

인공지능 기반 설계 이론 및 사례 연구
13차) Stage 1: Generative Design

2020년 11월

강남우

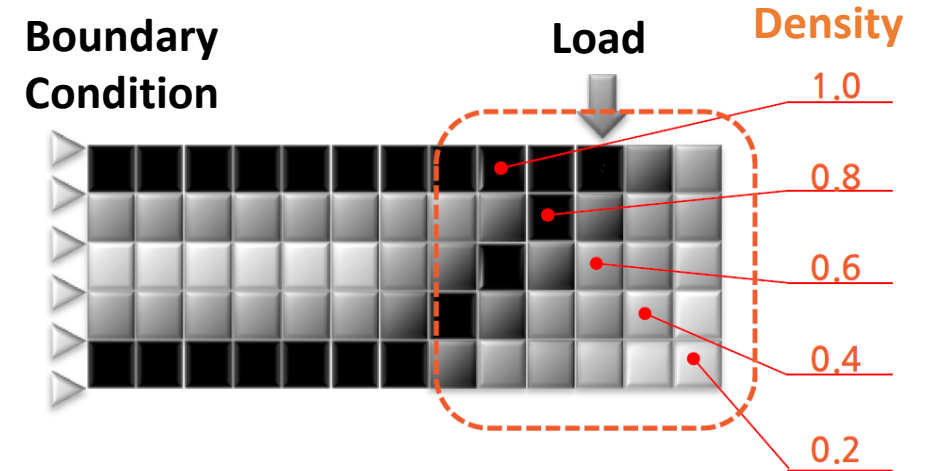
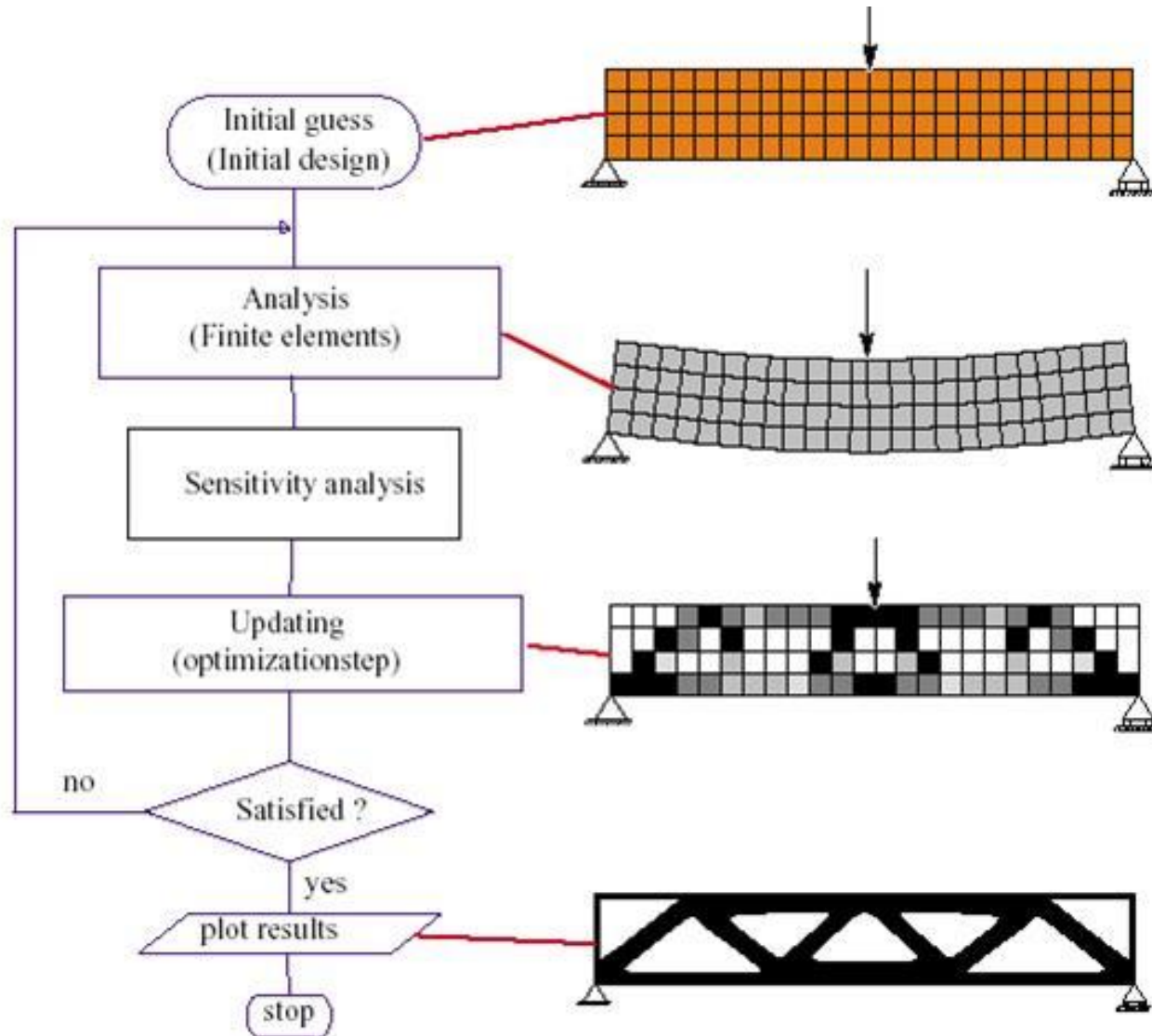
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1. Topology Optimization

