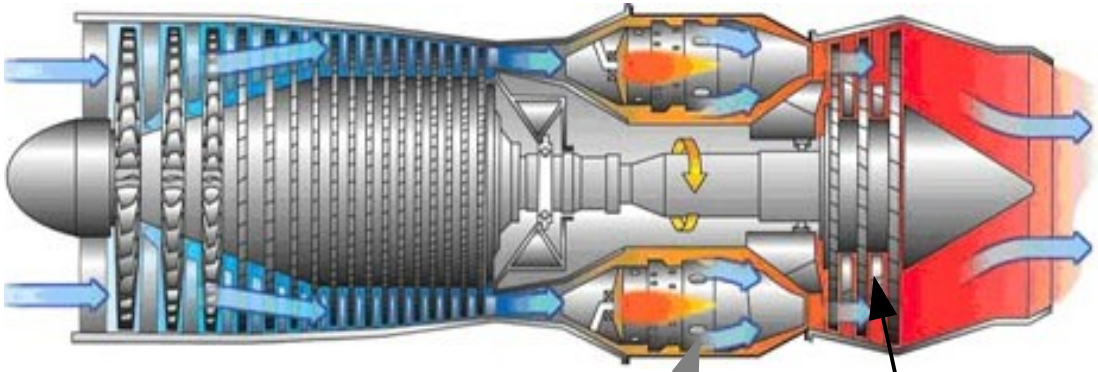


Combustor liner

2x disc alloy

2x forging hammer

Predicted alloys



Combustor liner

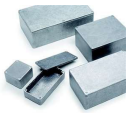
2x disc alloy

2x forging hammer

Case study: improved disc alloy








Ni	Cr	Co	Mo	Ti	Al	Ta	Hf	C	T	t
52	15	19	5	3.6	3	2	0.5	0.1	800	8

Case study: improved disc alloy

Ni

56

Cr

17

Co

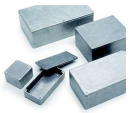
1.0

Mo

4.0

Ti

1.5



Al

4.3

Ta

0.2

Hf

0.1

C

0.2

T

980

t

