

# Tree Model Reconstruction Innovization Using Multi-objective Differential Evolution

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Introduction

Related Work

Algorithm Proposed

Experimental Results

Conclusion

# Motivation

- ▶ New approach for construction of trees
  - ▶ three-dimensional spatial models,
  - ▶ in computer graphics and animation,
  - ▶ the user had to sketch basic branches.
- ▶ Our **tree reconstruction** includes:
  - ▶ *evolutionary algorithms* and
  - ▶ *procedural modelling* of trees.
- ▶ An approach (CEC'09): uses procedural models in a 2D plane, we extend it:
  - ▶ on 3D procedural models and
  - ▶ more complex trees.
- ▶ Our approach combines open-sources:
  - ▶ ecosystem framework **EcoMod** and
  - ▶ algorithm **MOjDE**  
( $\text{DEMOwSA}_{\text{CEC2007}} + \text{jDE}_{\text{TEC2006}}$ ).

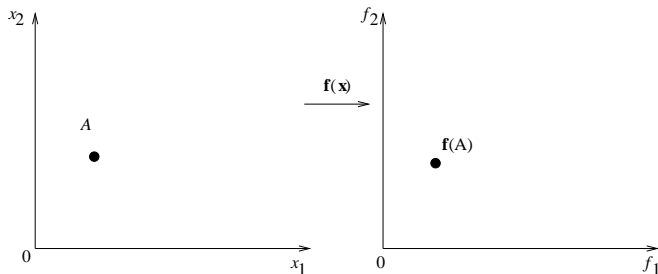


Source:  
→ CEC 2009



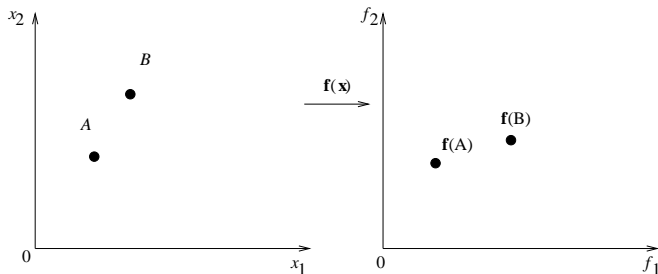
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- Numerical function, consisting multiple criteria ( $\mathbf{f}(\mathbf{x})$ ),



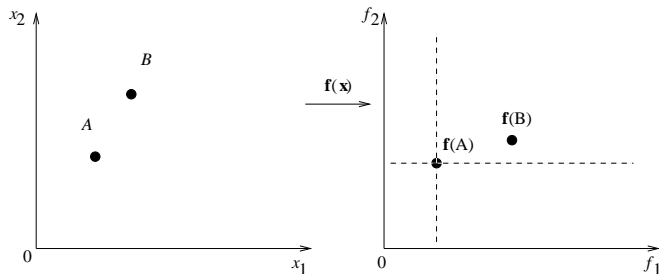
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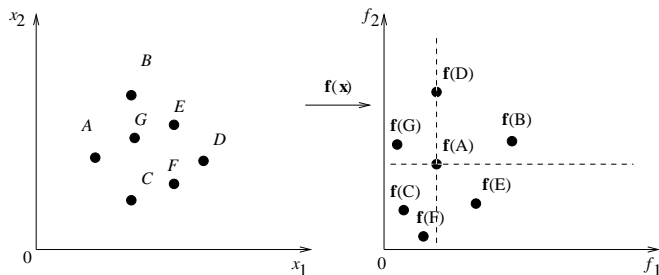
# Multi-objective optimization

- ▶ Numerical function, consisting multiple criteria ( $\mathbf{f}(\mathbf{x})$ ),
- ▶ meta-criterion ( $A \preceq B$ ): make criteria ordered by **dominance**,



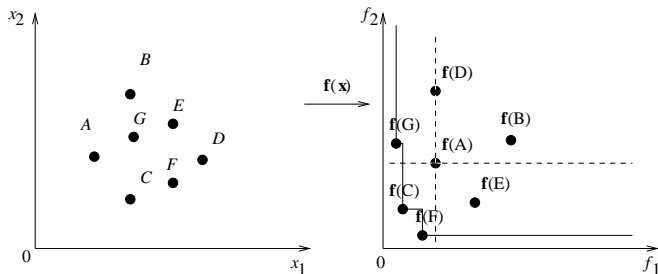
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# Multi-objective optimization

- ▶ Numerical function, consisting multiple criteria ( $f(\mathbf{x})$ ),
- ▶ meta-criterion ( $A \preceq B$ ): make criteria ordered by **dominance**,
- ▶ optimization: search for **Pareto optimal approximation set**.



