

# DYALOG

Elsinore 2023

## Grain Growth and Array Programming

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*Based on a true story*

# Problem statement

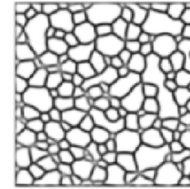
- ◆ Materials science students
- ◆ Choose modelling problem
- ◆ Scientific literature
- ◆ Reproduce model



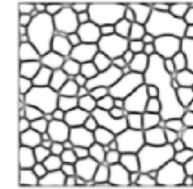
# Grain Growth

- Increment of size of grains in a polycrystal subjected to high temperatures

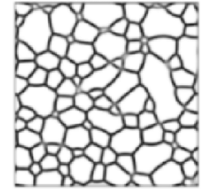
$t=2 \Delta t$



$t=8 \Delta t$

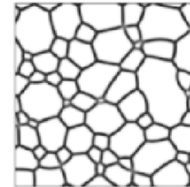


$t=20 \Delta t$

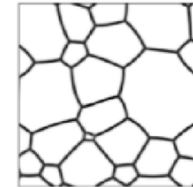


]

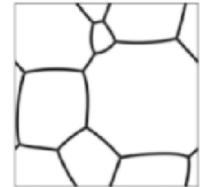
$t=50 \Delta t$



$t=250 \Delta t$



$t=800 \Delta t$



Dimokrati, A., & Benyoucef, M. (2016, September)  
Phase Field Simulation of Normal Grain Growth  
CONAT 2016



# Grain Growth (Geiger, 2001)

https://doi.org/10.1016/S1359-6454(00)00352-9

### 3.1. SIMULATION PROCEDURE

In CA, instead of each cell of the lattice representing a group of atoms, and the evolution of the microstructure is governed by the behavior of individual cells acting in response to their neighborhood. The motion equation of lattice sites is determined by the evolution of the simulation. It is typically in the form of  $\dot{x}_i = f(x_i)$ , depending on the current state of the simulation.

Our purpose in the present work is to develop a computer simulation method based on the cellular automata approach. Such will model several grain nucleating and competing, will be the physical model used in the simulation. This CA model is based on the following:

• The transition (jumping) of atoms (lattice) depends on the energy barrier and the energy difference of atoms before and after jumping.

• The energy barrier for the transition is defined as the energy of the grain boundary adjacent to the nucleation of the grain.

• The energy of the grain boundary depends on the nucleation of the grain.

The model developed allows the grain coarsening in two dimensions, as the process is generally measured and recorded on the surface of the sample and the size of the lattice. The CA model is in 2D.

Equation (1) is the grain energy stability of the transition of a grain  $i$  through the grain boundary:

$$\dot{x}_i = f(x_i) = \frac{1}{kT} \exp\left(-\frac{E_i}{kT}\right) \exp\left(-\frac{E_{ij}}{kT}\right) \quad (1)$$

where  $E_i$  is the energy of the grain boundary adjacent to the nucleation of the grain, and  $E_{ij}$  is the energy difference of the grain  $i$  and the grain  $j$ .

Equation (2) is the grain energy stability of the transition of a grain  $i$  through the grain boundary:

$$\dot{x}_i = f(x_i) = \frac{1}{kT} \exp\left(-\frac{E_i}{kT}\right) \exp\left(-\frac{E_{ij}}{kT}\right) \quad (2)$$

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where  $E_i$  is the energy of the grain boundary adjacent to the nucleation of the grain, and  $E_{ij}$  is the energy difference of the grain  $i$  and the grain  $j$ .

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# Grain Growth (Geiger, 2001)

[https://doi.org/10.1016/S1359-6454\(00\)00352-9](https://doi.org/10.1016/S1359-6454(00)00352-9)

- Thermal energy
- Boundary energy
- Activation energy

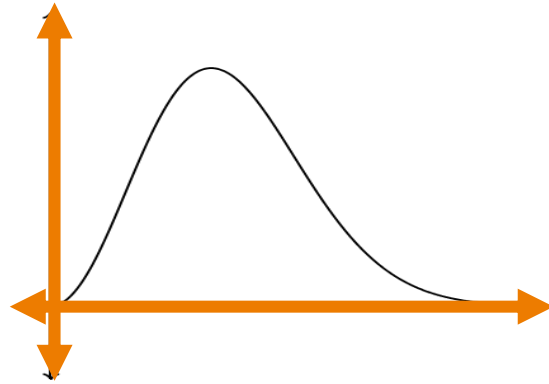


# Grain Growth (Geiger, 2001)

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- Thermal energy
- Boundary energy
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Maxwell-Boltzmann  
distribution



$$G_T(T) = -R T \log(x)$$

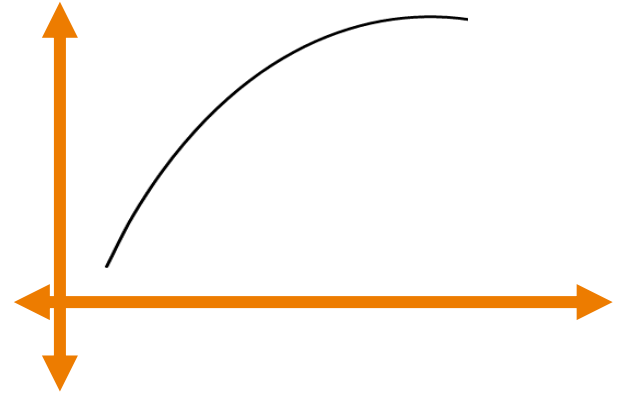
$$x \in U(0,1)$$



# Grain Growth (Geiger, 2001)

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- Thermal energy
- Boundary energy
- Activation energy



Read-Shockley  
equation

$$G_{Bij}(\Delta\theta_{ij}) = G_0 \sin(\Delta\theta_{ij}) (1 - \log(\sin(\Delta\theta_{ij})))$$

Misorientation

$$\Delta\theta_{ij} = \frac{\pi}{2} \frac{(q_i - q_j)}{q_{max}}$$





# Grain Growth (Geiger, 2001)

[https://doi.org/10.1016/S1359-6454\(00\)00352-9](https://doi.org/10.1016/S1359-6454(00)00352-9)

- Thermal energy
- Boundary energy
- Activation energy

$$G_A = 10000 \text{ J/mol}$$

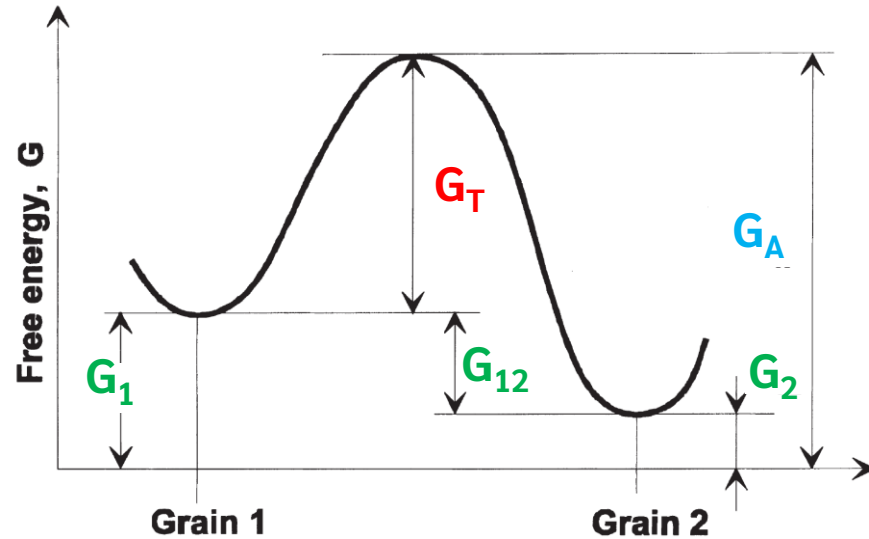


# Grain Growth (Geiger, 2001)

[https://doi.org/10.1016/S1359-6454\(00\)00352-9](https://doi.org/10.1016/S1359-6454(00)00352-9)

- Thermal energy
- Boundary energy
- Activation energy

Transform into neighbour with minimum new boundary energy



Geiger, J., Roósz, A., & Barkoczy, P. (2001). Simulation of grain coarsening in two dimensions by cellular-automaton. *Acta materialia*, 49(4), 623-629.