61

W

6.0

Mn

0.1

B

0.1

V

0.1

Si

0.1

Zr

0.2

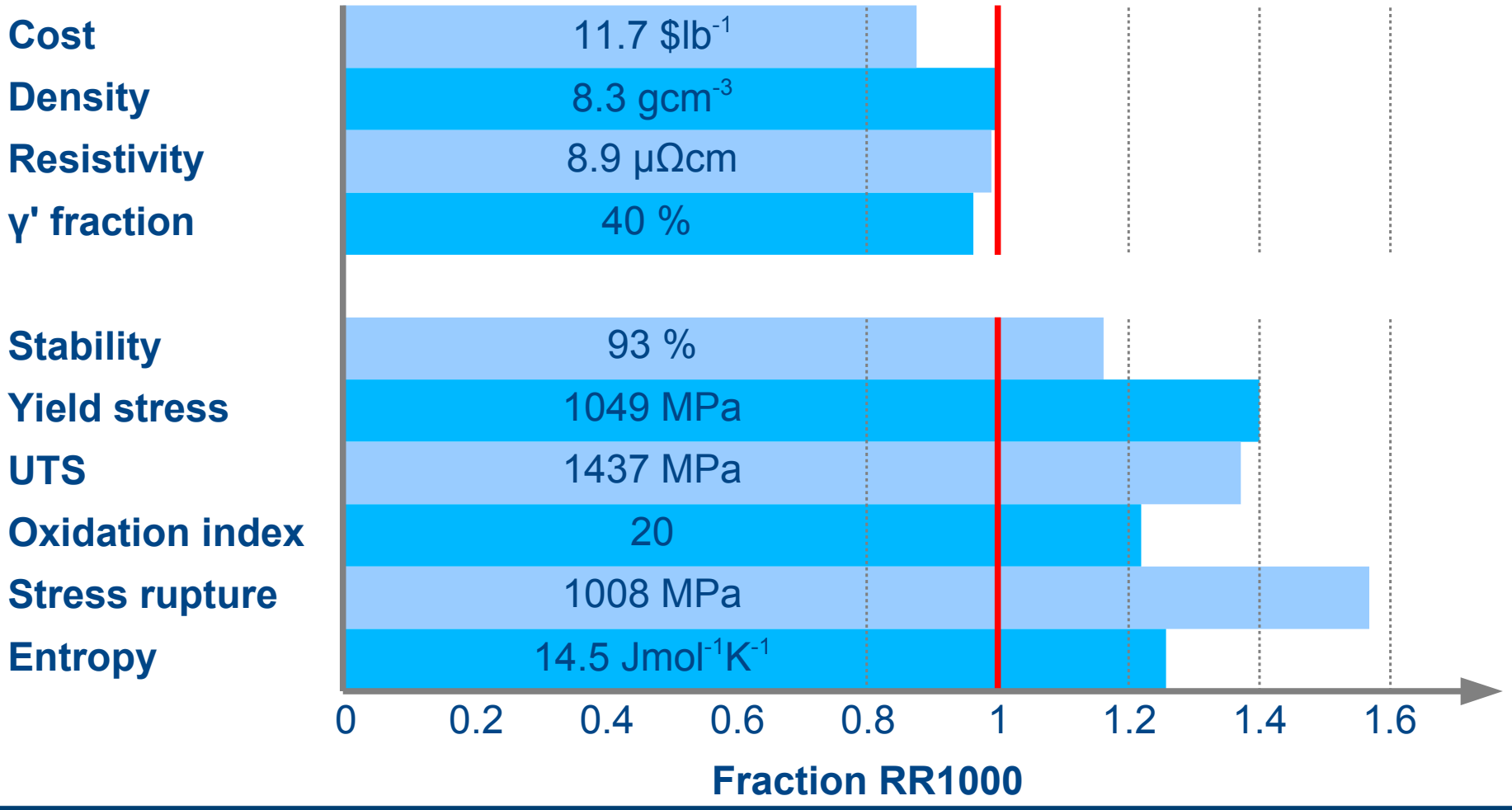
Nb

5.6

Fe

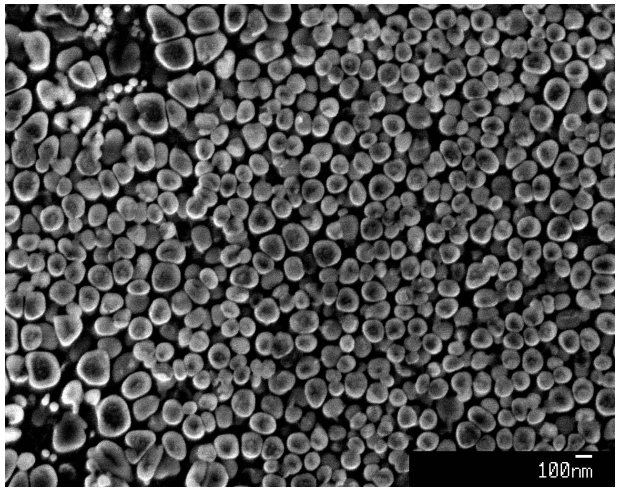
3.4

Case study: improved disc alloy



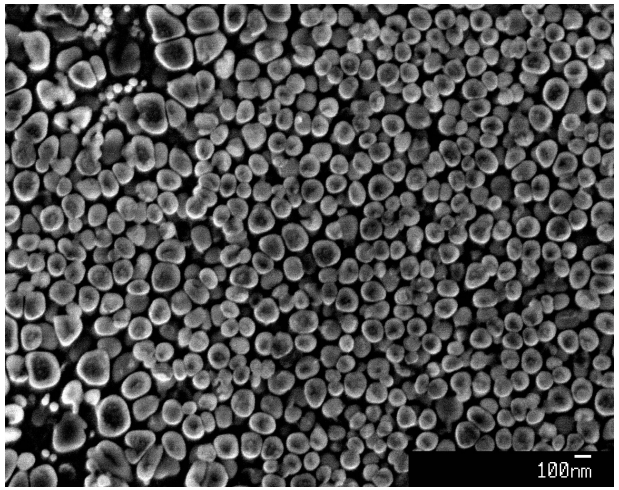
Electron micrograph – Ni disc alloy

Ni disc alloy

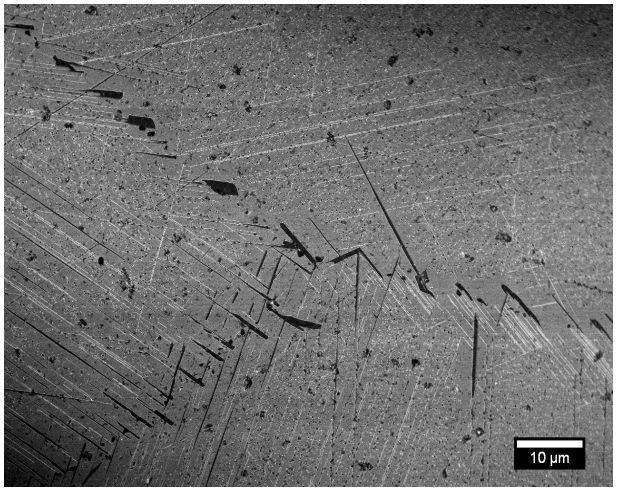


Electron micrograph – Ni disc alloy

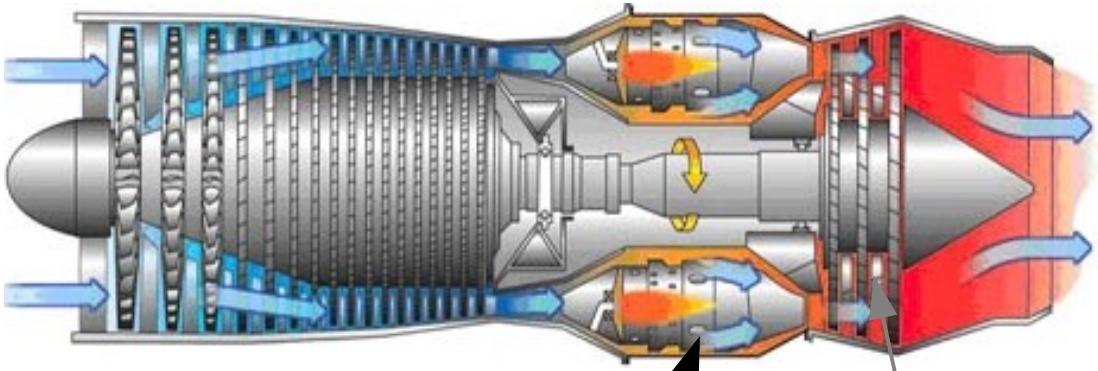
Ni disc alloy



η contaminated alloy



Predicted alloys

