

# Numerical Programming Final Project

## Combined Presentation (Tasks 1–3)

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### Problem statement

Simulate an illuminated drone show with shape preservation:

1. Static formation on a handwritten input.
2. Transition to “Happy New Year!” greeting.
3. Dynamic tracking of a moving object from video.

## 1 Task 1: Static Formation on a Handwritten Input

### Task statement

**Input:** handwritten name image (at least 8 characters), number of drones  $N$ , still initial configuration.

**Goal:** move the swarm from the still initial formation and form the handwritten name.

**Output:** trajectories and visualization.

#### 1.1 Overview of the pipeline

- **Target point extraction** from `task1/inputs/name.png` into `task1/outputs/target_points.csv`.
- **Trajectory generation** using swarm IVP with repulsion in `task1/simulate_drones.py`.
- **Visualization:** static trajectory plot and an animation (GIF).

#### 1.2 Input data and preprocessing

Binarization uses `THRESH_BINARY_INV` so ink becomes white and background black. Targets are sampled from contour / interior / skeleton, with optional minimum spacing. The number of drones  $N$  equals the number of sampled target points and is kept consistent across tasks.

#### 1.3 Inputs and parameters

- **Input image:** `task1/inputs/name.png` (handwritten name, at least 8 characters).
- **Number of drones:**  $N$  (example: 100).
- **Initial formation:** line or two-line placement below the target (options: `hline_below`, `hline_match_targets_x`, `two_hlines_below`).
- **Repulsion parameters:** gain  $k_{\text{rep}}$  and safety radius  $R_{\text{safe}}$  for collision avoidance.

## 1.4 Mathematical model

For drone  $i$ :

$$x_i(t) \in \mathbb{R}^2, \quad v_i(t) \in \mathbb{R}^2.$$

**Swarm IVP with repulsion:**

$$\dot{x}_i(t) = v_i(t), \tag{1}$$

$$\dot{v}_i(t) = \frac{1}{m} \left( k_p (T_i - x_i(t)) + \sum_{j \neq i} f_{\text{rep}}(x_i, x_j) - k_d v_i(t) \right). \tag{2}$$

**BVP (shooting, optional):**

$$x_i(0) = x_{i,0}, \quad x_i(T) = T_i.$$

## 1.5 Numerical methods

We use `solve_ivp` (RK45) for the coupled IVP. BVP shooting is available per drone without repulsion. Velocity saturation is applied as:

$$\dot{x}_i = v_i \cdot \min \left( 1, \frac{v_{\max}}{\|v_i\|} \right).$$

## 1.6 Validation

We validate final formation accuracy and safety:

- **Accuracy:** mean/max distance to targets at  $t = T$ .
- **Safety:** minimum inter-drone distance over time (diagnostic).

## 1.7 Reproducibility (commands)

```
python3 extract_target_points.py \  
  --image task1/inputs/name.png \  
  --n 200 --mode skeleton --min-target-spacing 5 \  
  --out-dir task1/outputs --debug-png --debug-point-radius 2
```

```
python3 task1/simulate_drones.py \  
  --model swarm --k-rep 160 --r-safe 50 \  
  --k-p 2.0 --k-d 2.5 --v-max 1e9 \  
  --t-end 12 --steps 120 \  
  --save-gif --save-traj-csv --save-traj-npy --save-traj-plot \  
  --drone-size 21 --target-size 35 --initial-size 21
```

## 1.8 Test cases

**Works well:** clear, high-contrast handwriting; moderate  $N$  with spacing; tuned  $R_{\text{safe}}$ .

**Does not work well:** low-contrast inputs; very large  $N$  with small  $R_{\text{safe}}$ ; shooting without repulsion.

## 2 Task 2: Transition to “Happy New Year!”

### Task statement

**Input:** swarm at Task 1 formation and greeting text.

**Goal:** move from handwritten name to greeting formation.

**Output:** trajectories and visualization.

### 2.1 Overview

Start positions are `task1/outputs/target_points.csv`. Greeting targets are extracted via `extract_target_points.py`. Transition trajectories are generated in `task2/transition.py` using BVP shooting or swarm IVP.