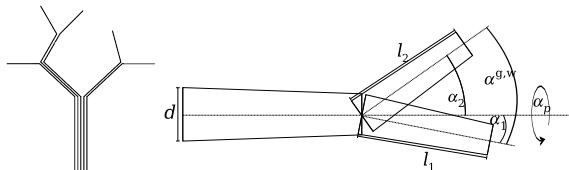


# Differential Evolution (DE)

- ▶ A floating point encoding EA for global optimization over continuous spaces,
  - ▶ trough **generations**,  
**the evolution process** improves **population of vectors**,
  - ▶ iteratively by combining a parent individual and several other individuals of the same population.
- ▶ We choose the **strategy** *jDE/rand/1/bin*
  - ▶ **mutation**:  $\mathbf{v}_{i,G+1} = \mathbf{x}_{r_1,G} + F \times (\mathbf{x}_{r_2,G} - \mathbf{x}_{r_3,G})$ ,
  - ▶ **crossover**:
$$u_{i,j,G+1} = \begin{cases} v_{i,j,G+1} & \text{if } \text{rand}(0, 1) \leq CR \text{ or } j = j_{rand} \\ x_{i,j,G} & \text{otherwise} \end{cases},$$
  - ▶ **selection**:  $\mathbf{x}_{i,G+1} = \begin{cases} \mathbf{u}_{i,G+1} & \text{if } f(\mathbf{u}_{i,G+1}) < f(\mathbf{x}_{i,G}) \\ \mathbf{x}_{i,G} & \text{otherwise} \end{cases},$
  - ▶ includes mechanism of  $F$  and  $CR$  control parameters self-adaptation.

# Problem Domain: Woody Plants Procedural Model

- ▶ 3D tree models are compactly represented using a procedural model
  - ▶ our EcoMod framework uses a numerically coded procedural model with fixed dimensionality
  - ▶ suitable for parameter estimation using DE/MOjDE.
- ▶ Parameterized procedural model builds a 3D structure of a tree and all its building parts:
  - ▶ by recursively executing a fixed procedure,
  - ▶ over a given set of numerically coded input parameters,
    - ▶ e.g. branch thickness, relative branch length and branching structure proportions.



# Trees Representable by EcoMod Framework

- ▶ Foliage or coniferous trees with very different branching structures,
- ▶ each branch and each leaf can be animated in real time to show the growth of a tree or its sway in the wind.



# 1: **procedure** branchsegment( $g, w, S_0, L_0, l_0, \mathbf{M}_0, \mathbf{M}_{m;0}^{-1}$ )

**Require:**  $g, w$  - Gravelius and Weibull index of base branch;  $S_0$  - number of strands in base branch;  $L_0, l_0$  - base branch relative and actual length;  $\mathbf{M}_0$  - base branch coordinate system;  $\mathbf{M}_{m;0}^{-1}$  - inverse matrix of rotations for gravimorphism in coordinate system for base branch; global (i.e. part of *breeder*)  $k_d, k_c, l_{type}, k_s^{g,w}, M^{g,w}, m^{g,w}, k_l^{g,w}, \alpha_m^{g,w}, \alpha^{g,w}, t, k_f, w_s, w_g$