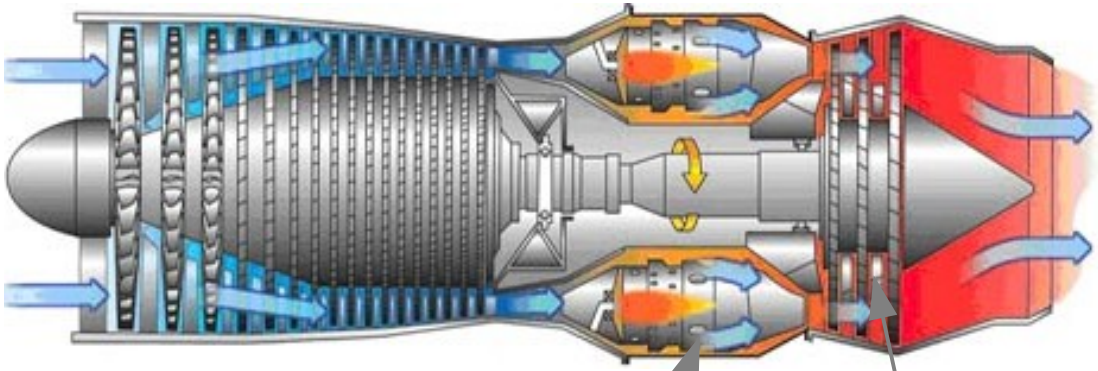


Combustor liner

2x disc alloy

2x forging hammer

Predicted alloys



Combustor liner

2x disc alloy

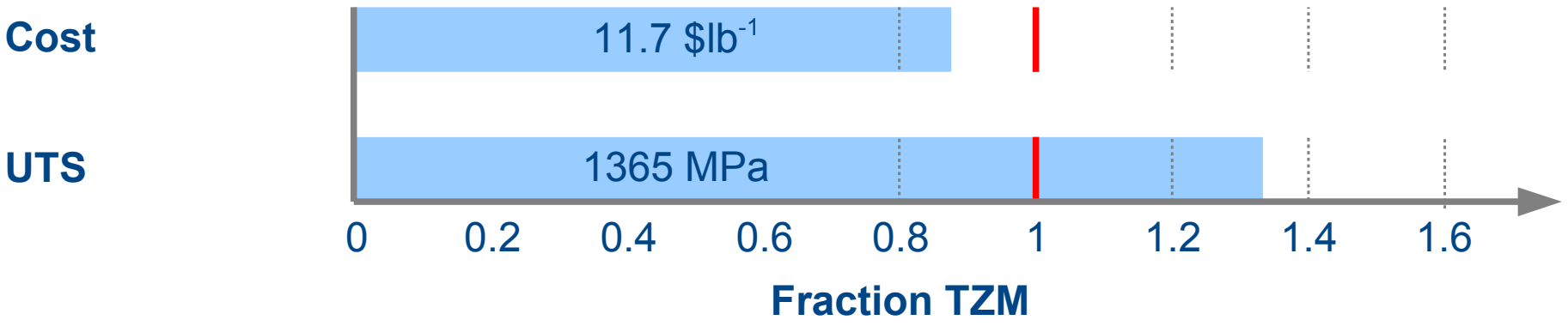
2x forging hammer

Case study: improved forging alloy




	Mo	Ti	C	Zr
TZM	99.4	0.5	0.02	0.08

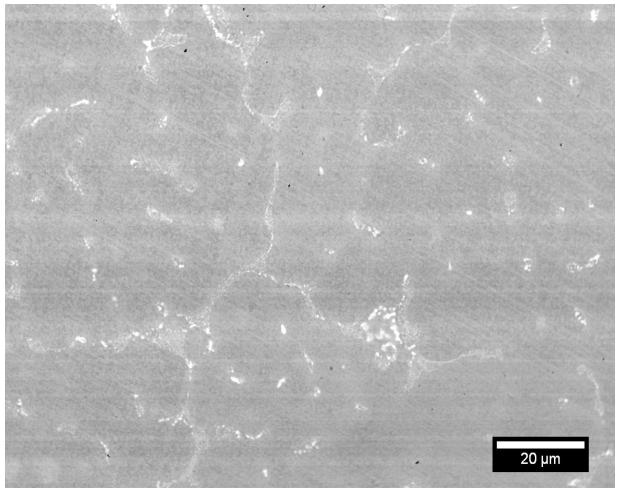
Case study: improved forging alloy



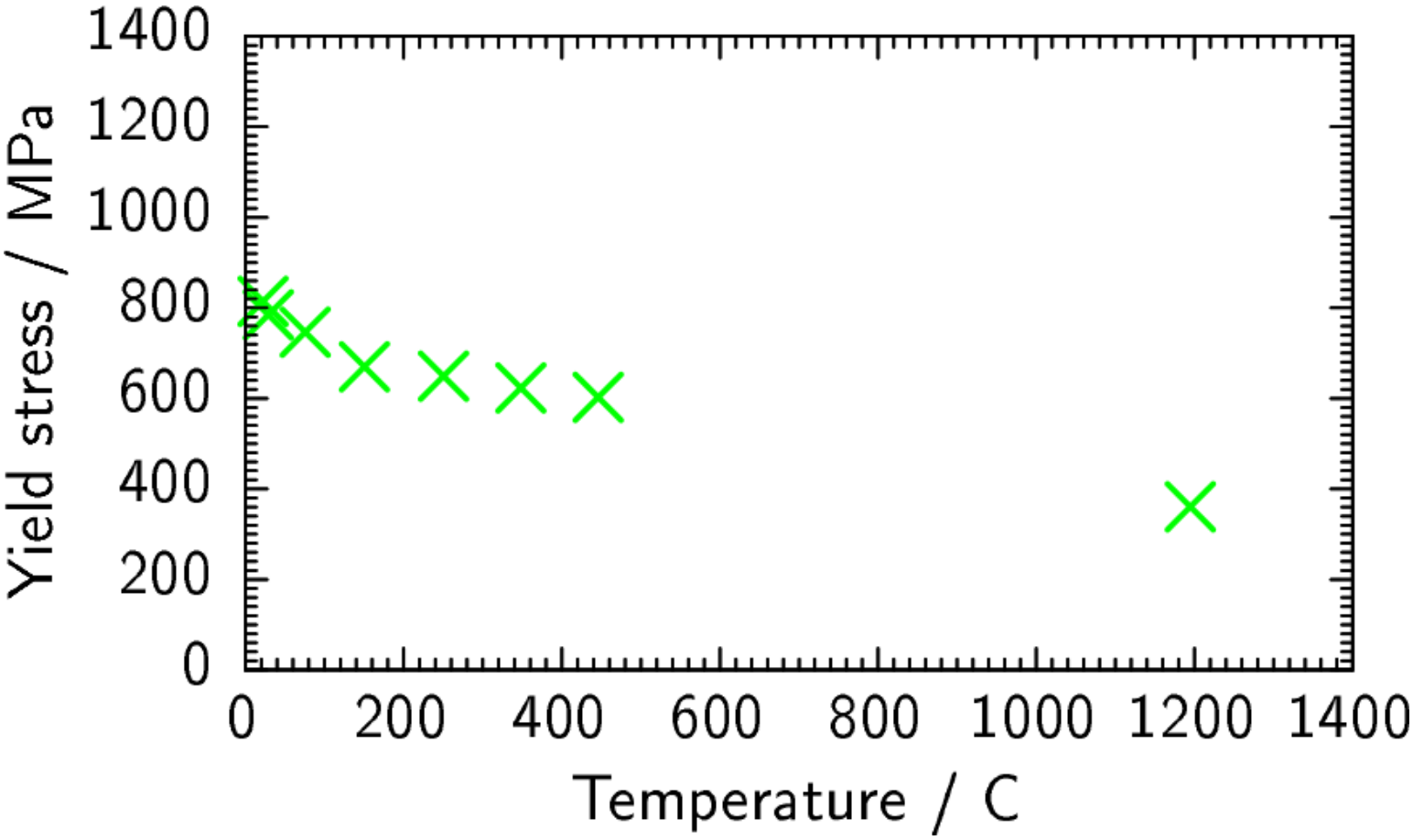
	Mo	Ti	C	Zr	Hf	W	Nb
TZM	99.4	0.5	0.02	0.08			
Optimal	82.7	1.0	0.2	0.9	9.0	0.5	5.7