2. Generative Design

위상최적화 (Topology Optimization)



- **Objective:** Minimize Compliance (=Maximize Stiffness)
- **Design Variables:** Density
- **Constraint:** Volume Fraction

❖ Beam 예제 (88line code)

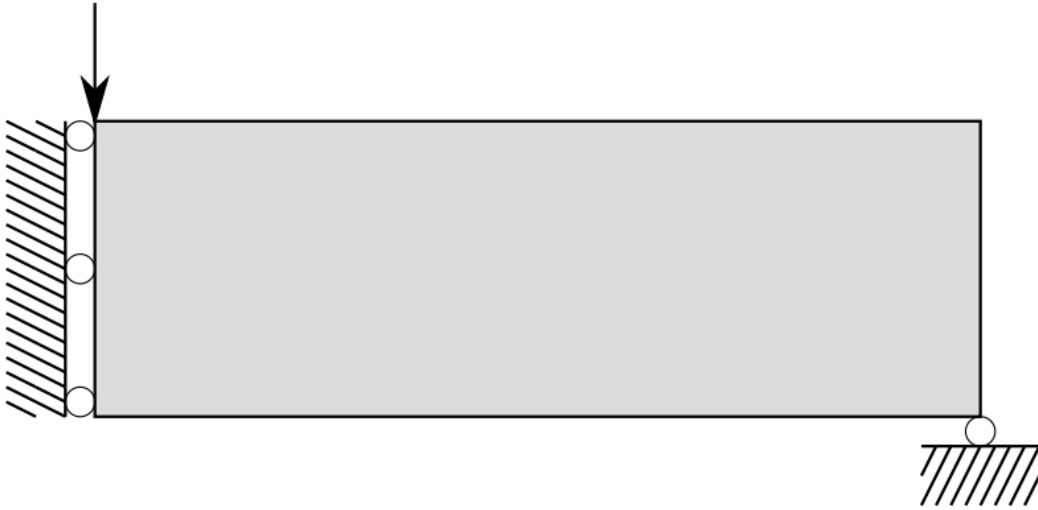


Fig. 1 The design domain, boundary conditions, and external load for the optimization of a symmetric MBB beam.

