

# OptiView Pro: UX/UI Handout

## Linked Dual-Space Visualization for 4-Objective Pareto Fronts with ML-Driven Design Space Exploration

Final Integrated Solution  
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### 1 Overview & Core Design Philosophy

OptiView Pro supports decision-making in a 4-objective multi-objective optimization (MOO) problem where three objectives are continuous ( $f_1, f_2, f_3$ ) and one objective is binary/discrete ( $f_4 \in \{0, 1\}$ ). In addition, users explore the *design/feature space* before objectives: a 2D input plane  $(x_1, x_2)$  with a PyTorch binary classifier producing a probability field  $P(\text{class} = 1 \mid x_1, x_2)$  and a decision boundary at 0.5.

**Design principle: Linked Dual-Space Understanding.** The interface connects *causes* (design space, ML predictions) to *effects* (objective space, Pareto optimality) through tightly linked views, enabling intuitive discovery and confident selection.

**Pareto emphasis principle.** Pareto points are always visually dominant via size/opacity/outline separation, while still grounded in the full feasible set.

**Clutter control refinement.** All cross-view synchronization remains on hover/selection, while *optional linking-line cues* can be toggled on/off to keep the workspace clean.

### 2 Data & System Context

- **Dataset A (Feasible set):** all feasible points with fields  $(x_1, x_2, f_1, f_2, f_3, f_4)$ ; potentially large.
- **Dataset B (Pareto set):** nondominated subset with the same fields.
- **ML model output:** probability grid  $Z(x_1, x_2) = P(\text{class} = 1 \mid x_1, x_2)$ , derived from a 2D mesh and PyTorch forward pass.

### 3 Primary User Goals

1. Identify Pareto-optimal solutions quickly.
2. Understand trade-offs among  $f_1, f_2, f_3$  and how  $f_4$  changes the front.
3. Relate feature-space regions to objective-space performance.
4. Filter, shortlist, and compare candidates for final decisions.

## 4 Dashboard Layout (Final)

The UI uses a split dashboard with linked panels:

- **Top-Left: Design Space (Input & ML).** 2D probability heatmap on  $(x_1, x_2)$  with decision boundary and point overlays.
- **Top-Right: Objective Space (Pareto & Feasible Context).** Two faceted 3D scatter views split by  $f_4$ : *Panel A*:  $f_4 = 0$ , *Panel B*:  $f_4 = 1$ .
- **Bottom-Left (Support): Parallel Coordinates and/or SPLOM.** Multi-dimensional trade-off reading and brushing across all four objectives.
- **Bottom-Right: Details & Inspector.** Exact values for hovered/selected points and pinned comparison table.
- **Right Sidebar: Controls & Filters.** Global view toggles, opacity sliders, range filters, and linking controls.

## 5 Panel Encodings & Interactions

### 5.1 Design Space Panel (2D Heatmap)

**Visual encoding:**

- Heatmap color  $\rightarrow Z(x_1, x_2)$  (probability of class 1).
- Decision boundary  $\rightarrow$  thick contour at  $Z = 0.5$  (plus optional contours at 0.1, 0.9).
- Feasible points  $\rightarrow$  tiny, low-opacity dots.
- Pareto points  $\rightarrow$  larger, saturated markers with outline/glow.
- Binary objective  $f_4 \rightarrow$  marker *shape* (e.g., circle vs triangle) or stroke style.

**Key interactions:**

- Hover a point  $\Rightarrow$  highlight same point in objective space and tradeoff views.
- Lasso/box select region  $\Rightarrow$  create a working set linked to all views.
- Probe cursor on grid  $\Rightarrow$  show  $Z(x_1, x_2)$  and nearest point objectives (optional).
- Zoom/pan for local inspection.

### 5.2 Objective Space Panel (Faceted 3D)

**Visual encoding:**

- Axes:  $(f_1, f_2, f_3)$  in 3D.
- Feasible cloud  $\rightarrow$  very faint points or density volumes.
- Pareto front  $\rightarrow$  bold points on top (size + glow).
- Faceting by  $f_4 \rightarrow$  two side-by-side panels for direct comparison.

**Key interactions:**

- Rotate/pan/zoom 3D.
- Hover a Pareto point  $\Rightarrow$  ring highlight at its  $(x_1, x_2)$  in heatmap.
- Brush-select in 3D  $\Rightarrow$  filters heatmap and tradeoff views.

### 5.3 Tradeoff View (Parallel Coordinates / SPLOM)

**Purpose:** reduce 3D occlusion and enable precise multi-criteria filtering.

**Encoding:**

- One axis per objective  $f_1, f_2, f_3, f_4$  (binary shown as two categorical levels).
- Feasible set  $\rightarrow$  thin, ghosted lines / density bands.
- Pareto set  $\rightarrow$  thicker, saturated lines.

**Interactions:**

- Range brushing on any axis  $\Rightarrow$  linked filtering everywhere.
- Axis reordering to reveal correlations.

### 5.4 Inspector Panel

- Shows  $(x_1, x_2)$ ,  $Z(x_1, x_2)$ , predicted class, and  $(f_1, f_2, f_3, f_4)$ .
- **Pin/Compare:** users can pin multiple candidates and compare deltas.

### 5.5 Controls & Filters Sidebar

**Core controls:**

- Toggle: show/hide feasible cloud.
- Slider: feasible cloud opacity.
- Toggle: density vs sampled feasible points.
- Filters: objective range sliders and/or categorical  $f_4$  filter.
- Toggle: split/combined view for  $f_4$  (default split).
- **Toggle: show linking line.** When off, highlighting remains but lines are hidden.
- Toggle: show sampled points on heatmap.

## 6 Pareto Emphasis Rules (Always-On)

- Opacity separation: feasible  $\alpha \approx 0.05$ – $0.15$ ; Pareto  $\alpha = 1$ .
- Size separation: feasible 1–2 px; Pareto 6–8 px.
- Edge/glow: Pareto only.
- Draw order: Pareto always on top.
- Persistent legends clarifying feasible vs Pareto layers.

## 7 Scalability & Performance

- If Dataset A is massive, default to density rendering and progressive refinement.
- Pareto set is never sampled.
- Cross-view linking updates only the highlight layer for responsiveness.

## 8 Accessibility & Clarity

- Do not encode  $f_4$  by color alone (use shape/stroke).
- Heatmap palette chosen for contrast and grayscale legibility.
- Axis direction (min/max) indicated with small arrows or labels.

## 9 Future Enhancements

- Knee-point detection and labeling in both spaces.
- Session save/load and export of views/selected solutions.
- What-if sliders for  $(x_1, x_2)$  with live ML and objective preview.
- Uncertainty visualization (ML confidence or Pareto robustness).