Electron micrograph – Mo forging alloy

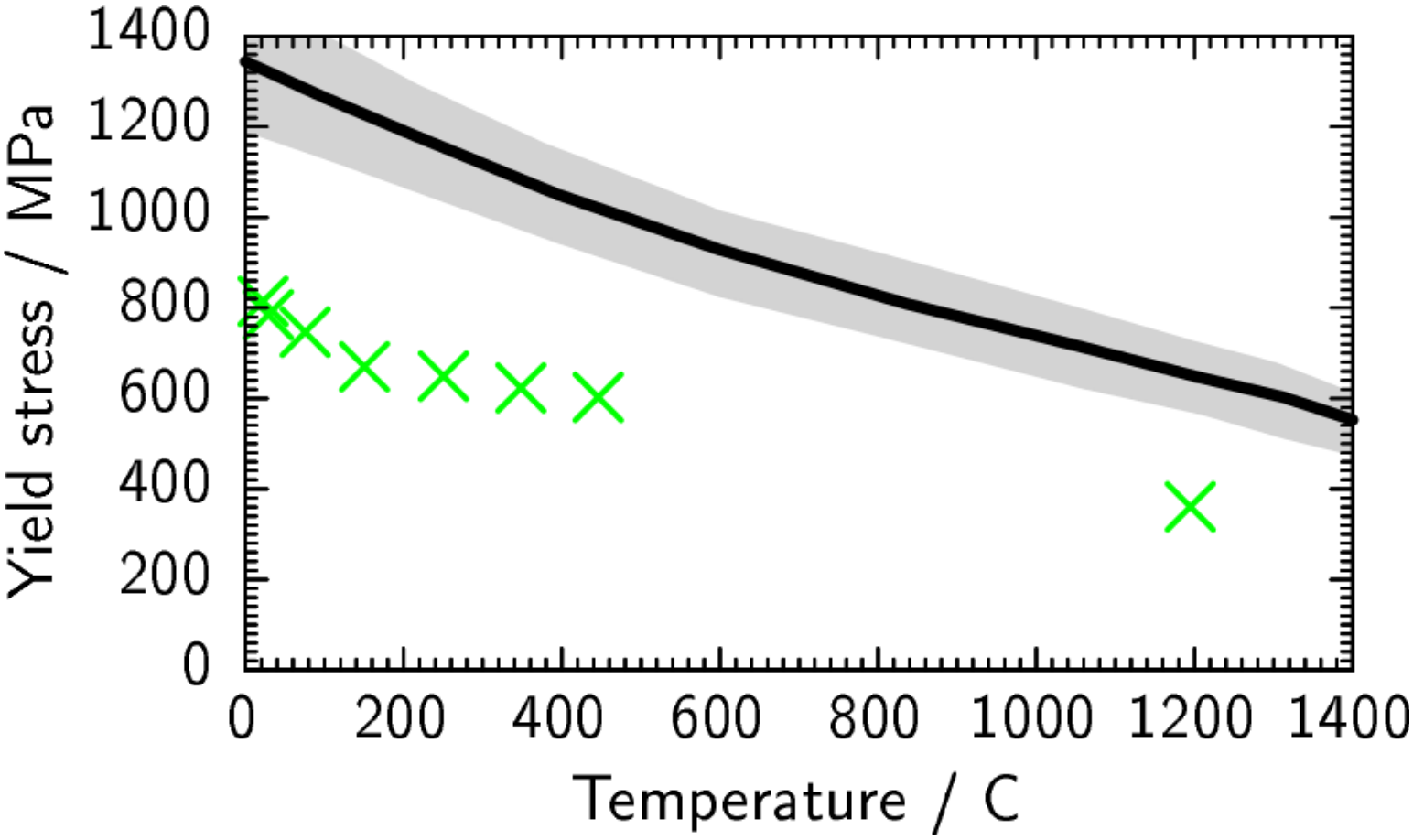
Mo forging alloy



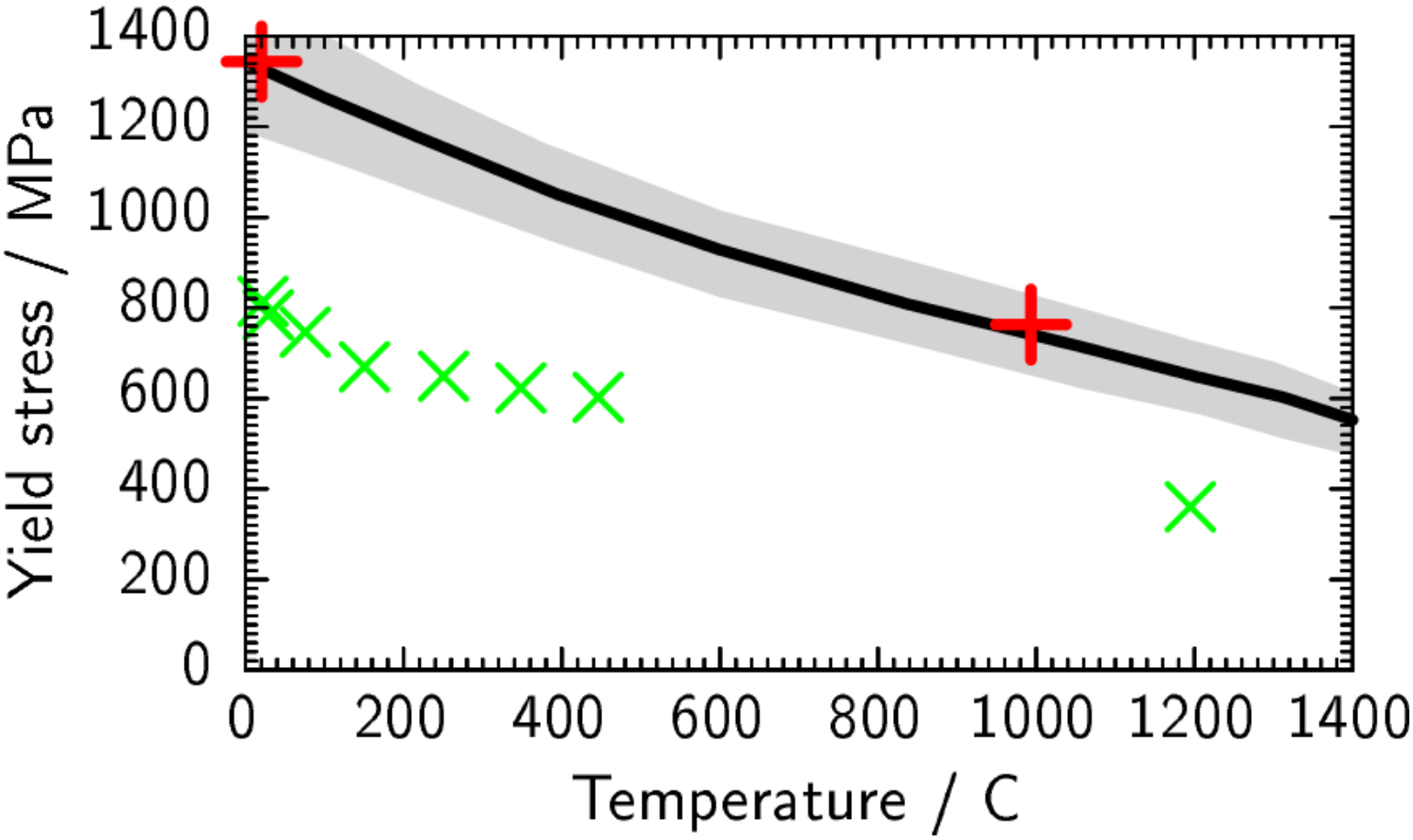
Yield stress



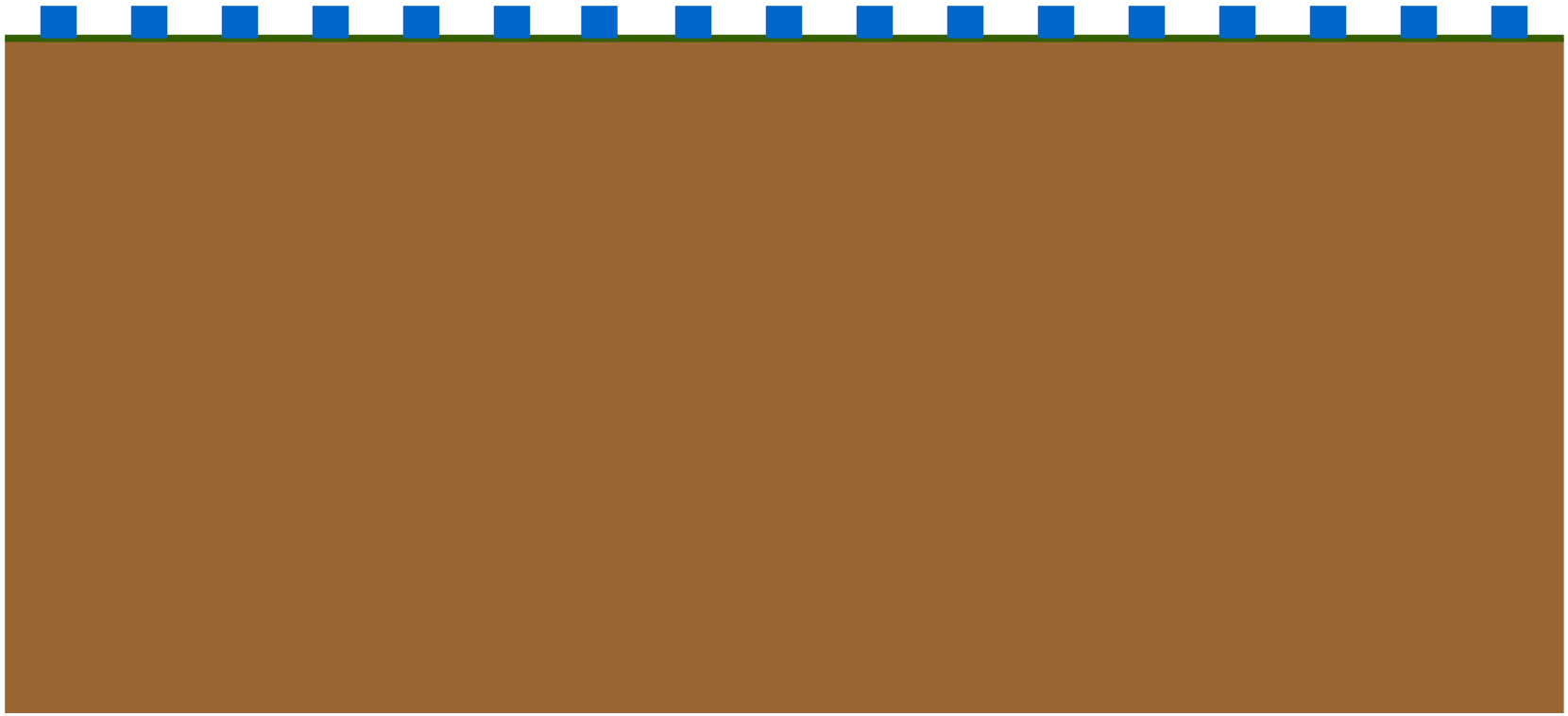
Yield stress