**Ensure:** rendered tree image

- 2:  $d := k_d \sqrt{S_0}$ ; {thickness calculation from Mandelbrot}
- 3: **render** base branch( $\mathbf{M}_0, l_0, d$ );
- 4: **if**  $S_0 = 1$  **then**
- 5:     **render** leaves( $l_{type}$ ); **return**;
- 6: **end if**
- 7:  $S_1 := \lceil 1 + k_s^{g,w} (S_0 - 2) \rceil, S_2 = S_0 - S_1$ ; {number of strands in subbranches}
- 8:  $r_1 := \max \left\{ \min \left\{ \sqrt{\frac{S_1}{S_0}}, M^{g,w} \right\}, m^{g,w} \right\}$ ; {branch length proportions based on strands}
- 9:  $r_2 := \max \left\{ \min \left\{ \sqrt{\frac{S_2}{S_0}}, M^{g,w} \right\}, m^{g,w} \right\}$ ;
- 10:  $L_1 := r_1 L_0, L_2 := r_2 L_0$ ; {relative lengths of subbranches}
- 11:  $l_1 := k_l^{g,w} L_1, l_2 := k_l^{g,w} L_2$ ; {active subbranch lengths}
- 12:  $\alpha_1 := k_c \sqrt{\frac{S_2}{S_0}} \alpha^{g,w}, \alpha_2 := \alpha^{g,w} - \alpha_1$ ; {branching angles}
- 13:  $\mathbf{M}_1 := \mathbf{R}_z(\alpha_1) \mathbf{R}_y(\alpha_p) \mathbf{R}_{y \times y_m}(\alpha_m^{g,w}) \mathbf{T}_y(l_0) \mathbf{M}_0$ ; {Translation and rotation matrices}
- 14:  $\mathbf{M}_2 := \mathbf{R}_z(\alpha_2) \mathbf{R}_y(\alpha_p) \mathbf{R}_{y \times y_m}(\alpha_m^{g,w}) \mathbf{T}_y(l_0) \mathbf{M}_0$ ;
- 15:  $\mathbf{M}_{m;1}^{-1} := \mathbf{R}_{y \times y_m}(-\alpha_m^{g,w}) \mathbf{R}_y(-\alpha_p) \mathbf{R}_x(-\alpha_x(t)) \mathbf{R}_z(-\alpha_1 - \alpha_2(t)) \mathbf{M}_{m;0}^{-1}$ ; {refreshing inverse matrix for construction of gravimorphism vector}
- 16:  $\mathbf{M}_{m;2}^{-1} := \mathbf{R}_{y \times y_m}(-\alpha_m^{g,w}) \mathbf{R}_y(-\alpha_p) \mathbf{R}_x(-\alpha_x(t)) \mathbf{R}_z(-\alpha_2 - \alpha_2(t)) \mathbf{M}_{m;0}^{-1}$ ;
- 17: branchsegment( $g + 1, w + 1, S_2, L_2, l_2, \mathbf{M}_2, \mathbf{M}_{m;2}^{-1}$ ); {minor branch development}
- 18: branchsegment( $g, w + 1, S_1, L_1, l_1, \mathbf{M}_1, \mathbf{M}_{m;1}^{-1}$ ); {major branch development}
- 19: **return**; {from recursive procedure call}

# Reconstruction: Image-based Approaches to Tree Modeling

- ▶ Image-based approaches have the best potential to produce realistically looking plants
  - ▶ they rely on images of real plants.
- ▶ Little work has been done to design trees with the use of a general reconstruction from images without user interaction
  - ▶ use of sketch based guide techniques or
  - ▶ the procedural models reconstructed were two-dimensional.
- ▶ We now extended this recognition to the domain of 3D procedural models
  - ▶ suitable to model woody plants without user interaction.



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# Tree Model Reconstruction Innovization Using Multi-objective Differential Evolution

- ▶ Based on an optimization procedure with three main parts:
  - ▶ Part I: genotype encoding,
  - ▶ Part II: genotype-phenotype mapping, and
  - ▶ Part III: fitness evaluation:
    - ▶ phenotype and reference image comparison.

# Genotype Encoding

- ▶ An individual genotype vector  $\mathbf{x}$  of a DE population represents a set of procedural model parameters,
  - ▶ by computing recursive procedure using a set of parameters, EcoMod renders a tree (woody plant),
  - ▶ dimensionality of the genotype  $\mathbf{x}$  is  $D = 4509$ ,
    - ▶ where  $g \in \{0, \dots, G = 15\}$ ,  $w \in \{0, \dots, W = 50\}$ , and
    - ▶ each local  $G \times W = 750$  real-coded parameter encodes: one matrix of a Gravelius and Weibull ordered parameter for recursive calculations, and
  - ▶ all  $x_{i,j} \in [0, 1]$ ,  $i \in \{1..NP\}$  and  $j \in \{1..D\}$  are linearly normalized.