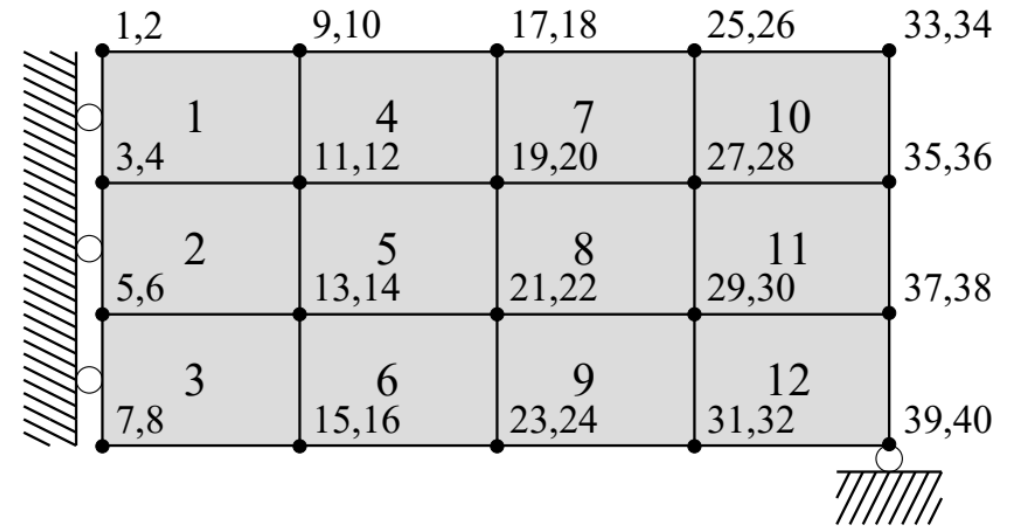


Fig. 2 The design domain with 12 elements.

❖ Optimization Problem

the total compliance (strain energy) is the summation of element-wise compliance

$$\begin{aligned} \min_{\mathbf{x}}: \quad & c(\mathbf{x}) = \mathbf{U}^T \mathbf{K} \mathbf{U} = \sum_{e=1}^N E_e(x_e) \mathbf{u}_e^T \mathbf{k}_0 \mathbf{u}_e \\ \text{subject to:} \quad & V(\mathbf{x})/V_0 = f \\ & \mathbf{K} \mathbf{U} = \mathbf{F} \\ & \mathbf{0} \leq \mathbf{x} \leq \mathbf{1} \end{aligned}$$

- c : compliance (=strain energy)
- \mathbf{U} : global displacement
- \mathbf{F} : force vectors
- \mathbf{K} : global stiffness matrix
- \mathbf{u}_e : element displacement vector,
- \mathbf{k}_0 : element stiffness matrix for an element with unit Young's modulus
- \mathbf{x} : vector of design variables (i.e. the element densities)
- N : number of elements used to discretize the design domain
- $V(\mathbf{x})$: material volume
- V_0 : design domain volume
- f : volume fraction

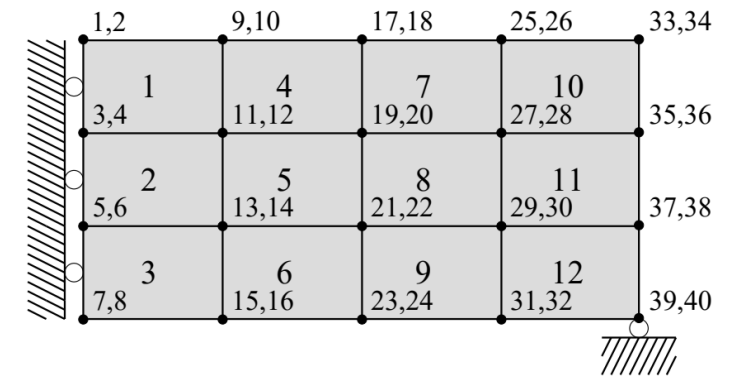


Fig. 2 The design domain with 12 elements.

❖ Modified SIMP approach

$$E_e(x_e) = E_{\min} + x_e^p(E_0 - E_{\min}), \quad x_e \in [0, 1]$$

- x_e : density
- E_e : Young's modulus
- E_0 : stiffness of the material
- E_{\min} : very small stiffness assigned to void regions
- p : penalization factor (typically $p = 3$) introduced to ensure black-and-white solutions.

❖ Sensitivities of the objective function

$$\begin{aligned} \min_{\mathbf{x}}: \quad & c(\mathbf{x}) = \mathbf{U}^T \mathbf{K} \mathbf{U} = \sum_{e=1}^N E_e(x_e) \mathbf{u}_e^T \mathbf{k}_0 \mathbf{u}_e \quad E_e(x_e) = E_{\min} + x_e^p (E_0 - E_{\min}) \\ \text{subject to:} \quad & V(\mathbf{x})/V_0 = f \\ & \mathbf{K} \mathbf{U} = \mathbf{F} \\ & 0 \leq \mathbf{x} \leq 1 \end{aligned}$$

$$\begin{aligned} \frac{\partial c}{\partial x_e} &= -p x_e^{p-1} (E_0 - E_{\min}) \mathbf{u}_e^T \mathbf{k}_0 \mathbf{u}_e \\ \frac{\partial V}{\partial x_e} &= 1 \end{aligned}$$