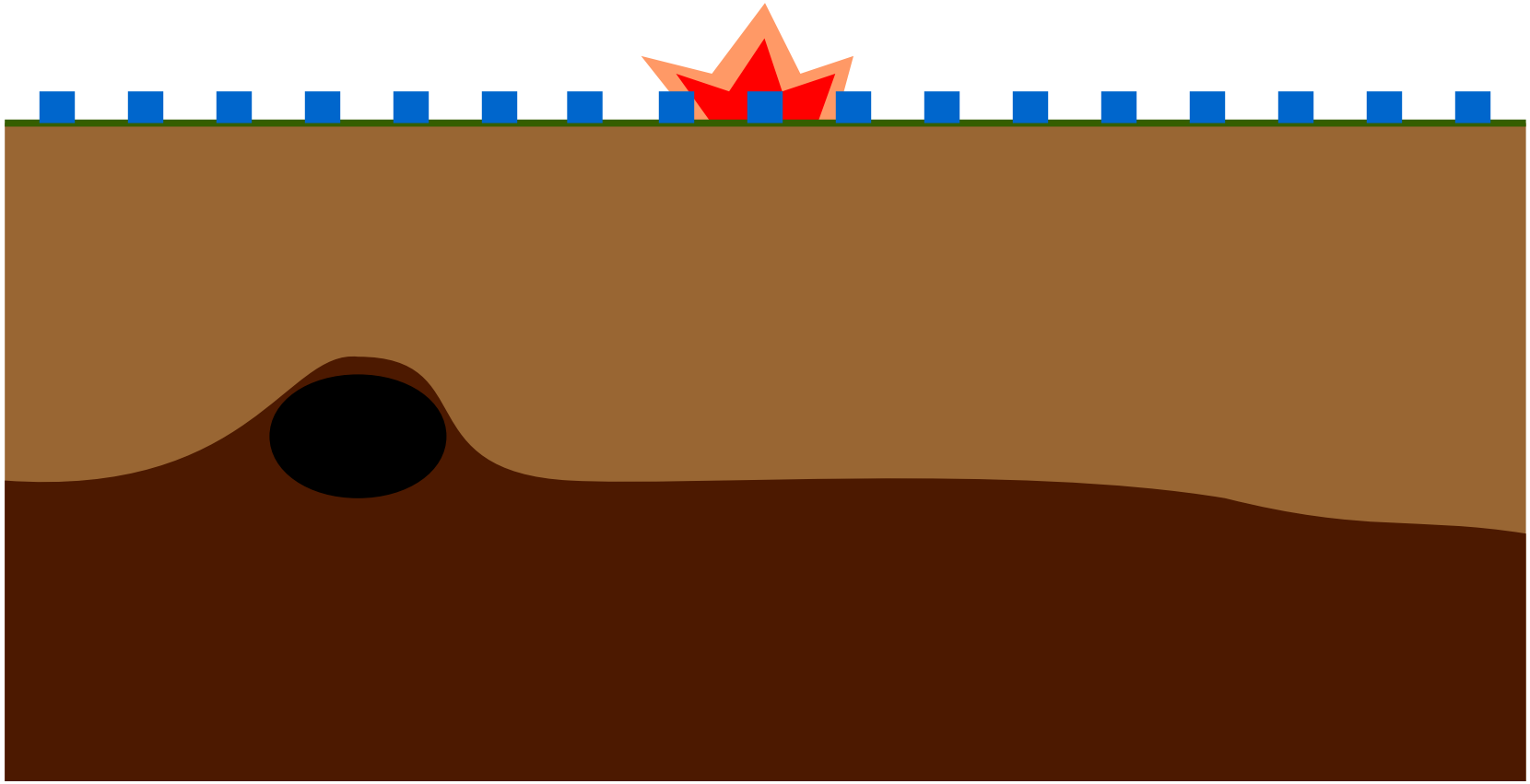
Yield stress



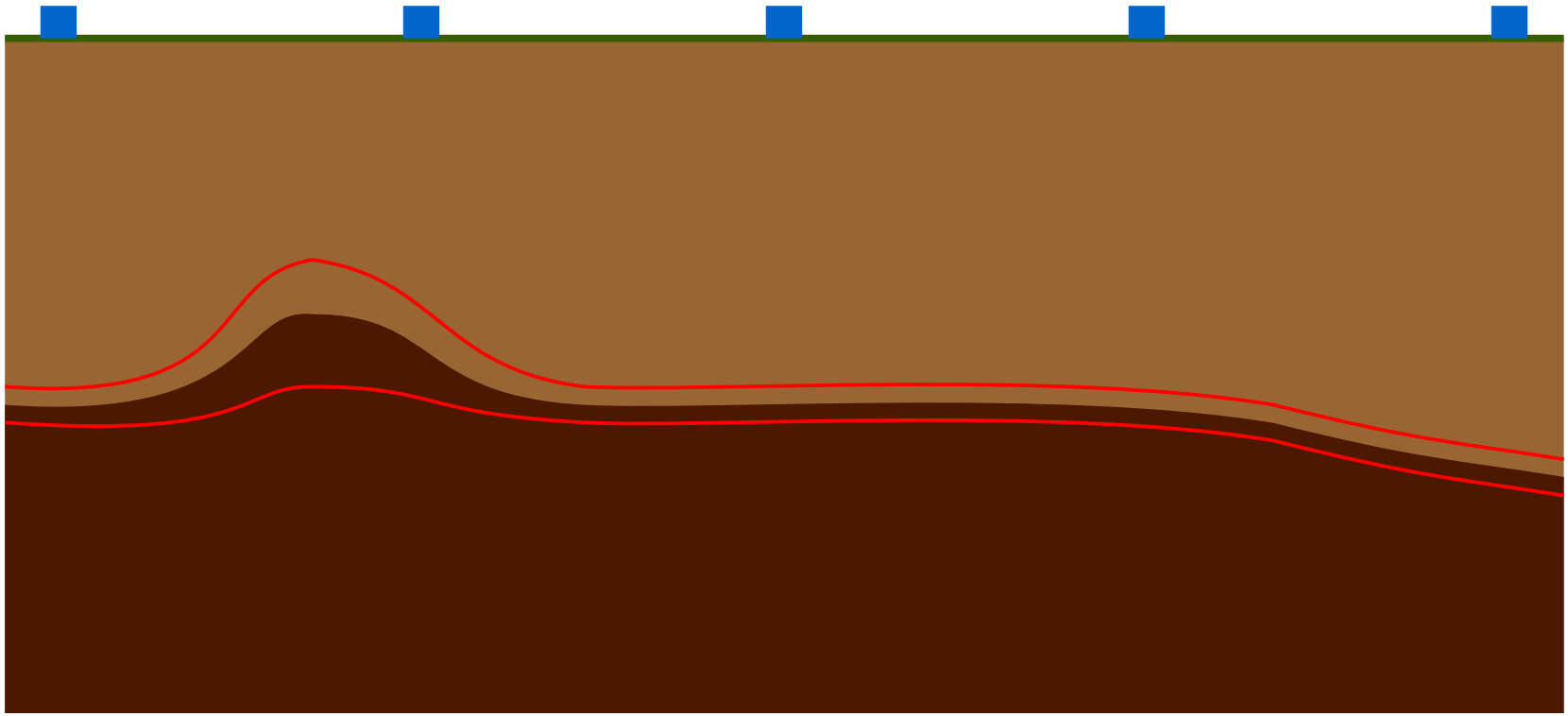
Seismic surveys



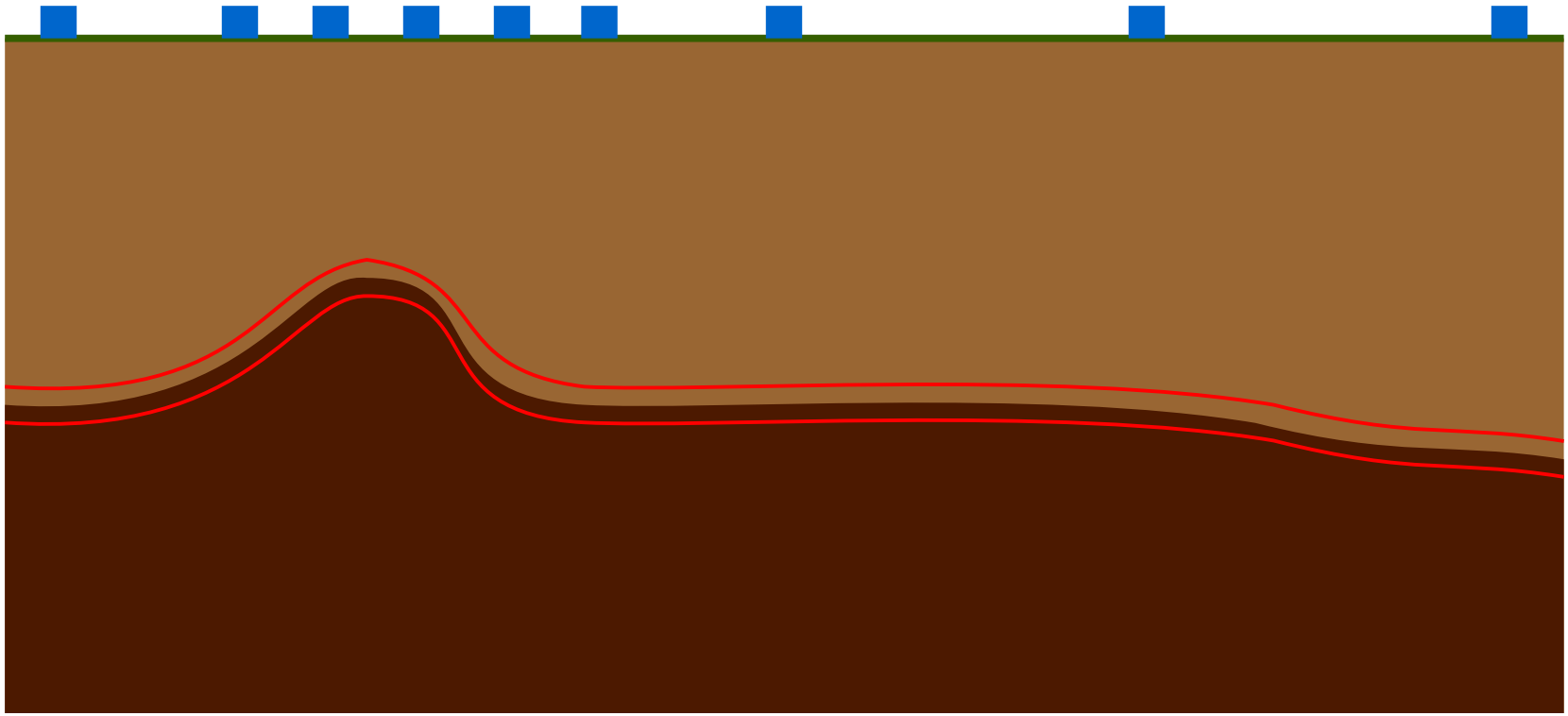
Seismic surveys



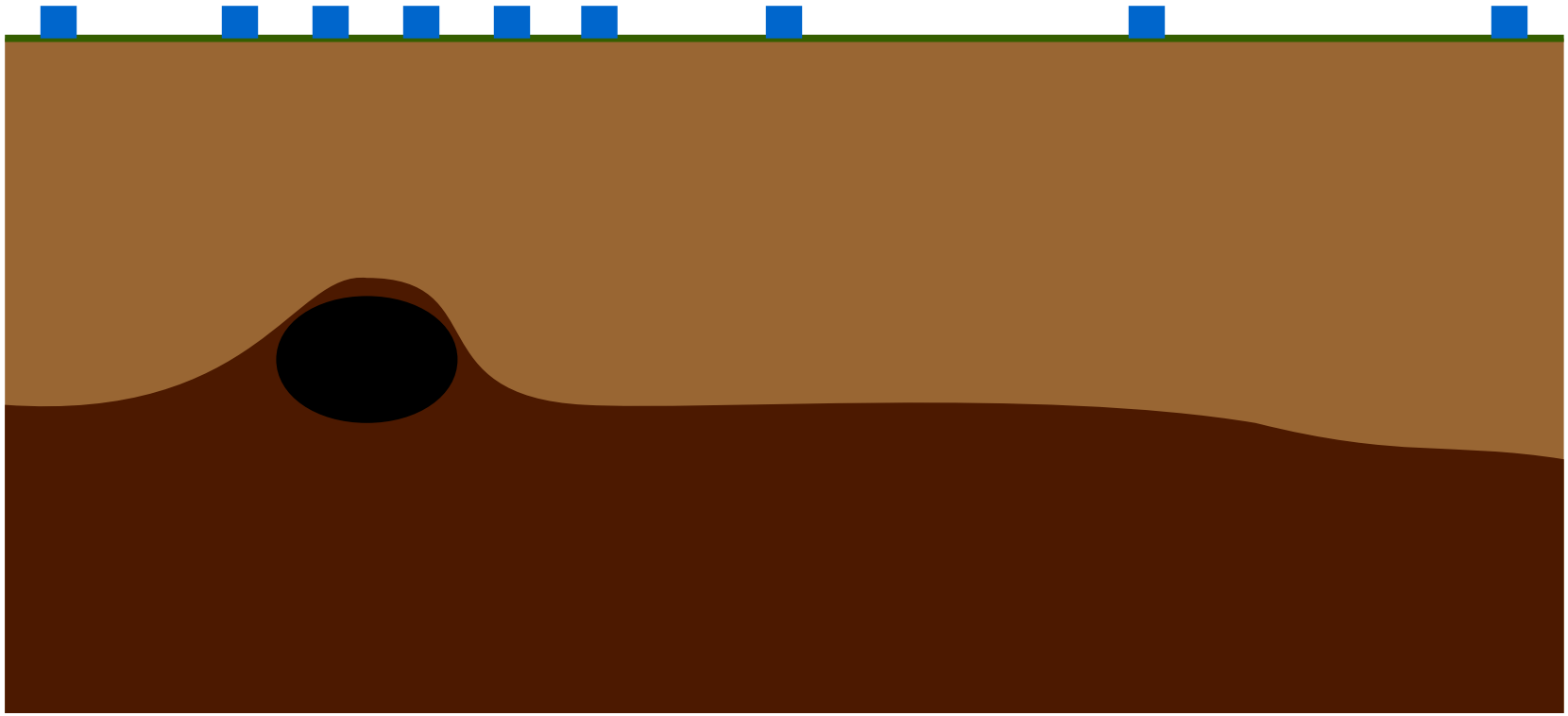