# First solution (Python)

- The students know some Python
- Python is easy!
- Python is there



# First solution (Python)

```
import numpy as np
import matplotlib.pyplot as plt
from skimage.measure import regionprops
from skimage.morphology import label
from timeit import timeit

R = 8.314 # ideal gas constant
GA = 10000 # activation energy
G0 = 3000 # maximum boundary energy

# thermal energy according to Maxwell-Boltzmann distribution at temperature T (kelvin)
def thermal(T):
    return -R*T*np.log(1-np.random.random())

# misorientation angle (in radians) between q and q1
def misorientation(q, q1, qmax):
    return (np.pi/2)*np.abs(q-q1)/qmax

# total boundary energy of cell with orientation q and neighbours with orientations q1
def cell_boundary(q, q1, qmax):
    gb = 0
    for qi in q1:
        dg = misorientation(q, qi, qmax)
        sin = np.sin(dg)
        if sin > 0:
            gb += G0*sin*(1-np.log(sin)) if sin > 0 else 0 # Read-Shockley equation
    return gb

def solve(q, T, f=0, n=0):
    i, d = 0, []
    qmax = np.max(q)
    while True:
        di = diameter(q)
        d.append(di)
        if f > 0 and i%f == 0:
            plot(q, qmax, "%d (%.2f)"%(i,di))
        i, (q, changed) = i+1, step(q, qmax, T)
        if (not changed and n == 0) or i == n:
            break
    # final microstructure
    if f > 0:
        plot(q, qmax, "%d (%.2f)"%(i,diameter(q)))
    return q, d

# return next orientation map q, after grain growth step at temperature T,
# and a boolean indicating if there was any change
def step(q, qmax, T):
    nr, nc = q.shape # number of rows and columns
    p, changed = np.copy(q), False # p is the map of product orientations
    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[i-1 if i>0 else nr-1, j]
            s = q[i+1 if i<nr-1 else 0, j]
            e = q[i, j+1 if j<nc-1 else 0]
            w = q[i, j-1 if j>0 else nc-1]
            q1 = [n, s, e, w]
            # check if a cell will transform into some neighbour
            if (q1.count(q[i, j]) == len(q1)) or (thermal(T) + cell_boundary(q[i, j], q1, qmax) < GA):
                continue
            # calculate new boundary energy when transforming into each neighbour
            gbn = cell_boundary(n, q1, qmax)
            gbs = cell_boundary(s, q1, qmax)
            gbe = cell_boundary(e, q1, qmax)
            gbw = cell_boundary(w, q1, qmax)
            gb1 = [gbn, gbs, gbe, gbw]
            # pick product for which the boundary energy is the lowest
            p[i, j] = q1[np.argmin(gb1)]
            # check if the orientation changed
            if p[i, j] != q[i, j]:
                changed = True
    return p, changed
```



# First solution (Python)

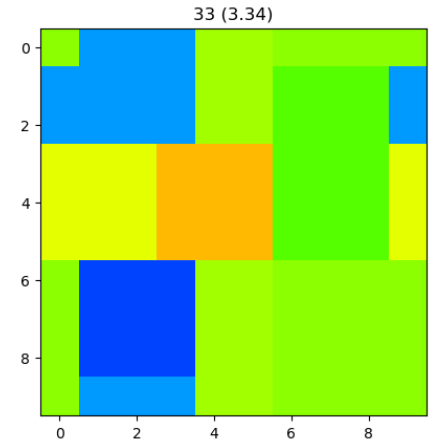
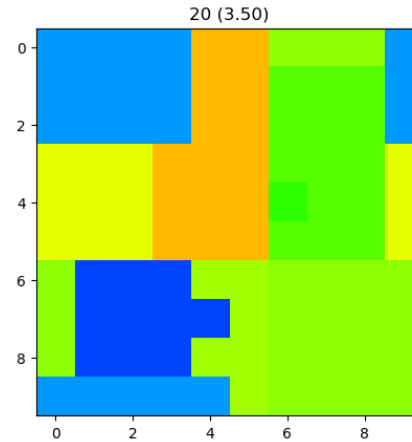
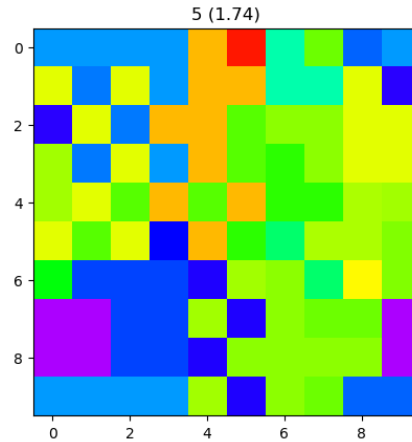
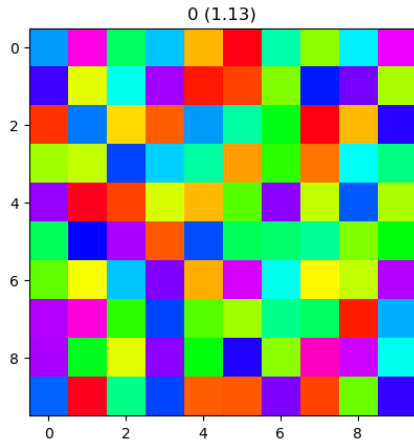
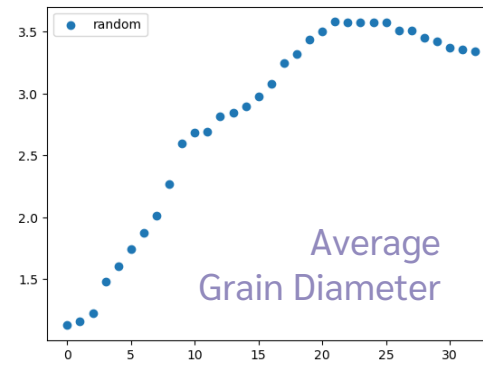
Loops

```
# return next orientation map q, after grain growth step at temperature T,
# and a boolean indicating if there was any change
def step(q, qmax, T):
    nr, nc = q.shape                # number of rows and columns
    p, changed = np.copy(q), False # p is the map of product orientations
    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[i-1 if i>0 else nr-1, j]
            s = q[i+1 if i<nr-1 else 0, j]
            e = q[i, j+1 if j<nc-1 else 0]
            w = q[i, j-1 if j>0 else nc-1]
            q1 = [n, s, e, w]
            # check if a cell will transform into some neighbour
            if (q1.count(q[i, j]) == len(q1)) or (thermal(T) + cell_boundary(q[i, j], q1, qmax) < GA):
                continue
            # calculate new boundary energy when transforming into each neighbour
            gbn = cell_boundary(n, q1, qmax)
            gbs = cell_boundary(s, q1, qmax)
            gbe = cell_boundary(e, q1, qmax)
            gbw = cell_boundary(w, q1, qmax)
            gb1 = [gbn, gbs, gbe, gbw]
            # pick product for which the boundary energy is the lowest
            p[i, j] = q1[np.argmin(gb1)]
            # check if the orientation changed
            if p[i, j] != q[i, j]:
                changed = True
    return p, changed
```



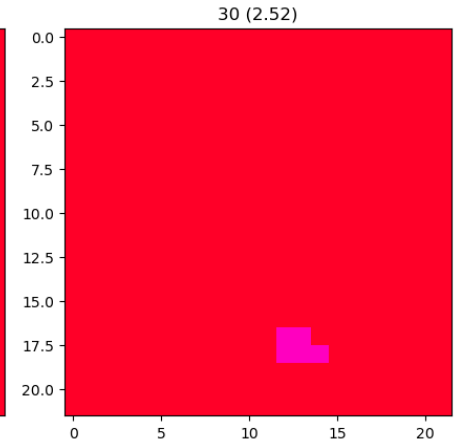
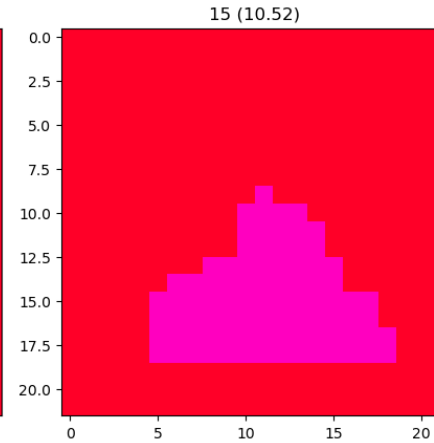
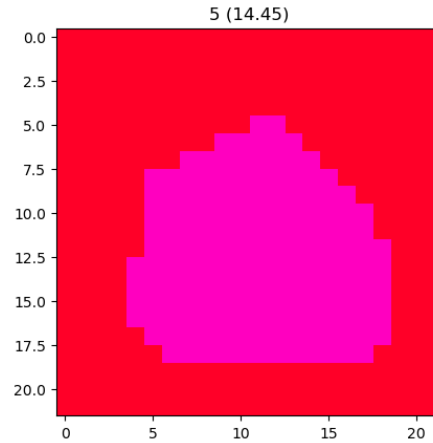
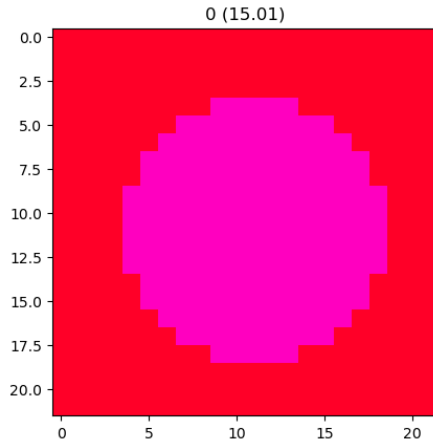
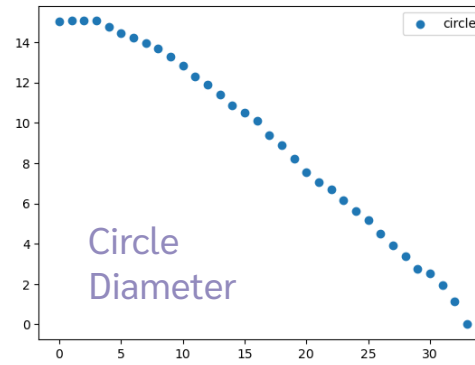
# First solution (Python)