Assumption: each element has unit volume

Reduced Gradient Method

$$\frac{df}{d\mathbf{x}} = \frac{\partial f}{\partial \mathbf{x}} - \frac{\partial f}{\partial \mathbf{u}} \left(\frac{\partial \mathbf{h}}{\partial \mathbf{u}} \right)^{-1} \frac{\partial \mathbf{h}}{\partial \mathbf{x}},$$



$$\frac{df}{d\mathbf{x}} = -\mathbf{u}^T \frac{\partial \mathbf{K}}{\partial \mathbf{x}} \mathbf{u}.$$

❖ Optimality criteria method (OC)

$$x_e^{\text{new}} = \begin{cases} \max(0, x_e - m) & \text{if } x_e B_e^\eta \leq \max(0, x_e - m) \\ \min(1, x_e + m) & \text{if } x_e B_e^\eta \geq \min(1, x_e + m) \\ x_e B_e^\eta & \text{otherwise} \end{cases}$$

- m : positive move limit
- η ($= 1/2$): numerical damping coefficient

Optimality
condition

$$B_e = \frac{-\frac{\partial c}{\partial x_e}}{\lambda \frac{\partial V}{\partial x_e}}$$

Lagrangian multiplier λ must be chosen so that the volume constraint is satisfied

❖ Filtering

In order to ensure existence of solutions to the topology optimization problem and to avoid the formation of checkerboard patterns

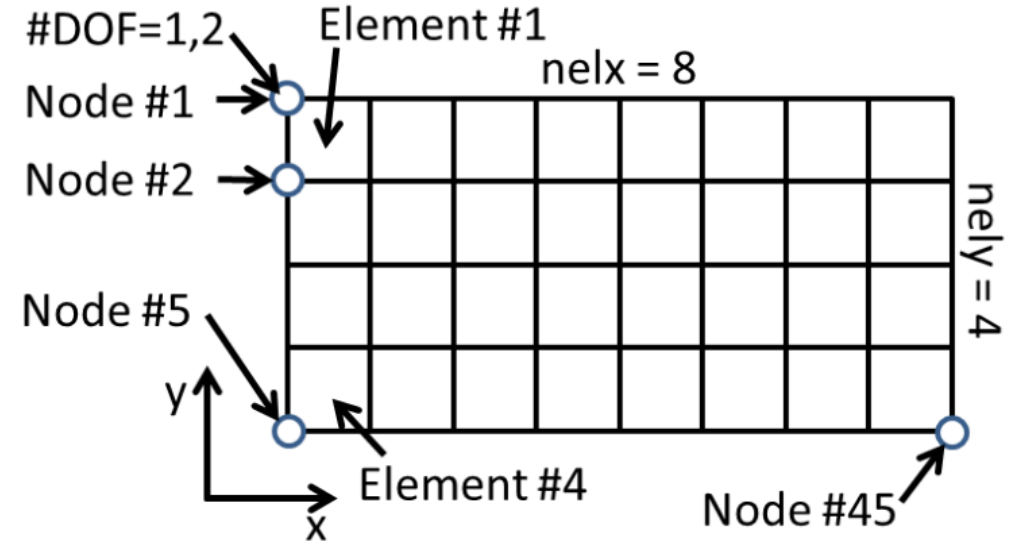
- Sensitivity filtering
$$\widehat{\frac{\partial c}{\partial x_e}} = \frac{1}{\max(\gamma, x_e) \sum_{i \in N_e} H_{ei}} \sum_{i \in N_e} H_{ei} x_i \frac{\partial c}{\partial x_i}$$

$$H_{ei} = \max(0, r_{\min} - \Delta(e, i))$$

- Density filtering
$$\tilde{x}_e = \frac{1}{\sum_{i \in N_e} H_{ei}} \sum_{i \in N_e} H_{ei} x_i$$

MATLAB Code

$$\begin{aligned} \min_{\mathbf{x}}: \quad & c(\mathbf{x}) = \mathbf{U}^T \mathbf{K} \mathbf{U} = \sum_{e=1}^N E_e(x_e) \mathbf{u}_e^T \mathbf{k}_0 \mathbf{u}_e \\ \text{subject to:} \quad & V(\mathbf{x})/V_0 = f \\ & \mathbf{K} \mathbf{U} = \mathbf{F} \\ & 0 \leq \mathbf{x} \leq 1 \end{aligned}$$