Seismic surveys



Seismic surveys

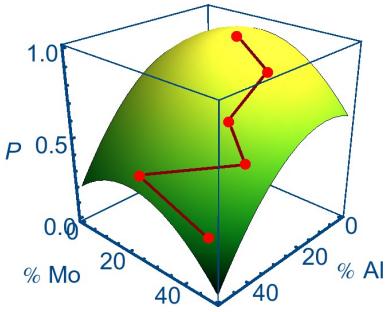


Seismic surveys

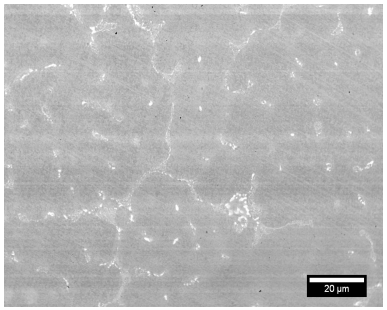


Concurrent materials design

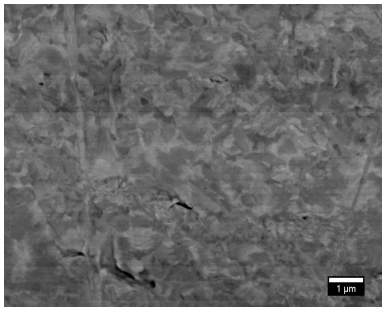