Random microstructure



# First solution (Python)

Single circular grain



# First solution (Python)

```
# return next orientation map q, after grain growth step at temperature T,
# and a boolean indicating if there was any change
def step(q, qmax, T):
    nr, nc = q.shape # number of rows, number of columns
    p, changed = np.copy(q), False # p is the previous orientation map
    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[i-1 if i>0 else nr-1, j]
            s = q[i+1 if i<nr-1 else 0, j]
            e = q[i, j+1 if j<nc-1 else 0]
            w = q[i, j-1 if j>0 else nc-1]
            q1 = [n, s, e, w]
            # check if a cell will transform into a different neighbour
            if (q1.count(q[i, j]) == len(q1)) or (math.exp(-GA/(T + cell_boundary(q[i, j], q1, qmax))) < GA):
                continue
            # calculate new boundary energy for each neighbour
            gbn = cell_boundary(n, q1, qmax)
            gbs = cell_boundary(s, q1, qmax)
            gbe = cell_boundary(e, q1, qmax)
            gbw = cell_boundary(w, q1, qmax)
            gb1 = [gbn, gbs, gbe, gbw]
            # pick product for which the boundary energy is the lowest
            p[i, j] = q1[np.argmin(gb1)]
            # check if the orientation changed
            if p[i, j] != q[i, j]:
                changed = True
    return p, changed
```





# Partial Results

- Thermal energy
- Current boundary energy
- Boundary energy when transforming into each neighbour
- New orientations (transformed cells)



# First solution (Python)

```
# return next orientation map q, after grain growth step at temperature T,
# and a boolean indicating if there was any change
def step(q, qmax, T):
    nr, nc = q.shape                # number of rows and columns
    p, changed = np.copy(q), False # p is the map of product orientations
    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[i-1 if i>0 else nr-1, j]
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            e = q[i, j+1 if j<nc-1 else 0]
            w = q[i, j-1 if j>0 else nc-1]
            q1 = [n, s, e, w]
            # check if a cell will transform into some neighbour
            if (q1.count(q[i, j]) == len(q1)) or (thermal(T) + cell_boundary(q[i, j], q1, qmax) < GA):
                continue
            # calculate new boundary energy when transforming into each neighbour
            gbn = cell_boundary(n, q1, qmax)
            gbs = cell_boundary(s, q1, qmax)
            gbe = cell_boundary(e, q1, qmax)
            gbw = cell_boundary(w, q1, qmax)
            gb1 = [gbn, gbs, gbe, gbw]
            # pick product for which the boundary energy is the lowest
            p[i, j] = q1[np.argmin(gb1)]
            # check if the orientation changed
            if p[i, j] != q[i, j]:
                changed = True
    return p, changed
```



# Partial Results (Python)

```
# new version of step that plots partial results if plt is True
def step_partial(q, qmax, T, plt):
    nr, nc = q.shape                # number of rows and columns
    p, changed = np.copy(q), False  # p is the map of product orientations
    gt = np.zeros(q.shape)         # thermal energy
    gb = np.zeros(q.shape)         # current grain boundary energy
    gbn = np.zeros(q.shape)        # new grain boundary energy if transforms into north neighbour
    gbs = np.zeros(q.shape)        # new grain boundary energy if transforms into south neighbour
    gbe = np.zeros(q.shape)        # new grain boundary energy if transforms into east neighbour
    gbw = np.zeros(q.shape)        # new grain boundary energy if transforms into west neighbour
    m = np.zeros(q.shape, dtype=int) # neighbour for which new grain boundary energy is minimum
    t = np.zeros(q.shape, dtype=bool) # cells that will transform

    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[(i-1 if i>0 else nr-1),j]
            s = q[(i+1 if i<nr-1 else 0),j]
            e = q[i,(j+1 if j<nc-1 else 0)]
            w = q[i,(j-1 if j>0 else nc-1)]
            q1 = [n, s, e, w]
            # calculate current boundary energy
            gb[i,j] = cell_boundary(q[i,j], q1, qmax)
            # and when transforming into each neighbour
            gbn[i,j] = cell_boundary(n, q1, qmax)
            gbs[i,j] = cell_boundary(s, q1, qmax)
            gbe[i,j] = cell_boundary(e, q1, qmax)
            gbw[i,j] = cell_boundary(w, q1, qmax)
            gb1 = [gbn[i,j], gbs[i,j], gbe[i,j], gbw[i,j]]
            # find neighbour for which new boundary energy is minimum
            m[i,j] = np.argmin(gb1)
            # thermal energy
            gt[i,j] = thermal(T)
            # check if a cell will transform into some neighbour
            t[i,j] = (q1.count(q[i,j]) != len(q1)) and (gt[i,j] + gb[i,j]) >= GA)
            if not t[i,j]:
                continue
            # pick orientation of neighbour with minimum new boundary energy
            p[i,j] = q1[m[i,j]]
            # check if the cell changed of orientation
            if p[i,j] != q[i,j]:
                changed = True

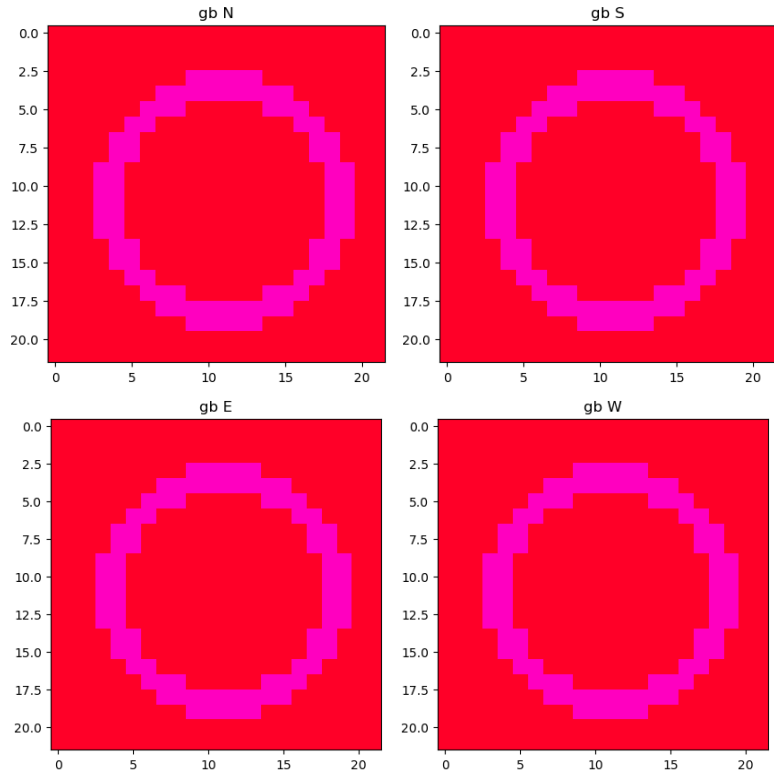
    if plt:
        plot_partial(gt, gb, [gbn, gbs, gbe, gbw], m, t)
    return p, changed
```