`top88(nelx,nely,volfrac,penal,rmin,ft)`

- nelx and nely: the number of elements in the horizontal and vertical direction, respectively
- volfrac: the prescribed volume fraction f ,
- penal: the penalization power p
- rmin: the filter radius rmin (divided by the element size)
- ft: specifies whether sensitivity filtering (ft = 1) or density filtering (ft = 2) should be used

$$E_e(x_e) = E_{\min} + x_e^p (E_0 - E_{\min})$$

$$H_{ei} = \max(0, r_{\min} - \Delta(e, i))$$

`top88(60,20,0.5,3,1.5,1)`

MATLAB Code

```
1  %%%% AN 88 LINE TOPOLOGY OPTIMIZATION CODE Nov, 2010 %%%%
2  function top88(nelx,nely,volfrac,penal,rmin,ft)
3  %% MATERIAL PROPERTIES
4  E0 = 1;
5  Emin = 1e-9;
6  nu = 0.3;
7  %% PREPARE FINITE ELEMENT ANALYSIS
8  A11 = [12  3 -6 -3;  3 12  3  0; -6  3 12 -3; -3  0 -3 12];
9  A12 = [-6 -3  0  3; -3 -6 -3 -6;  0 -3 -6  3;  3 -6  3 -6];
10 B11 = [-4  3 -2  9;  3 -4 -9  4; -2 -9 -4 -3;  9  4 -3 -4];
11 B12 = [ 2 -3  4 -9; -3  2  9 -2;  4  9  2  3; -9 -2  3  2];
12 KE = 1/(1-nu^2)/24*([A11 A12;A12' A11]+nu*[B11 B12;B12' B11]);
13 nodenrs = reshape(1:(1+nelx)*(1+nely),1+nely,1+nelx);
14 edofVec = reshape(2*nodenrs(1:end-1,1:end-1)+1,nelx*nely,1);
15 edofMat = repmat(edofVec,1,8)+repmat([0 1 2*nely+[2 3 0 1] -2 -1],nelx*nely,1);
16 iK = reshape(kron(edofMat,ones(8,1))',64*nelx*nely,1);
17 jK = reshape(kron(edofMat,ones(1,8))',64*nelx*nely,1);
18 % DEFINE LOADS AND SUPPORTS (HALF MBB-BEAM)
19 F = sparse(2,1,-1,2*(nely+1)*(nelx+1),1);
20 U = zeros(2*(nely+1)*(nelx+1),1);
21 fixeddofs = union([1:2:2*(nely+1)],[2*(nelx+1)*(nely+1)]);
22 alldofs = [1:2*(nely+1)*(nelx+1)];
23 freedofs = setdiff(alldofs,fixeddofs);
```

Boundary Condition & Load Vectors

MATLAB Code

```
24 %% PREPARE FILTER
25 iH = ones(nelx*nely*(2*(ceil(rmin)-1)+1)^2,1);
26 jH = ones(size(iH));
27 sH = zeros(size(iH));
28 k = 0;
29 for il = 1:nelx
30     for jl = 1:nely
31         e1 = (il-1)*nely+jl;
32         for i2 = max(il-(ceil(rmin)-1),1):min(il+(ceil(rmin)-1),nelx)
33             for j2 = max(jl-(ceil(rmin)-1),1):min(jl+(ceil(rmin)-1),nely)
34                 e2 = (i2-1)*nely+j2;
35                 k = k+1;
36                 iH(k) = e1;
37                 jH(k) = e2;
38                 sH(k) = max(0,rmin-sqrt((il-i2)^2+(jl-j2)^2));
39             end
40         end
41     end
42 end
43 H = sparse(iH,jH,sH);
44 Hs = sum(H,2);
```

```

45 %% INITIALIZE ITERATION
46 x = repmat(volfrac,nely,nelx);
47 xPhys = x;
48 loop = 0;
49 change = 1;
50 %% START ITERATION
51 while change > 0.01
52     loop = loop + 1;
53     %% FE-ANALYSIS
54     sK = reshape(KE(:)*(Emin+xPhys(:)'.^penal*(E0-Emin)),64*nelx*nely,1);
55     K = sparse(iK,jK,sK); K = (K+K')/2;
56     U(freedofs) = K(freedofs,freedofs)\F(freedofs);
57     %% OBJECTIVE FUNCTION AND SENSITIVITY ANALYSIS
58     ce = reshape(sum((U(edofMat)*KE).*U(edofMat)),2),nely,nelx);
59     c = sum(sum((Emin+xPhys.^penal*(E0-Emin)).*ce));
60     dc = -penal*(E0-Emin)*xPhys.^(penal-1).*ce;
61     dv = ones(nely,nelx);
62     %% FILTERING/MODIFICATION OF SENSITIVITIES
63     if ft == 1
64         dc(:) = H*(x(:).*dc(:))./Hs./max(1e-3,x(:));
65     elseif ft == 2
66         dc(:) = H*(dc(:)./Hs);
67         dv(:) = H*(dv(:)./Hs);
68     end
69     %% OPTIMALITY CRITERIA UPDATE OF DESIGN VARIABLES AND PHYSICAL DENSITIES
70     l1 = 0; l2 = 1e9; move = 0.2;
71     while (l2-l1)/(l1+l2) > 1e-3
72         lmid = 0.5*(l2+l1);
73         xnew = max(0,max(x-move,min(1,min(x+move,x.*sqrt(-dc./dv/lmid))));
74         if ft == 1
75             xPhys = xnew;
76         elseif ft == 2
77             xPhys(:) = (H*xnew(:))./Hs;
78         end
79         if sum(xPhys(:)) > volfrac*nelx*nely, l1 = lmid; else l2 = lmid; end
80     end
81     change = max(abs(xnew(:)-x(:)));
82     x = xnew;