Discovery algorithm
Patent GB1302743.8 (2013)



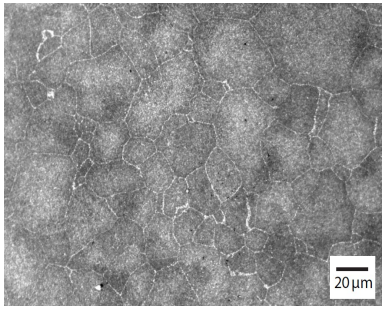
Mo-Hf forging alloy
Patent GB1307533.8 (2013)



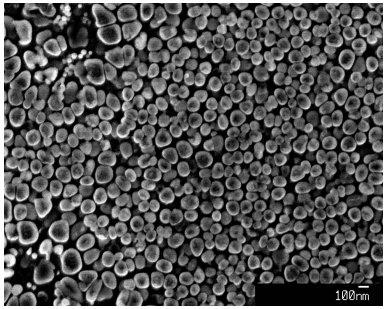
Mo-Nb forging alloy
Patent GB1307535.3 (2013)



RR1000 grain growth
Acta Materialia, **61**, 3378 (2013)



Ni disc alloy
Rolls-Royce invention
NC12261 (2012)



Ni combustor liner
Rolls-Royce invention
NC13006 (2013)