Define  
arrays

Fill arrays  
inside loop



# Partial Results (Python)

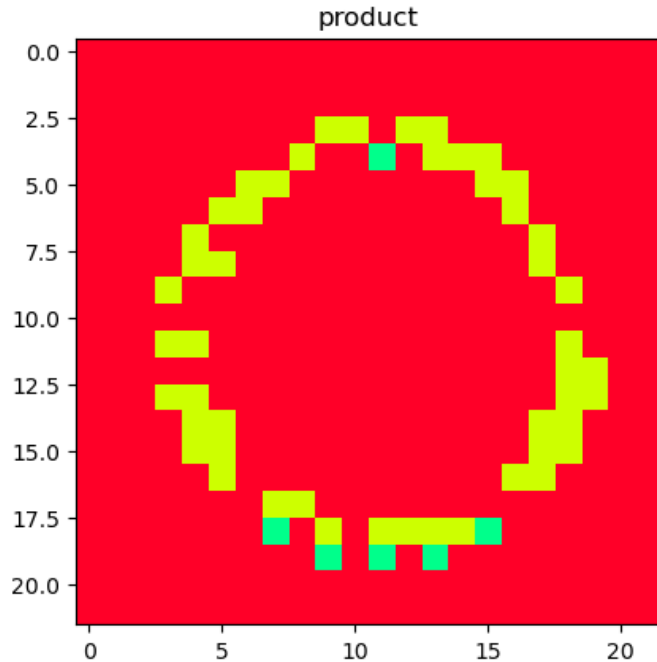


New  
boundary  
energy

Same change of  
energy for all  
neighbours



# Partial Results (Python)



Cells transform only  
into north and south  
neighbours



Randomize choice of  
neighbour with  
minimum energy



# Random Perturbation (Python)

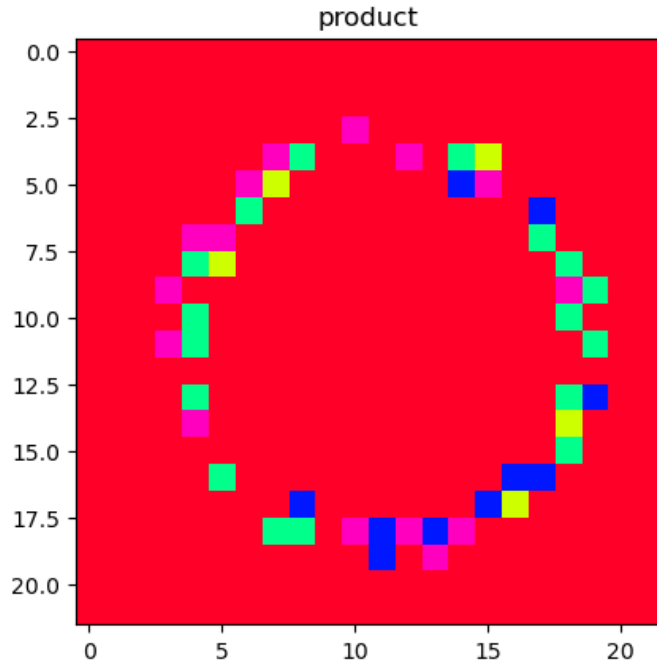
```
# new version of step_partial which applies a random perturbation to
# new boundary energies before searching the minimum
def step_rnd(q, qmax, T, plt):
    nr, nc = q.shape # number of rows and columns
    p, changed = np.copy(q), False # p is the map of product orientations
    gt = np.zeros(q.shape) # thermal energy
    gb = np.zeros(q.shape) # current grain boundary energy
    gbn = np.zeros(q.shape) # new grain boundary energy if transforms into north neighbour
    gbs = np.zeros(q.shape) # new grain boundary energy if transforms into south neighbour
    gbe = np.zeros(q.shape) # new grain boundary energy if transforms into east neighbour
    gbw = np.zeros(q.shape) # new grain boundary energy if transforms into west neighbour
    m = np.zeros(q.shape, dtype=int) # neighbour for which new grain boundary energy is minimum
    t = np.zeros(q.shape, dtype=bool) # cells that will transform
    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[(i-1 if i>0 else nr-1),j]
            s = q[(i+1 if i<nr-1 else 0),j]
            e = q[i,(j+1 if j<nc-1 else 0)]
            w = q[i,(j-1 if j>0 else nc-1)]
            q1 = [n, s, e, w]
            # calculate current boundary energy
            gb[i,j] = cell_boundary(q[i,j], q1, qmax)
            # and when transforming into each neighbour
            gbn[i,j] = cell_boundary(n, q1, qmax) * (1 + RP * np.random.random())
            gbs[i,j] = cell_boundary(s, q1, qmax) * (1 + RP * np.random.random())
            gbe[i,j] = cell_boundary(e, q1, qmax) * (1 + RP * np.random.random())
            gbw[i,j] = cell_boundary(w, q1, qmax) * (1 + RP * np.random.random())
            gb1 = [gbn[i,j], gbs[i,j], gbe[i,j], gbw[i,j]]
            # find neighbour for which new boundary energy is minimum
            m[i,j] = np.argmin(gb1)
            # thermal energy
            gt[i,j] = thermal(T)
            # check if a cell will transform into some neighbour
            t[i,j] = (q1.count(q[i,j]) != len(q1)) and (gt[i,j] + gb[i,j]) >= GA)
            if not t[i,j]:
                continue
            # pick orientation of neighbour with minimum new boundary energy
            p[i,j] = q1[m[i,j]]
            # check if the cell changed of orientation
            if p[i,j] != q[i,j]:
                changed = True
    if plt:
        plot_partial(gt, gb, [gbn, gbs, gbe, gbw], m, t)
    return p, changed
```

Apply small  
random  
perturbation to  
new boundary  
energies

RP = 0.01 # random perturbation



# Random Perturbation (Python)

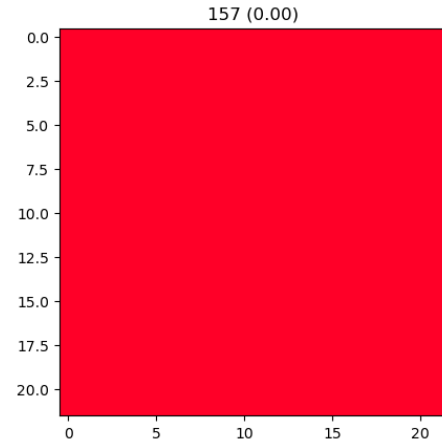
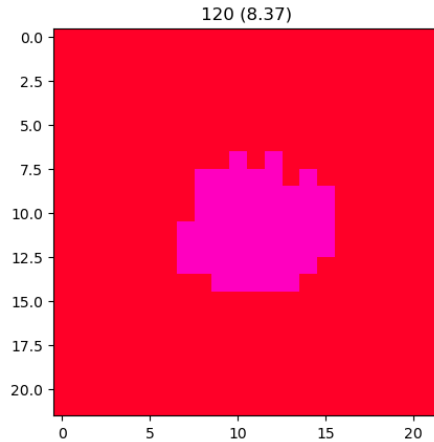
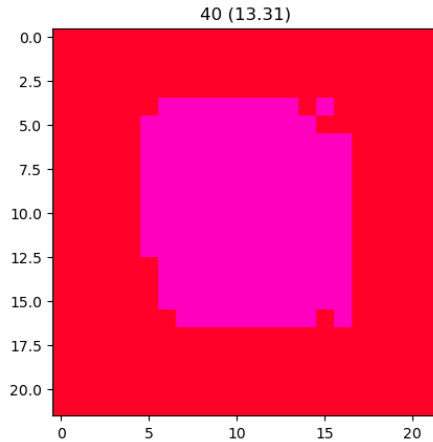
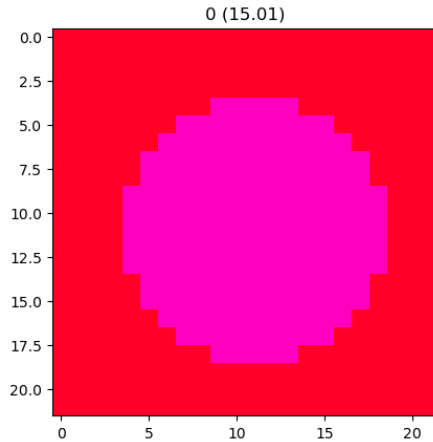
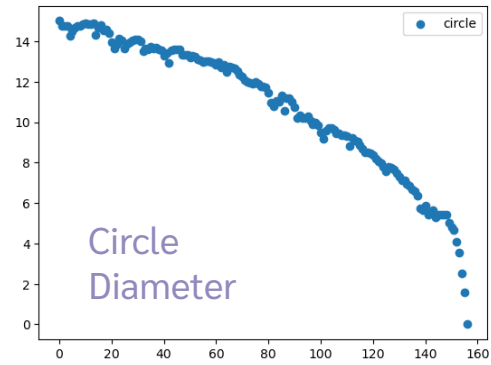


Neighbour into which  
to transform is  
randomly chosen



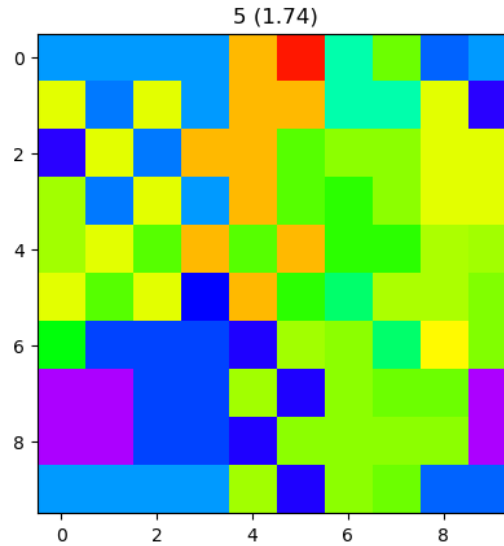
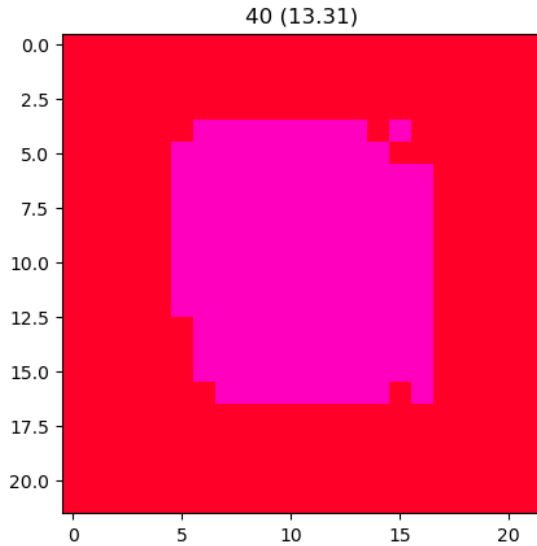
# Random Perturbation

Single circular grain





# Other problems



Unrealistic  
“checkerboard pattern”



Consider second  
order neighbours



# Random Perturbation (Python)

```
# new version of step_partial which applies a random perturbation to
# new boundary energies before searching the minimum
def step_rnd(q, qmax, T, plt):
    nr, nc = q.shape # number of rows and columns
    p, changed = np.copy(q), False # p is the map of product orientations
    gt = np.zeros(q.shape) # thermal energy
    gb = np.zeros(q.shape) # current grain boundary energy
    gbn = np.zeros(q.shape) # new grain boundary energy if transforms into north neighbour
    gbs = np.zeros(q.shape) # new grain boundary energy if transforms into south neighbour
    gbe = np.zeros(q.shape) # new grain boundary energy if transforms into east neighbour
    gbw = np.zeros(q.shape) # new grain boundary energy if transforms into west neighbour
    m = np.zeros(q.shape, dtype=int) # neighbour for which new grain boundary energy is minimum
    t = np.zeros(q.shape, dtype=bool) # cells that will transform
    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[(i-1 if i>0 else nr-1),j]
            s = q[(i+1 if i<nr-1 else 0),j]
            e = q[i,(j+1 if j<nc-1 else 0)]
            w = q[i,(j-1 if j>0 else nc-1)]
            q1 = [n, s, e, w]
            # calculate current boundary energy
            gb[i,j] = cell_boundary(q[i,j], q1, qmax)
            # and when transforming into each neighbour
            gbn[i,j] = cell_boundary(n, q1, qmax) * (1 + RP * np.random.random())
            gbs[i,j] = cell_boundary(s, q1, qmax) * (1 + RP * np.random.random())
            gbe[i,j] = cell_boundary(e, q1, qmax) * (1 + RP * np.random.random())
            gbw[i,j] = cell_boundary(w, q1, qmax) * (1 + RP * np.random.random())
            gb1 = [gbn[i,j], gbs[i,j], gbe[i,j], gbw[i,j]]
            # find neighbour for which new boundary energy is minimum
            m[i,j] = np.argmin(gb1)
            # thermal energy
            gt[i,j] = thermal(T)
            # check if a cell will transform into some neighbour
            t[i,j] = (q1.count(q[i,j]) != len(q1)) and (gt[i,j] + gb[i,j] >= GA)
            if not t[i,j]:
                continue
            # pick orientation of neighbour with minimum new boundary energy
            p[i,j] = q1[m[i,j]]
            # check if the cell changed of orientation
            if p[i,j] != q[i,j]:
                changed = True
    if plt:
        plot_partial(gt, gb, [gbn, gbs, gbe, gbw], m, t)
    return p, changed
```