### 2.2 Mathematical model

Same second-order point-mass model as Task 1 with fixed targets  $T_i$  for the greeting.

$$x_i(0) = x_{i,0}, \quad x_i(T) = T_i.$$

### 2.3 Reproducibility (commands)

```
python3 extract_target_points.py \  
--image task2/inputs/greeting.png \  
--n 200 --mode skeleton --min-target-spacing 5 \  
--out-dir task2/outputs --debug-png
```

```
python3 task2/transition.py \  
--start task1/outputs/target_points.csv \  
--targets task2/outputs/target_points.csv \  
--bg-target task2/inputs/greeting.png \  
--model swarm --k-rep 160 --r-safe 50 \  
--k-p 2.0 --k-d 2.5 --v-max 1e9 \  
--t-end 12 --steps 120 \  
--collision-report --collision-threshold 50 \  
--save-gif --save-traj-csv --save-traj-npy --save-traj-plot \  
--full-view --output-prefix full_transition
```

```
python3 task2/transition.py \  
--start task1/outputs/target_points.csv \  
--targets task2/outputs/target_points.csv \  
--bg-target task2/inputs/greeting.png \  
--model swarm --k-rep 160 --r-safe 50 \  
--k-p 2.0 --k-d 2.5 --v-max 1e9 \  
--t-end 12 --steps 120 \  
--collision-report --collision-threshold 50 \  
--save-gif --save-traj-csv --save-traj-npy --save-traj-plot \  
--output-prefix transition
```

### 2.4 Test cases

**Works well:** same  $N$  for start/target; swarm IVP with repulsion; skeleton extraction for greeting.

**Does not work well:** mismatched  $N$ ; shooting without repulsion in dense cases; very small  $R_{\text{safe}}$ .

### 3 Task 3: Dynamic Tracking and Shape Preservation

#### Task statement

**Input:** swarm at Task 2 greeting and a video.

**Goal:** track a moving object with shape preservation.

**Output:** trajectories and visualization.

#### 3.1 Overview

We segment the moving object, extract its contour per frame, sample  $N$  boundary points, and maintain stable correspondence across frames. The implementation is in `task3/dynamic_tracking.py`.

#### 3.2 Contour targets

For each frame  $k$ , sample  $N$  points at equal arc-length:

$$T_i^{(k)} = C_k(s_i), \quad s_i = \frac{i}{N} L(C_k).$$

We align consecutive frames via cyclic shift and orientation checks to keep point correspondence stable.

#### 3.3 Controllers

- **Direct (kinematic):**  $x_i^{(k)} = T_i^{(k)}$  (zero tracking error).
- **Dynamics (IVP RK4):** second-order damped model integrated by RK4 to follow moving targets.

#### 3.4 Dynamic model (IVP)

For each drone in dynamics mode:

$$\dot{x}_i(t) = v_i(t), \tag{3}$$

$$\dot{v}_i(t) = \frac{1}{m} (k_p (T_i(t) - x_i(t)) - k_d v_i(t)), \tag{4}$$

with velocity saturation  $v_i \leftarrow v_i \cdot \min\left(1, \frac{v_{\max}}{\|v_i\|}\right)$  before integration.

#### 3.5 Reproducibility (commands)

```
python3 task3/dynamic_tracking.py \  
--tracking-mode contour \  
--controller direct \  
--video-step 1 \  
--contour-upscale 3.0 --contour-smooth 9 \  
--segmenter greenscreen \  
--save-gif --save-traj-csv --save-traj-npy --gif-fps 30 \  
--output-gif task3/outputs/task3_contour_direct_hi_with_video.gif \  
--drone-size 13
```

#### 3.6 Test cases

**Works well:** green-screen video; direct controller; higher contour upscaling.

**Does not work well:** complex backgrounds; dynamics controller at high speed (lag); very small  $N$ .

## AI usage disclosure

This project was developed with AI assistance (ChatGPT) for explanation, implementation, and debugging support.

## Files included in the submission

- Code: `extract_target_points.py`, `task1/simulate_drones.py`, `task2/transition.py`, `task3/dynamic_tracking.py`
- Inputs: `task1/inputs/name.png`, `task2/inputs/greeting.png`, `task3/video.mp4`
- Outputs: `task1/outputs/`, `task2/outputs/`, `task3/outputs/`
- Presentation (this file): `report/presentation.pdf`