# Bounds and Scaling of Genotype-encoded Parameters $x_{i,j}$

Parameter	Formula	Interval
Number of strands in a tree (tree complexity)	$S = 400x_{i,0} + 10$	$S \in [10, 410]$
Height of base trunk	$l_0^{0,0} = 10x_{i,1}$	$l_0^{0,0} \in [0 \text{ m}, 10 \text{ m}]$
Coefficient of branch thickness	$k_d = 0.05x_{i,2}$	$k_d \in [0, 0.05]$
Phyllotaxis angle	$\alpha_p = 360x_{i,3}$	$\alpha_p \in [0^\circ, 360^\circ]$
Branching ratio of subbranch strands distribution	$k_s^{g,w} = 0.5x_{i,j} + 0.5, \forall j \in [4, 753]$	$k_s^{g,w} \in [\frac{1}{2}, 1]$
Branching angle between dividing sub-branches	$\alpha^{g,w} = 180x_{i,j} \forall j \in [754, 1503]$	$\alpha^{g,w} \in 0^\circ, 180^\circ$
Maximum relative sub-branch to base branch length	$M^{g,w} = 20x_{i,j} \forall j \in [1504, 2253]$	$M^{g,w} \in [0, 20]$
Minimum relative sub-branch to base branch length	$m^{g,w} = 20x_{i,j} \forall j \in [2254, 3003]$	$m^{g,w} \in [0, 20]$
Branch length scaling factor	$k_l^{g,w} = 20x_{i,j}, \forall j \in [3004, 3753]$	$k_l^{g,w} \in [0, 20]$
Gravicentrism impact	$k_c = x_{i,3754}$	$k_c \in [0, 1]$
Gravimorphism impact (i.e. gravitational bending of branches)	$\alpha_m^{g,w} = 360x_{i,j} - 180, \forall j \in [3755, 4504]$	$\alpha_m^{g,w} \in [-180^\circ, 180^\circ]$
Enabling leaves display on a tree	$B_l = x_{i,4505} < 0.5 ? 0 : 1$	$B_l \in \{0, 1\}$
Size of leaves	$l_l = 0.3x_{i,4506}$	$l_l \in [0, 0.3]$
Density of leaves	$\rho_l = 30x_{i,4507}$	$\rho_l \in [0, 30]$
Leaf distribution type	$l_{type} = 5x_{i,4508}$	<i>Spiral, Stacked, Staggered, Bunched, or Coniferous</i> <sup>1</sup>

# Genotype-phenotype Mapping

- ▶ Reconstruction method is based on reconstruction of two-dimensional images of woody plants  $\mathbf{z}^*$  (photo),
- ▶ to compare the three-dimensional tree evolved with the use of genotype  $\mathbf{x}$  to the reference image  $\mathbf{z}^*$ , genotype  $\mathbf{x}$  must be transformed to its phenotype first,
  - ▶ phenotype is a rendered two-dimensional image  $\mathbf{z}$ ,
  - ▶ images  $\mathbf{z}^*$  and  $\mathbf{z}$  are all of dimensionality  $X \times Y$  pixels,
    - ▶ the reference image is scaled to the given resolution, if necessary.
  - ▶ both images are converted to black and white, where white (0) pixels mark *background* and black (1) pixels mark *material*, e.g. wood,
- ▶ An evolved procedural model is rendered for comparison twice
  - ▶ to favor three-dimensional procedural models generation,
  - ▶ projections differ by  $\beta = 90^\circ$  camera view angle along the trunk base (i.e.  $z$  axis for OpenGL).



# Phenotype and Reference Image Comparison

- ▶ The recognition success is measured by similarity of
  - ▶ the reference original images (2D) and
  - ▶ the rendered image (2D) projections of evolved parametrized procedural models.
- ▶ Images are compared pixel-wise by **two criteria**:
  1. in the evolved image, for each pixel rendered as material (1):
    - ▶ the Manhattan distance to the nearest material pixel in the reference image is computed
    - ▶ and vice-versa, for each material (1) pixel of an evolved model image,
  2. count of differing pixels (0/1) among comparing images.

# Calculus of the Two Criteria for Image Comparison

- Fitness evaluation of each phenotype are the sums:

$$f_1(\mathbf{x}) = f(\mathbf{g}(\mathbf{x}, \beta_1), \mathbf{g}(\mathbf{x}, \beta_2)) = h_1(\mathbf{z}_1) + h_1(\mathbf{z}_2)$$

$$h_1(\mathbf{z}_i) = \sum_{x,y} m_1(z_{x,y}^i, z_{x,y}^*) + \sum_{x,y} m_1(z_{x,y}^*, z_{x,y}^i)$$

$$f_2(\mathbf{x}) = f(\mathbf{g}(\mathbf{x}, \beta_1), \mathbf{g}(\mathbf{x}, \beta_2)) = h_2(\mathbf{z}_1) + h_2(\mathbf{z}_2)$$

$$h_2(\mathbf{z}_i) = \sum_{x,y} w(z_{x,y}^i, z_{x,y}^*) + \sum_{x,y} w(z_{x,y}^*, z_{x,y}^i),$$

- where  $i \in \{1, 2\}$  for the two orthogonal evolved projections,
- $m_1$  denotes a function computing Manhattan distance to the nearest material (value 1) pixel in the image  $\mathbf{z}^*$ , and
- $w$  denotes a function to count differing pixels in same coordinates.