RP = 0.01 # random perturbation



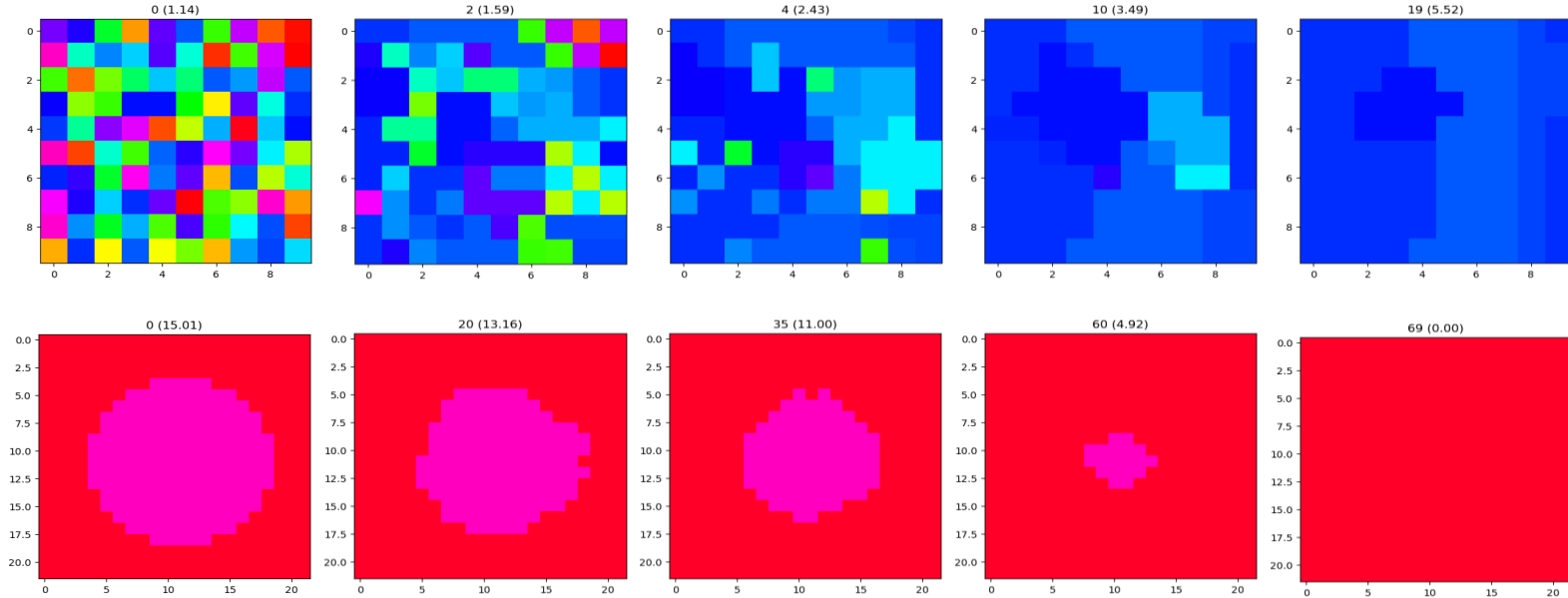
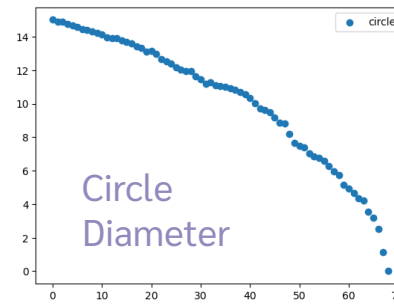
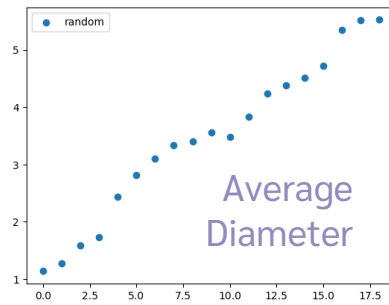
# 2<sup>nd</sup> order neighbours (Python)

```
# new version of step_rnd which takes into account the influence of secon-order neighbours
def step_rnd2(q, qmax, T, plt):
    nr, nc = q.shape # number of rows and columns
    p, changed = np.copy(q), False # p is the map of product orientations
    gt = np.zeros(q.shape) # thermal energy
    gb = np.zeros(q.shape) # current grain boundary energy
    gbn = np.zeros(q.shape) # new grain boundary energy if transforms into north neighbour
    gbs = np.zeros(q.shape) # new grain boundary energy if transforms into south neighbour
    gbe = np.zeros(q.shape) # new grain boundary energy if transforms into east neighbour
    gbw = np.zeros(q.shape) # new grain boundary energy if transforms into west neighbour
    m = np.zeros(q.shape, dtype=int) # neighbour for which new grain boundary energy is minimum
    t = np.zeros(q.shape, dtype=bool) # cells that will transform
    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[(i-1 if i>0 else nr-1),j]
            s = q[(i+1 if i<nr-1 else 0),j]
            e = q[i,(j+1 if j<nc-1 else 0)]
            w = q[i,(j-1 if j>0 else nc-1)]
            q1 = [n, s, e, w]
            # calculate current boundary energy
            gb[i,j] = cell_boundary(q[i,j], q1, qmax)
            # and when transforming into each neighbour
            gbn[i,j] = cell_boundary(n, q1, qmax)
            gbs[i,j] = cell_boundary(s, q1, qmax)
            gbe[i,j] = cell_boundary(e, q1, qmax)
            gbw[i,j] = cell_boundary(w, q1, qmax)
            if SF > 0:
                nw = q[(i-1 if i>0 else nr-1),(j-1 if j>0 else nc-1)]
                se = q[(i+1 if i<nr-1 else 0),(j+1 if j<nc-1 else 0)]
                ne = q[(i-1 if i>0 else nr-1),(j+1 if j<nc-1 else 0)]
                sw = q[(i+1 if i<nr-1 else 0),(j-1 if j>0 else nc-1)]
                q2 = [nw, se, ne, sw]
                gbn[i,j] += SF * cell_boundary(n, q2, qmax)
                gbs[i,j] += SF * cell_boundary(s, q2, qmax)
                gbe[i,j] += SF * cell_boundary(e, q2, qmax)
                gbw[i,j] += SF * cell_boundary(w, q2, qmax)
            gbn[i,j] = 1 + RP * np.random.random()
            gbs[i,j] = 1 + RP * np.random.random()
            gbe[i,j] = 1 + RP * np.random.random()
            gbw[i,j] = 1 + RP * np.random.random()
            gbi = [gbn[i,j], gbs[i,j], gbe[i,j], gbw[i,j]]
            # find neighbour for which new boundary energy is minimum
            m[i,j] = np.argmin(gbi)
            # thermal energy
            gt[i,j] = thermal(T)
            # check if a cell will transform into some neighbour
            t[i,j] = (q1.count(q[i,j]) != len(q1)) and (gt[i,j] + gb[i,j]) >= GA
            if not t[i,j]:
                continue
            # pick orientation of neighbour with minimum new boundary energy
            p[i,j] = q1[m[i,j]]
            # check if the cell changed of orientation
            if p[i,j] != q[i,j]:
                changed = True
    if plt:
        plot_partial(gt, gb, [gbn, gbs, gbe, gbw], 'NSEW', m, t)
    return p, changed
```