```

$$\frac{\partial c}{\partial x_e} = -p x_e^{p-1} (E_0 - E_{\min}) \mathbf{u}_e^T \mathbf{k}_0 \mathbf{u}$$

$$\frac{\partial V}{\partial x_e} = 1$$

$$x_e^{\text{new}} = \begin{cases} \max(0, x_e - m) & \text{if } x_e B_e^\eta \leq \max(0, x_e - m) \\ \min(1, x_e + m) & \text{if } x_e B_e^\eta \geq \min(1, x_e + m) \\ x_e B_e^\eta & \text{otherwise} \end{cases}$$

MATLAB Code

```
83 %% PRINT RESULTS
84 fprintf(' It.:%5i Obj.:%11.4f Vol.:%7.3f ch.:%7.3f\n',loop,c, ...
85         mean(xPhys(:)),change);
86 %% PLOT DENSITIES
87 colormap(gray); imagesc(1-xPhys); caxis([0 1]); axis equal; axis off; drawnow;
```

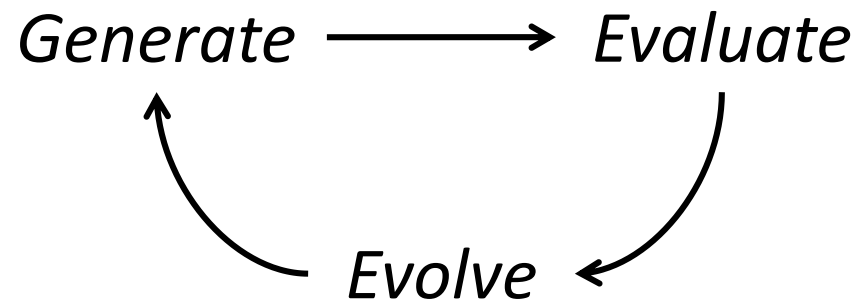
1. Topology Optimization

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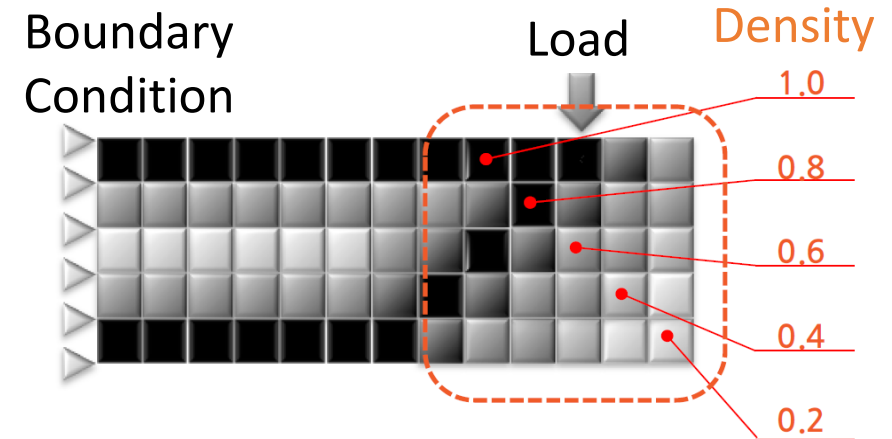
What is Generative Design?