# 1: procedure MO reconstruction( $\mathbf{z}^*$ )

**Require:**  $S_0$  - maximum number of strands in base branch; also, other default parameters for MOjDE and EcoMod

**Ensure:** Pareto set of reconstructed parameterized procedural 3D woody plant models

2: uniform randomly generate DE initial population  $\mathbf{x}_{i,0} \in [0, 1]$  for  $i = 1..NP$ ;

3: for DE generation loop  $g$  (while FEs < 10000) do

4:   for DE iteration loop  $i$  (for all individuals  $\mathbf{x}_{i,g}$  of a population) do

5:     DE individual  $\mathbf{x}_{i,g}$  creation (adaptation, mutation, crossover):

6:         
$$F_{i,G+1} = \begin{cases} F_i + rand_1 \times F_u & \text{if } rand_2 < \tau_1, \\ F_{i,G} & \text{otherwise} \end{cases}; CR_{i,G+1} = \begin{cases} rand_3 & \text{if } rand_4 < \tau_2, \\ CR_{i,G} & \text{otherwise} \end{cases};$$

8:          $\mathbf{v}_{i,G+1} = \mathbf{x}_{r_1,G} + F_{i,G+1}(\mathbf{x}_{r_2,G} - \mathbf{x}_{r_3,G});$

9:         
$$u_{i,j,G+1} = \begin{cases} v_{i,j,G+1} & \text{if } rand(0, 1) \leq CR_{i,G+1} \text{ or } j = j_{rand}, \\ x_{i,j,G} & \text{otherwise} \end{cases};$$

10:     DE fitness evaluation (genotype-phenotype mapping, rendering, and comparison):

11:          $\mathbf{z}_1 = \mathbf{g}(\mathbf{u}_{i,g}, \beta_1), \mathbf{z}_2 = \mathbf{g}(\mathbf{u}_{i,g}, \beta_2)$  {Execute Algorithm branchsegment twice}

12:          $h_1(\mathbf{z}_1) = \sum_{x,y} m_1(z_{x,y}^1, z_{x,y}^*) + \sum_{x,y} m_1(z_{x,y}^*, z_{x,y}^1)$ ; {First difference metric, at  $0^\circ$ }

13:          $h_1(\mathbf{z}_2) = \sum_{x,y} m_1(z_{x,y}^2, z_{x,y}^*) + \sum_{x,y} m_1(z_{x,y}^*, z_{x,y}^2)$ ; {First difference metric, at  $90^\circ$ }

14:          $f_1(\mathbf{x}) = f(\mathbf{g}(\mathbf{x}, \beta_1), \mathbf{g}(\mathbf{x}, \beta_2)) = h_1(\mathbf{z}_1) + h_1(\mathbf{z}_2)$ ; {Fitness evaluation, 1st criterion}

15:          $h_2(\mathbf{z}_1) = \sum_{x,y} w(z_{x,y}^1, z_{x,y}^*) + \sum_{x,y} w(z_{x,y}^*, z_{x,y}^1)$ ; {Second difference metric,  $0^\circ$ }

16:          $h_2(\mathbf{z}_2) = \sum_{x,y} w(z_{x,y}^2, z_{x,y}^*) + \sum_{x,y} w(z_{x,y}^*, z_{x,y}^2)$ ; {Second difference metric,  $90^\circ$ }

17:          $f_2(\mathbf{x}) = f(\mathbf{g}(\mathbf{x}, \beta_1), \mathbf{g}(\mathbf{x}, \beta_2)) = h_2(\mathbf{z}_1) + h_2(\mathbf{z}_2)$ ; {Fitness evaluation, 2nd criterion}

18:          $\mathbf{f}(\mathbf{x}) = \{f_1(\mathbf{x}), f_2(\mathbf{x})\}$ ; {Fitness evaluation, all criteria combined done}

19:     DE selection:

20:         
$$\mathbf{x}_{i,G+1} = \begin{cases} \mathbf{u}_{i,G+1} & \text{if } \mathbf{f}(\mathbf{u}_{i,G+1}) \preceq \mathbf{f}(\mathbf{x}_{i,G}) \\ \mathbf{x}_{i,G} & \text{otherwise} \end{cases};$$
 {Multi-objective comparison operator}

21:         if not  $(\mathbf{u}_{i,G+1} \preceq \mathbf{x}_{i,G} \text{ or } \mathbf{x}_{i,G} \preceq \mathbf{u}_{i,G+1})$  then add  $\mathbf{u}_{i,G+1}$  to population archive;

22:     end for

23:     Truncate DE population archive to a size of  $NP$  using SPEA2 mechanism.