RP = 0.01 # random perturbation  
SF = 0.25 # second-order-neighbours factor



# 2<sup>nd</sup> order



# 2<sup>nd</sup> order neighbours (Python)

```
# new version of step_rnd which takes into account the influence of second-order neighbours
def step_rnd2(q, qmax, T, plt):
    nr, nc = q.shape # number of rows and columns
    p, changed = np.copy(q), False # p is the map of product orientations
    gt = np.zeros(q.shape) # thermal energy
    gb = np.zeros(q.shape) # current grain boundary energy
    gbn = np.zeros(q.shape) # new grain boundary energy if transforms into north neighbour
    gbs = np.zeros(q.shape) # new grain boundary energy if transforms into south neighbour
    gbe = np.zeros(q.shape) # new grain boundary energy if transforms into east neighbour
    gbw = np.zeros(q.shape) # new grain boundary energy if transforms into west neighbour
    m = np.zeros(q.shape, dtype=int) # neighbour for which new grain boundary energy is minimum
    t = np.zeros(q.shape, dtype=bool) # cells that will transform
    for i in range(nr):
        for j in range(nc):
            # 1st-order neighbours: north, south, east, west
            n = q[(i-1 if i>0 else nr-1),j]
            s = q[(i+1 if i<nr-1 else 0),j]
            e = q[i,(j+1 if j<nc-1 else 0)]
            w = q[i,(j-1 if j>0 else nc-1)]
            q1 = [n, s, e, w]
            # calculate current boundary energy
            gb[i,j] = cell_boundary(q[i,j], q1, qmax)
            # and when transforming into each neighbour
            gbn[i,j] = cell_boundary(n, q1, qmax)
            gbs[i,j] = cell_boundary(s, q1, qmax)
            gbe[i,j] = cell_boundary(e, q1, qmax)
            gbw[i,j] = cell_boundary(w, q1, qmax)
            if SF > 0:
                nw = q[(i-1 if i>0 else nr-1),(j-1 if j>0 else nc-1)]
                se = q[(i+1 if i<nr-1 else 0),(j+1 if j<nc-1 else 0)]
                ne = q[(i-1 if i>0 else nr-1),(j+1 if j<nc-1 else 0)]
                sw = q[(i+1 if i<nr-1 else 0),(j-1 if j>0 else nc-1)]
                q2 = [nw, se, ne, sw]
                gbn[i,j] += SF * cell_boundary(n, q2, qmax)
                gbs[i,j] += SF * cell_boundary(s, q2, qmax)
                gbe[i,j] += SF * cell_boundary(e, q2, qmax)
                gbw[i,j] += SF * cell_boundary(w, q2, qmax)
            gbn[i,j] = 1 + RP * np.random.random()
            gbs[i,j] = 1 + RP * np.random.random()
            gbe[i,j] = 1 + RP * np.random.random()
            gbw[i,j] = 1 + RP * np.random.random()
            gbi = [gbn[i,j], gbs[i,j], gbe[i,j], gbw[i,j]]
            # find neighbour for which new boundary energy is minimum
            m[i,j] = np.argmin(gbi)
            # thermal energy
            gt[i,j] = thermal(T)
            # check if a cell will transform into some neighbour
            t[i,j] = (q1.count(q[i,j]) != len(q1)) and (gt[i,j] + gb[i,j] >= GA)
            if not t[i,j]:
                continue
            # pick orientation of neighbour with minimum new boundary energy
            p[i,j] = q1[m[i,j]]
            # check if the cell changed of orientation
            if p[i,j] != q[i,j]:
                changed = True
    if plt:
        plot_partial(gt, gb, [gbn, gbs, gbe, gbw], 'NSEW', m, t)
    return p, changed
```

Too much repetitive code

Hard to understand what the model is doing

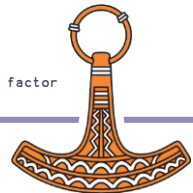
Easy to make mistakes

Slow



ARRAY PROGRAMMING

RP = 0.01 # random perturbation  
SF = 0.25 # second-order-neighbours factor



# 2<sup>nd</sup> order neighbours (NumPy)

```
# new version of step_rnd2 using numpy functions
def step_numpy2(q, qmax, T, plt):
    q1 = [np.roll(q, i, axis=a) for i, a in zip([-1,1,1,-1], [0,0,1,1])] # 1st-order neighbours
    gt, gb = -R*T*np.log(1-np.random.random(q.shape)), boundary(q, q1, qmax) # thermal and boundary energies
    gb1 = np.array([boundary(qi, q1, qmax) for qi in q1]) # new boundary energies
    if SF > 0:
        q2 = [np.roll(qi, i, axis=a) for qi, i, a in zip(q1, [-1,1,-1,1], [1,1,0,0])] # 2nd-order neighbours
        gb1 += SF * np.array([boundary(qi, q2, qmax) for qi in q1]) # their boundary energies
    gb1 *= 1 + RP * np.random.random(gb1.shape) # apply random perturbation
    t, m = (gt + gb >= GA), np.argmin(gb1, axis=0) # cells to transform and minimum
    p = np.where(t, np.choose(m, q1), q) # product orientations
    if plt:
        plot_partial(gt, gb, gb1, 'NSEW', m, t)
    return p, not np.array_equal(q, p)
```

Less code, less mistakes, more performance

Less likely to be written by students

```
RP = 0.01 # random perturbation
SF = 0.25 # second-order-neighbours factor
```



# 2<sup>nd</sup> order neighbours (NumPy)

```
# new version of step_rnd2 using numpy functions
def step_numpy2(q, qmax, T, plt):
    q1 = [np.roll(q, i, axis=a) for i, a in zip([-1,1,1,-1], [0,0,1,1])] # 1st-order neighbours
    gt, gb = -R*T*np.log(1-np.random.random(q.shape)), boundary(q, q1, qmax) # thermal and boundary energies
    gb1 = np.array([boundary(qi, q1, qmax) for qi in q1]) # new boundary energies
    if SF > 0:
        q2 = [np.roll(qi, i, axis=a) for qi, i, a in zip(q1, [-1,1,-1,1], [1,1,0,0])] # 2nd-order neighbours
        gb1 += SF * np.array([boundary(qi, q2, qmax) for qi in q1]) # their boundary energies
    gb1 *= 1 + RP * np.random.random(gb1.shape) # apply random perturbation
    t, m = (gt + gb >= GA), np.argmin(gb1, axis=0) # cells to transform and minimum
    p = np.where(t, np.choose(m, q1), q) # product orientations
    if plt:
        plot_partial(gt, gb, gb1, 'NSEW', m, t)
    return p, not np.array_equal(q, p)

# return boundary energies for orientations q with neighbours q1
def boundary(q, q1, qmax):
    dg = np.array([misorientation(q, qi, qmax) for qi in q1])
    sin = np.sin(dg)
    gbi = G0 * sin * (1 - np.log(sin, out=np.ones_like(dg), where=dg>0))
    return np.sum(gbi, axis=0)

RP = 0.01 # random perturbation
SF = 0.25 # second-order-neighbours factor
```



# NumPy

- Array objects
  - The N-dimensional array ([ndarray](#))
  - Scalars
  - Data type objects ([dtype](#))
  - Indexing routines
  - Iterating over arrays
  - Standard array subclasses
  - Masked arrays
  - The array interface protocol
  - Datetimes and Timedeltas
- Array API Standard Compatibility
  - Table of Differences between [numpy.array\\_api](#) and [numpy](#)
- Constants
  - [Inf](#)
  - [Infinity](#)
  - [NaN](#)
  - [NINF](#)
  - [NZERO](#)
  - [NaN](#)
  - [PINF](#)
  - [PZERO](#)
  - [e](#)
  - [euler\\_gamma](#)
  - [inf](#)
  - [infty](#)
  - [nan](#)
  - [newaxis](#)
  - [pi](#)
- Universal functions ([ufunc](#))
- [ufunc](#)
  - Available [ufuncs](#)
- Routines
  - Array creation routines
  - Array manipulation routines
  - Binary operations
  - String operations
  - C-Types foreign function interface ([numpy.ctypeslib](#))
  - Datetime support functions
  - Data type routines
  - Mathematical functions with automatic domain
  - Floating point error handling
  - Discrete Fourier Transform ([numpy.fft](#))
  - Functional programming
  - NumPy-specific help functions
  - Input and output
  - Linear algebra ([numpy.linalg](#))
  - Logic functions
  - Masked array operations
  - Mathematical functions
  - Matrix library ([numpy.matlib](#))
  - Miscellaneous routines
  - Padding Arrays
  - Polynomials
  - Random sampling ([numpy.random](#))
  - Set routines
  - Sorting, searching, and counting
- Statistics
  - Test Support ([numpy.testing](#))
  - Support for testing overrides ([numpy.testing.overrides](#))
  - Window functions
- Typing ([numpy.typing](#))
  - [Mypy](#) plugin
  - Differences from the runtime NumPy API
  - API
- Global state
  - Performance-related options
  - Debugging-related options
  - Testing planned future behavior
- Packaging ([numpy.distutils](#))
  - Modules in [numpy.distutils](#)
  - Configuration class
  - Building Installable C libraries
  - Conversion of [.src](#) files
- NumPy [distutils](#) - users guide
  - [SciPy](#) structure
  - Requirements for [SciPy](#) packages
  - The [setup.py](#) file
  - The [\\_\\_init\\_\\_.py](#) file
  - Extra features in NumPy [Distutils](#)
- Status of [numpy.distutils](#) and migration advice
  - Migration advice
  - Interaction of [numpy.distutils](#) with [setuptools](#)
- NumPy C-API
- Python Types and C-Structures
  - System configuration
  - Data Type API
  - Array API
  - Array Iterator API
  - UFunc API
  - Generalized Universal Function API
  - NumPy core libraries
  - C API Deprecations
  - Memory management in NumPy
- CPU/SIMD Optimizations
  - CPU build options
  - How does the CPU dispatcher work?
- NumPy security
  - Advice for using NumPy on untrusted data
- NumPy and SWIG
  - [numpy.i](#): a SWIG Interface File for NumPy
  - Testing the [numpy.i](#) Typemaps