Generative Design



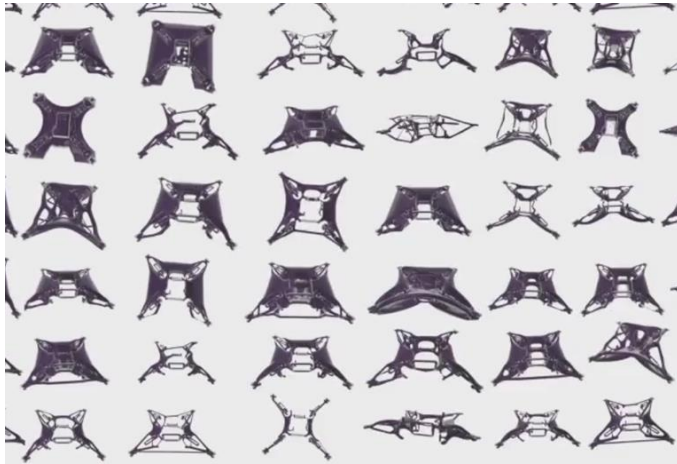
Topology Optimization



- Objective: Minimize Compliance
- Design Variables: Density
- Constraint: Volume Fraction

What if we vary
*Parameters of **Problem Definition***
in Topology Optimization?

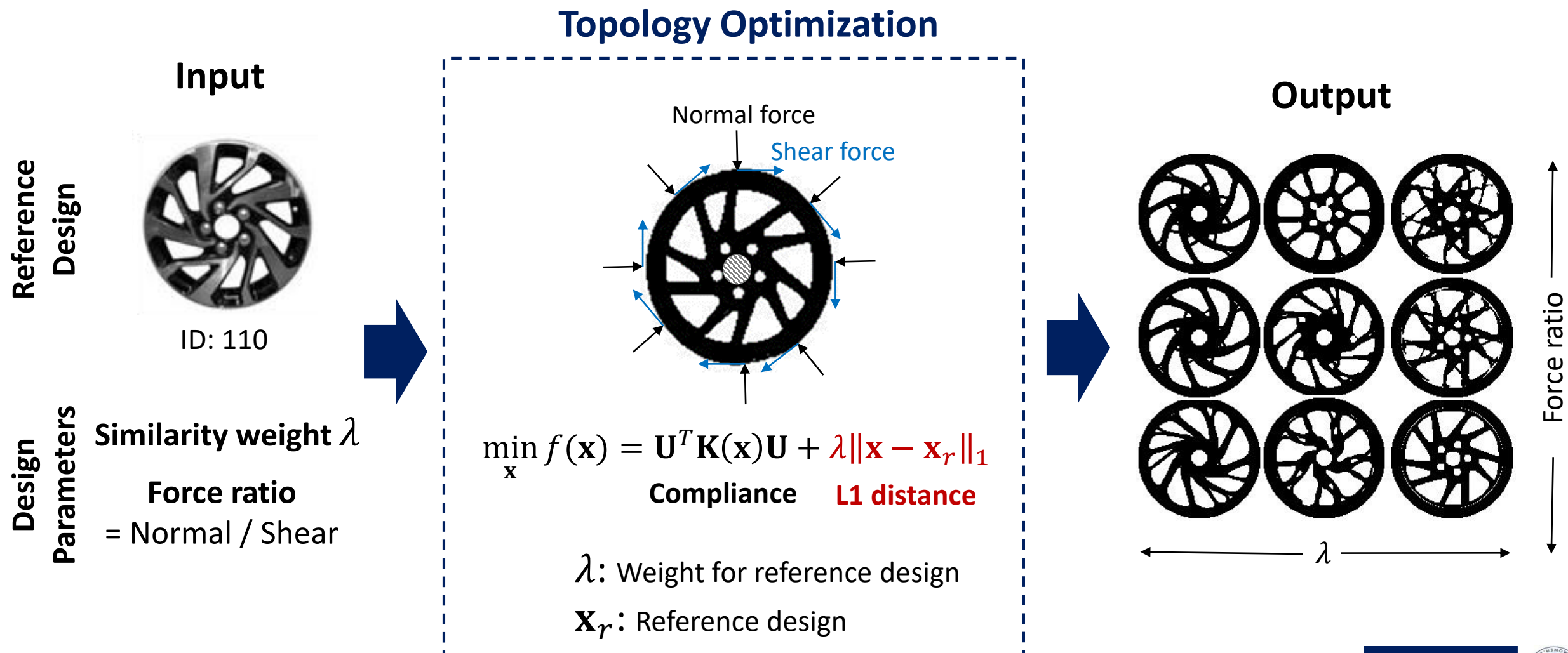
What is Generative Design?