24: end for

25: return the best individuals obtained;

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# Experiment Design

- ▶ Sampling rate dimension of the rendered parametrized procedural model set to  $250 \times 250$ ,
  - ▶ the maximal number of strands in the tree  $S = 410$ ,
  - ▶ size of MOjDE population (approximation set)  $NP = 100$ ,
  - ▶ maximal number of fitness evaluations (FEs)  $FEs = 10,000$ ,
  - ▶ experiment repeat for  $N_{runs} = 30$  independent runs,
  - ▶ the remaining parameters were kept default as in original algorithms from their literature.
- 
- ▶ Reference model to reconstruct:

(rendered in  
EcoMod)



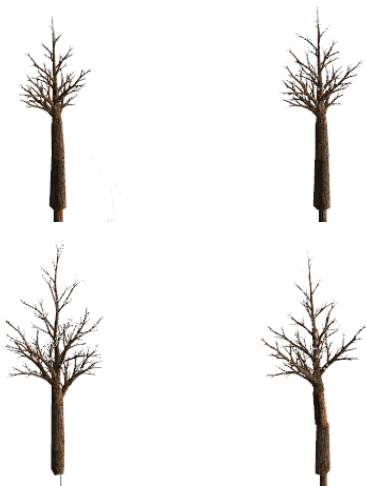
# Evolution of a Sample Procedural Model



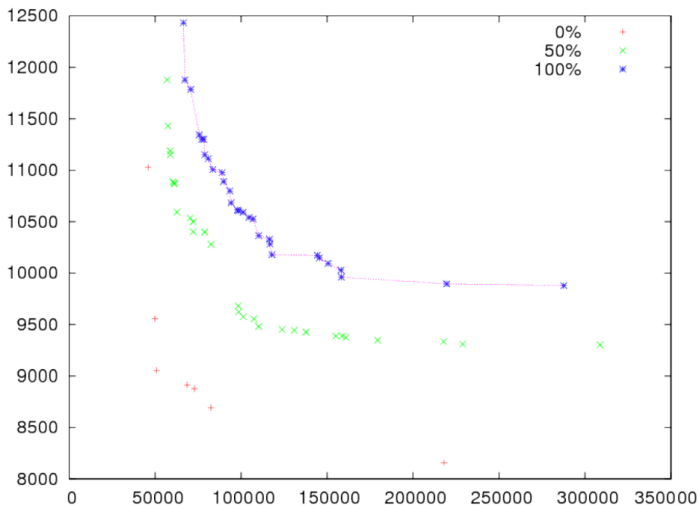
- ▶ Evolved projection image samples for first run ( $RN_0 = 1$ ):
  - ▶ Pareto optimal individuals of a current population, through generations.



# Rendered evolved models of final approximation set (run 1)



## Attainment Surfaces: 0%, 50%, and 100% for $NP=100$ .



# Summary Conclusions

- ▶ An approach to design woody plant geometrical models,
- ▶ parameters of the procedural model are iteratively evolved using multi-objective differential evolution MOjDE algorithm
  - ▶ sampled randomly to reconstruct geometrical models,
- ▶ procedural models are rendered using EcoMod framework,
- ▶ rendered images are compared to the reference source images, for reconstruction, to guide the optimization process.
  - ▶ fitness is evaluated by two criteria, which are not pre-weighted,
  - ▶ multi-objective optimization finds multiple criteria trade-offs.
- ▶ Reconstruction approach results assessment
  - ▶ attainment surfaces (trade-offs distribution),
  - ▶ renderings of sample evolved models,
  - ▶ rendered final approximation set models.

# Future Guidelines

- ▶ Prepare a combined approach of bi-criteria into single criteria by some sort of tradeoff,
- ▶ analyze assigned criteria tradeoff weights,
- ▶ segmentation procedures for automatic image alignment and base trunk recognition from natural landscape photography,
- ▶ parallel reconstruction of trees (LIDAR scanned forest),
- ▶ use additional metric: volumetric data instead of black and white images,
- ▶ implement the application for mobile devices (+AR), and
- ▶ add interactive methods for optimization.

Thank you very much for listening!  
YOU are welcome to join this research.



Also, greetings from Slovenia by A. Zamuda.

Questions and suggestions?