# NumPy

## Available ufuncs

### Math operations

- Array operations
  - add(x1, x2, /[, out, where, casting, order, ...])
  - subtract(x1, x2, /[, out, where, casting, ...])
  - multiply(x1, x2, /[, out, where, casting, ...])
  - matmul(x1, x2, /[, out, casting, order, ...])
  - divide(x1, x2, /[, out, where, casting, ...])
  - logaddexp(x1, x2, /[, out, where, casting, ...])
  - logaddexp2(x1, x2, /[, out, where, casting, ...])
  - true\_divide(x1, x2, /[, out, where, ...])
  - floor\_divide(x1, x2, /[, out, where, ...])
  - negative(x, /[, out, where, casting, order, ...])
  - positive(x, /[, out, where, casting, order, ...])
- Array arithmetic
  - power(x1, x2, /[, out, where, casting, ...])
  - float\_power(x1, x2, /[, out, where, ...])
- Constants
  - remainder(x1, x2, /[, out, where, casting, ...])
  - mod(x1, x2, /[, out, where, casting, order, ...])
  - fmod(x1, x2, /[, out, where, casting, ...])
  - divmod(x1, x2[, out1, out2], / [[, out, ...])
  - absolute(x, /[, out, where, casting, order, ...])
  - fabs(x, /[, out, where, casting, order, ...])
  - rint(x, /[, out, where, casting, order, ...])
  - sign(x, /[, out, where, casting, order, ...])
  - heaviside(x1, x2, /[, out, where, casting, ...])
  - conj(x, /[, out, where, casting, order, ...])
  - conjugate(x, /[, out, where, casting, ...])
  - exp(x, /[, out, where, casting, order, ...])
  - exp2(x, /[, out, where, casting, order, ...])
  - log(x, /[, out, where, casting, order, ...])
  - log2(x, /[, out, where, casting, order, ...])
  - log10(x, /[, out, where, casting, order, ...])
  - expm1(x, /[, out, where, casting, order, ...])
  - Universal functions
    - log1p(x, /[, out, where, casting, order, ...])

- sqrt(x, /[, out, where, casting, order, ...])
- square(x, /[, out, where, casting, order, ...])
- cbqrt(x, /[, out, where, casting, order, ...])
- reciprocal(x, /[, out, where, casting, ...])
- gcd(x1, x2, /[, out, where, casting, order, ...])
- lcm(x1, x2, /[, out, where, casting, order, ...])

### Trigonometric functions

- sin(x, /[, out, where, casting, order, ...])
- cos(x, /[, out, where, casting, order, ...])
- tan(x, /[, out, where, casting, order, ...])
- arcsin(x, /[, out, where, casting, order, ...])
- arccos(x, /[, out, where, casting, order, ...])
- arctan(x, /[, out, where, casting, order, ...])
- arctan2(x1, x2, /[, out, where, casting, ...])
- hypot(x1, x2, /[, out, where, casting, ...])
- sinh(x, /[, out, where, casting, order, ...])
- cosh(x, /[, out, where, casting, order, ...])
- tanh(x, /[, out, where, casting, order, ...])
- arcsinh(x, /[, out, where, casting, order, ...])
- arccosh(x, /[, out, where, casting, order, ...])
- arctanh(x, /[, out, where, casting, order, ...])
- degrees(x, /[, out, where, casting, order, ...])
- radians(x, /[, out, where, casting, order, ...])
- deg2rad(x, /[, out, where, casting, order, ...])
- rad2deg(x, /[, out, where, casting, order, ...])

### Bit-twiddling functions

- bitwise\_and(x1, x2, /[, out, where, ...])
- bitwise\_or(x1, x2, /[, out, where, casting, ...])
- bitwise\_xor(x1, x2, /[, out, where, ...])
- invert(x, /[, out, where, casting, order, ...])
- left\_shift(x1, x2, /[, out, where, casting, ...])
- right\_shift(x1, x2, /[, out, where, ...])

### Comparison functions

- greater(x1, x2, /[, out, where, casting, ...])
- greater\_equal(x1, x2, /[, out, where, ...])
- less(x1, x2, /[, out, where, casting, ...])
- less\_equal(x1, x2, /[, out, where, casting, ...])
- not\_equal(x1, x2, /[, out, where, casting, ...])
- equal(x1, x2, /[, out, where, casting, ...])
- logical\_and(x1, x2, /[, out, where, ...])
- logical\_or(x1, x2, /[, out, where, casting, ...])
- logical\_xor(x1, x2, /[, out, where, ...])
- logical\_not(x, /[, out, where, casting, ...])
- maximum(x1, x2, /[, out, where, casting, ...])
- minimum(x1, x2, /[, out, where, casting, ...])
- fmax(x1, x2, /[, out, where, casting, ...])
- fmin(x1, x2, /[, out, where, casting, ...])

### Floating functions

- isfinite(x, /[, out, where, casting, order, ...])
- isinf(x, /[, out, where, casting, order, ...])
- isnan(x, /[, out, where, casting, order, ...])
- isnat(x, /[, out, where, casting, order, ...])
- fabs(x, /[, out, where, casting, order, ...])
- signbit(x, /[, out, where, casting, order, ...])
- copysign(x1, x2, /[, out, where, casting, ...])
- nextafter(x1, x2, /[, out, where, casting, ...])
- spacing(x, /[, out, where, casting, order, ...])
- modf(x[, out1, out2], / [[, out, where, ...])
- ldexp(x1, x2, /[, out, where, casting, ...])
- frexp(x[, out1, out2], / [[, out, where, ...])
- fmod(x1, x2, /[, out, where, casting, ...])
- floor(x, /[, out, where, casting, order, ...])
- ceil(x, /[, out, where, casting, order, ...])
- trunc(x, /[, out, where, casting, order, ...])

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## Array objects

### The N-dimensional array (ndarray)

- Constructing arrays
- Indexing arrays
- Internal memory layout of an ndarray
- Array attributes
- Array methods
- Arithmetic, matrix multiplication, and comparison operations
- Special methods

### Scalars

- Built-in scalar types
- Attributes
- Indexing
- Methods
- Defining new types

### Data type objects (dtype)

- Specifying and constructing data types
- dtype

## Indexing routines

- Generating index arrays
- Indexing-like operations
- Inserting data into arrays
- Iterating over arrays

### Iterating over arrays

- Single array iteration
- Broadcasting array iteration
- Putting the Inner Loop in Cython

### Standard array subclasses

- Special attributes and methods
- Matrix objects
- Memory-mapped file arrays
- Character arrays (numpy.char)
- Record arrays (numpy.rec)
- Masked arrays (numpy.ma)
- Standard container class
- Array Iterators

### Masked arrays

- The numpy.ma module
- Using numpy.ma

## Examples

- Constants of the numpy.ma module
- The MaskedArray class
- MaskedArray methods
- Masked array operations

### The array interface protocol

- Python side
- C-struct access
- Type description examples
- Differences with Array interface (Version 2)

### Datetimes and Timedeltas

- Datetime64 Conventions and Assumptions
- Basic Datetimes
- Datetime and Timedelta Arithmetic
- Datetime Units
- Business Day Functionality
- Datetime64 shortcomings

## Functions

- out, where, casting, ...])
- 2, /[, out, where, ...])
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- ..., out, where, casting, ...])
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- remainder(x1, x2, /[, out, where, casting, ...])
- mod(x1, x2, /[, out, where, casting, order, ...])
- fmod(x1, x2, /[, out, where, casting, ...])
- divmod(x1, x2[, out1, out2], / [[, out, ...])
- absolute(x, /[, out, where, casting, order, ...])
- fabs(x, /[, out, where, casting, order, ...])
- rint(x, /[, out, where, casting, order, ...])
- sign(x, /[, out, where, casting, order, ...])
- heaviside(x1, x2, /[, out, where, casting, ...])
- conj(x, /[, out, where, casting, order, ...])
- conjugate(x, /[, out, where, casting, ...])
- exp(x, /[, out, where, casting, order, ...])
- exp2(x, /[, out, where, casting, order, ...])
- log(x, /[, out, where, casting, order, ...])
- log2(x, /[, out, where, casting, order, ...])
- log10(x, /[, out, where, casting, order, ...])
- expm1(x, /[, out, where, casting, order, ...])
- log1p(x, /[, out, where, casting, order, ...])

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- sinh(x, /[, out, where, casting, order, ...])
- cosh(x, /[, out, where, casting, order, ...])
- tanh(x, /[, out, where, casting, order, ...])
- arcsinh(x, /[, out, where, casting, order, ...])
- arccosh(x, /[, out, where, casting, order, ...])
- arctanh(x, /[, out, where, casting, order, ...])
- degrees(x, /[, out, where, casting, order, ...])
- radians(x, /[, out, where, casting, order, ...])
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- rad2deg(x, /[, out, where, casting, order, ...])

### Bit-tiddling functions

- bitwise\_and(x1, x2, /[, out, where, ...])
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- right\_shift(x1, x2, /[, out, where, ...])

### Floating functions

- isfinite(x, /[, out, where, casting, order, ...])
- isinf(x, /[, out, where, casting, order, ...])
- isnan(x, /[, out, where, casting, order, ...])
- isnat(x, /[, out, where, casting, order, ...])
- fabs(x, /[, out, where, casting, order, ...])
- signbit(x, /[, out, where, casting, order, ...])
- copysign(x1, x2, /[, out, where, casting, ...])
- nextafter(x1, x2, /[, out, where, casting, ...])
- spacing(x, /[, out, where, casting, order, ...])
- modf(x[, out1, out2], / [[, out, where, ...])
- ldexp(x1, x2, /[, out, where, casting, ...])
- frexp(x[, out1, out2], / [[, out, where, ...])
- fmod(x1, x2, /[, out, where, casting, ...])
- floor(x, /[, out, where, casting, order, ...])
- ceil(x, /[, out, where, casting, order, ...])
- trunc(x, /[, out, where, casting, order, ...])