	Generative Design	Topology Optimization	Parametric Design
Objective	Explore feasible <i>design sets</i> (<i>thousands of designs</i>)	Find the <i>optimal design</i>	Explore <i>design sets</i>
Method	Vary parameters of <i>problem definition</i> in Topology Optimization	Optimize material layout within given design space	Vary parameters of <i>geometry</i> directly

Data-driven Generative Design

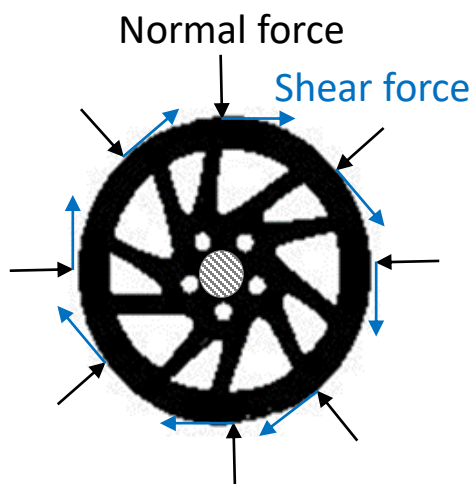
❖ Wheel Generation based on Reference Designs



Data-driven Generative Design

❖ Wheel Generation based on Reference Designs

Topology Optimization



$$\min_{\mathbf{x}} f(\mathbf{x}) = \mathbf{U}^T \mathbf{K}(\mathbf{x}) \mathbf{U} + \lambda \|\mathbf{x} - \mathbf{x}_r\|_1$$

Compliance **L1 distance**

λ : Weight for reference design

\mathbf{x}_r : Reference design

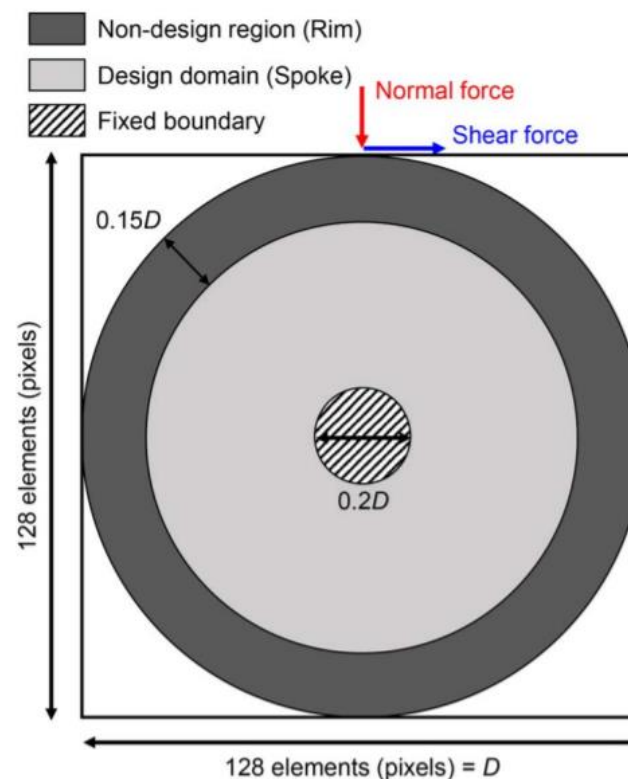


Fig. 2 Design domain and boundary conditions of a 2D wheel design

$$\frac{\partial}{\partial \mathbf{x}} (\lambda \|\mathbf{x}^* - \mathbf{x}\|_1) \cong -\lambda \mathbf{x}^*$$

$$x^* = 1 \rightarrow -\lambda$$

$$x^* = 0 \rightarrow 0$$

What Questions Do You Have?

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