## Array objects

### The N-dimensional array (ndarray)

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### Scalars

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Attribute

Indexing

Methods

Defining

### Data type objects (dtype)

Specifying

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mod(x)

fmod(x)

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fabs(x)

rint(x)

sign(x)

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conj(x)

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exp(x)

exp2(x)

log(x)

log2(x)

log10(x)

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log1p

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## Indexing routines

Generating index arrays

Indexing-like operations

Inserting data into arrays

Iterating over arrays

## Examples

Constants of the numpy.ma module

The MaskedArray class

MaskedArray methods

Masked array operations

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From existing data

Creating record arrays (numpy.rec)

Creating character arrays (numpy.char)

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The Matrix class

#### Array manipulation routines

Basic operations

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Changing kind of array

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String operations

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ndpointer

c\_intp

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numpy.result\_type

numpy.common\_type

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numpy.matlib.eye

numpy.matlib.identity

numpy.matlib.repmat

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Various windows



Learning Python might be easy

Mastering NumPy is hard





# APL Grain Growth Model

```
gg←{
  area←{0⇒ω:+/0≠,ω ∘ (ι2)×.((≠,ω){αα÷1[+/,ω≠1φ[α]ω})cω}
  boundary←([/,ω){+G0×s×1-⊗@(0<)s←100|↑(ω-cα)÷2×αα}
  step←{gt←-R×α×⊗?≠~ω ∘ gb←ω boundary↑q1←1 -1(θ'',φ'')cω ∘ t←(⇒(cω)v.≠q1)^GA≤gt+gb
    gb1←(1+RP×?≠~q1)×{ω+SF×q1 boundary''c1 -1(θ''∘(2↓↑),φ''∘(2↓φ))q1}*(SF>0)boundary''c~q1
    m←⇒∠ö1q↑gb1 ∘ (t/ö,ϕm[]ö0 1q↑q1)@(t~)ω}
  R←8.314 ∘ (GA G0 SF RP)←αα ∘ ⇒{α(2×(ω÷01)*÷2)}/α{q((⇒φω),area↑q←α step⇒ω)}*ωω↑ω(area ω)
}
```

Whole solution: step calculation, solve, calculate average diameter

Only APL primitives, no external libraries or system commands

Partial results available in the interpreter



# APL Grammar

```

gg←{
  area←{0=ω:
  boundary←(
  step←{gt←-R
  gb1←(1+
  m←÷○Δö1
  R←8.314 ◇ (
}

```

Whole solution

Only APL program

Partial result

```

gg←{ A α: parameters (GA, GO, SF, RP), ωω: repetitions, α: temperature, ω: initial orientations

  R←8.314
  (GA GO SF RP)←α
  qmax←[/,ω
  misorientations←{○|t(ω-α)÷2×qmax}
  boundary←{
    s←1○α misorientations ω
    +fGO×s×1-ε@{0<ω}s
  }
  boundary2←{
    SF≤0:0
    q2←(1 -1ε"2tω),1 -1φ"2tφω
    SF×ω boundary"cq2
  }
  step←{
    gt←-Rα×ε?≠ω
    q1←(1 -1ε"ω),1 -1φ"ω
    gb←ω boundary q1
    t←(vft(εω)≠q1)∧GA≤gt+gb
    gb1←q1 boundary"cq1
    gb1←gb1+boundary2 q1
    gb1←gb1×1+RP×?≠q1
    m←÷ö1Δö1φtgb1
    p←φm]ö0 1φtq1
    ((,t)/,p)@(t"ω)
  }
  area←{
    0=ω:+/0≠,ω
    v←(≠,ω)÷1[/,ω≠1φω
    h←(≠,ω)÷1[/,ω≠1εω
    w×h
  }
  next←{(q a)+ω ◇ n+α step q ◇ n(a,area n)}
  (q a)+α(next×ωω)ω(area ω)
  q(2×(a÷ö1)×÷2)
}

```

```

sgt+gb
)boundary"ε<~q1
)}*ωω-ω(area ω)

```

eter



# APL Grain Growth Model

```
gg←{
  area←{0⇒ω:+/0≠,ω ∘ (ι2)×.((≠,ω){αα÷1[+/,ω≠1φ[α]ω})cω}
  boundary←([/,ω){+≠G0×s×1-⊗@(0◦<)s←100|↑(ω-cα)÷2×αα}
  step←{gt←-R×α×⊗?≠~ω ∘ gb←ω boundary↑q1←1 -1(θ'',φ'')cω ∘ t←(⇒(cω)v.≠q1)^GA≤gt+gb
    gb1←(1+RP×?≠~q1)×{ω+SF×q1 boundary''c1 -1(θ''◦(2↓↑),φ''◦(2↓φ))q1}*(SF>0)boundary''◦c~q1
    m←⇒◦Δ◦1◦↑gb1 ∘ (t/◦,Δm[]◦0 1◦↑q1)@(t~)ω}
  R←8.314 ∘ (GA G0 SF RP)←αα ∘ ⇒{α(2×(ω÷◦1)*÷2)}/α{q((⇒φω),area↑q←α step⇒ω)}*ωω↑ω(area ω)
}
```





# Py'N'APL Grain Growth Model

```
from pynapl import APL
```

```
apl = APL.APL() # APL session
```

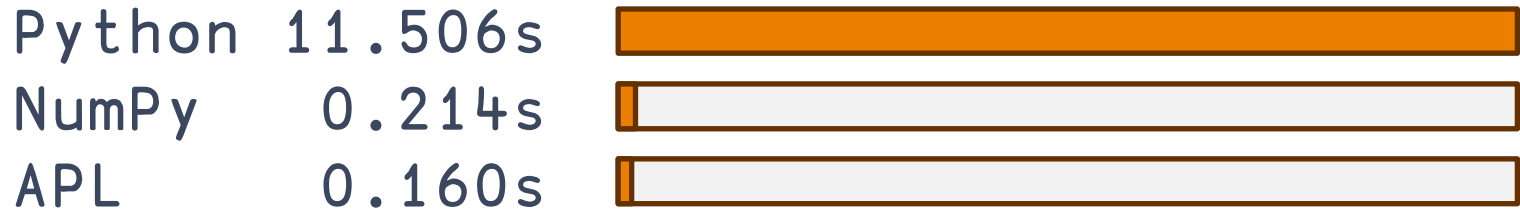
```
gg = apl.op('''{
    area←{0=ω:+/0≠,ω ⋄ (ι2)×.((≠,ω){αα÷1[+/ ,ω≠1φ[α]ω})cω}
    boundary←([/,ω){+/G0×s×1-⊖@(0<)s←1⊖|↑(ω-cα)÷2×αα}
    step←{gt←-R×α×⊖?≠~ω ⋄ gb←ω boundary↑q1←1 -1(⊖,φ)⋄ t←(cω)∨.≠q1)^GA≤gt+gb
        gb1←(1+RP×?≠~q1)×{ω+SF×q1 boundary"⋄1 -1(⊖∘(2↑),φ∘(2↑φ))q1}*(SF>0)boundary"⋄c~q1
        m←∘Φ∘1⋄↑gb1 ⋄ (t/ö,⋄m[]∘0 1⋄↑q1)@(t~)ω}
    R←8.314 ⋄ (GA G0 SF RP)←αα ⋄ ∘{α(2×(ω÷01)*÷2)} / α{q((cφω),area↑q←α step>ω)}*ωω↑ω(area ω)
}''')
```

```
def solve_apl(q, T, n=0):
    return gg((GA, G0, SF, RP), apl.fn('≡ö>') if n == 0 else n)(T, q)
```



# Benchmarks

Circle problem (radius 25, random perturbation, 2<sup>nd</sup> order)



The APL version runs faster in the interpreter, and it could be faster

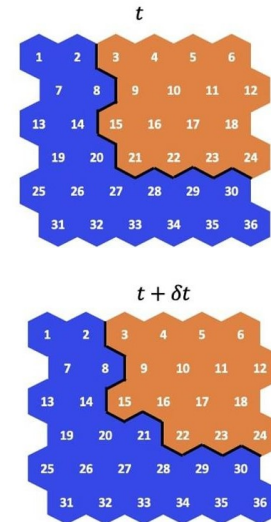
The NumPy version could be faster too



# Going further

- Crystallographic orientations
- Growth velocity (boundary mobility)
- Influence of 2<sup>nd</sup> order neighbours in formation of new boundaries

Traka, Konstantina, Karo Sedighiani, Cornelis Bos, Jesus Galan Lopez, Katja Angenendt, Dierk Raabe, and Jilt Sietsma. "Topological aspects responsible for recrystallization evolution in an IF-steel sheet—Investigation with cellular-automaton simulations." *Computational Materials Science* 198 (2021): 110643.



# Conclusions

- Array programming offers clear advantages in code clarity, size and performance
- The availability of intermediate results is an often-overlooked advantage
- Python is a widely known, easy to learn, nice little language, but NumPy is another story
- APL can be a serious contender



# What APL does great

- Powerful array programming with a few primitives
- Easy (and fast) to develop, once you learn
- High performance “out of the box”
- Dyalog (IDE, documentation, support)
- Community



# Room for improvement

- Unknown in some circles
- Introduction for domain experts
- Specialized “libraries”
- Community and ecosystem



# To sum up

- APL is great
- We need:
  - More people
  - More code



# Thank you







Elsinore 2023

# Grain Growth and Array Programming

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