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Optimised multi objective design of low-head diversion structures

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Keywords: DAMS, BARRAGES & RESERVOIRS, DESIGN METHODS & AIDS, CUT-OFF WALLS & BARRIERS, WATER MANAGEMENT IN DEVELOPING COUNTRIES, MUNICIPAL & PUBLIC SERVICE ENGINEERING

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16 **Abstract**

17 Diversion head works, also called weirs or barrages, are structures constructed across rivers or canals
18 to store water or raise the water level. Design of diversion structures involves calculating the depth,
19 length and thickness parameters of its horizontal and sloping aprons and cut-off piles. These are set to
20 achieve structural stability against multiple forms of failure such as scour, uplift, sliding, piping and
21 overturning. The components of a diversion structure, which have complex non-linear relationships, are
22 traditionally calculated with empirical recommendations derived from However, currently practiced
23 design approaches such as Khosla's method of independent variable do not explore the trade-offs
24 between the many relevant design objectives, failing to reveal possibly superior designs with less cost.
25 In this article we propose a multi objective optimisation design approach for diversion structures and
26 forward a free and open source code. The method is demonstrated on a stylised design problem. The
27 results show substantial improvement in stability and cost of the structure.

28 **Key words:** Modelling; Structural design; Municipal & public service engineering;
29 Environmental engineering; Dams, barrages & reservoirs; Design methods & aids;
30 Buildings, structures & design; Hydrology & water resource; Cut-off walls & barriers;
31 Developing countries

32 **Key Points**

- 33 1. We propose a design formulation for multi objective optimization of low-head diversion structures
- 34 2. Current design approaches for diversion structures on permeable media do not explore the trade-
35 offs between the many relevant design objectives, failing to reveal possibly superior designs
- 36 3. Multi objective optimization achieves substantial improvement in stability of the structure and cost.

37

38 **1. Introduction**

39 Costly expansion of infrastructure is needed to meet increasing water, food and energy demands and to
40 adapt to climate change while ensuring minimal environmental impact (November, 2010; UK
41 government, 2011; United Nations, 2020). Among these, diversion structures, a name commonly used
42 for structures that are used to raise the water level of a river, are ubiquitous in water supply, irrigation
43 and small scale hydropower applications among others (Tschantz, 2014). For example in the United
44 Kingdom there are 13,000 such diversion structures (weirs), a large number compared to the number of
45 large dams (486). Despite their continuous, wider and increasing use, the design methods used in their
46 design has arguably lagged behind in adapting the computation, visualisation and decision making
47 approaches prominent in other water resource fields such as the design of large dams. In this article we
48 propose a multi objective optimisation approach for diversion structures that reveals trade-off between
49 various design objectives.

50 Design of diversion structures involve calculating the parameters of a diversion structure's components
51 that are set in consideration of surface flow, subsurface flow, nature of foundation soil, structural
52 stability and economy. The parameters of the components of these structures are interrelated.

53 A range of surface and subsurface flow theories have been forwarded for the safe and economical design
54 of low-head diversion structures on permeable foundation. One of the oldest is Bligh's creep theory
55 (Bligh, 1912). The theory assumed the total head loss up to a point along the base of the structure to be
56 proportional to the distance of the point from the upstream of the foundation. This theory has been
57 found to be defective from actual field observations. Pavlovsky (1922) developed a general theory of
58 the conformal transformation problem to weir-foundation design. Lane (1935) proposed an empirical
59 method in which the creep is weighted to allow for the variation in creep along vertical and horizontal
60 directions based on his experiment on large number of dams. Casagrand (1935) formalized the method
61 of flow nets, a graphical solution of the Laplace equation for steady state flow, first developed by
62 Forcheimer. A graphical solution of the Laplace equation is a trial and error method and arduous.
63 Khosla (1935) used method of independent variables based on Schwarz - Christoffel transformation
64 (Christoffel, 1867). By splitting the complex foundation profile into several elementary forms the
65 Khosla's method achieves an approximate result. According to this widely applied method, the
66 components of a diversion structure are calculated based on rule of thumbs such as basing the
67 downstream pile depth to be at least 150 % of the expected scour depth at high flood. The overall apron
68 length is set based on the downstream pile depth, head difference between upstream and downstream
69 points to keep the safe exit gradient within acceptable limit to prevent progressive erosion commonly
70 called piping failure. The thickness of the aprons is set to balance the uplift pressure due to residual
71 seepage pressure as it is dissipated from upstream to downstream.

72 Recent works have suggested solving the complex non-linear design problem with optimisation. (Singh,
73 2011; Hassan and Al-Shukur, 2016) presented an optimization-based procedure that uses genetic
74 algorithm to minimize the overall cost as well as satisfy the safety and functionality requirements. Al-
75 Shukur & Fadhil Hassan, (2015) conducted a parametric analysis to investigate the effect of variation
76 in the design parameters values on the dimensions and on overall cost of the barrage to find optimal
77 hydraulic design of the barrage. Each parameter was taken separately while the others remained
78 constant. Singh (2011) demonstrated the fuzzy based framework for uncertainty characterization in
79 optimal cost for imprecise hydrologic parameter such as seepage head. The nonlinear optimisation
80 formulation is then solved using GA. Garg, Bhagat, & Asthana, (2002) presented method for
81 minimizing the cost of a barrage using an optimization technique along with a parametric analysis to
82 reveal the effects of various parameters on the optimal design barrage. To assess the effect of uncertainty
83 in seepage due to variations in hydraulic conductivity on optimum design using coupled simulation-
84 optimization methodology, Al-Juboori & Datta (2018) trained meta-models on multiple datasets of
85 simulated seepage scenarios. Safety factors and other hydraulic design requirements are imposed as
86 constraints of the optimization model within the simulation model

87 As reviewed above, the recent literature shows a cost effective design can be achieved with optimisation
88 and that many alternative parameter combinations can lead to acceptable designs. However, all
89 demonstrations are limited to reducing cost through a single objective optimization and do not explore
90 the trade-offs between the many relevant design objectives, failing to reveal possibly superior designs
91 by exploring the full parameter space.

92 In this article we present a multi objective optimisation design approach for diversion structures that
93 relies on trade-off analysis with visual analytics. The problem formulation transforms the traditional
94 constraints in the design of diversion structures as performance objectives. The approach results in
95 multiple alternative designs from which decision makers can chose one based on the acceptable trade-
96 offs between metrics of stability against piping, sliding, overturning and uplift failure, and construction
97 cost. The proposed approach is demonstrated on a sample design. A free and open source code is
98 provided for as convenient design and decision tool that simultaneously provide cost efficiency and
99 higher factors of safety against multiple modes of failure such as sliding, rotation, uplift and piping.

100 The following section revises the widely used (traditional) Khosla's method of independent variables.
101 Section 3 presents the proposed reformulation of the diversion structure design problem to a multi
102 objective optimisation one. The simplified design problem with hypothetical hydrologic and soil
103 properties is given in section 4 where the results is also presented. Sections 5 and 6 present discussion
104 and conclusion respectively.

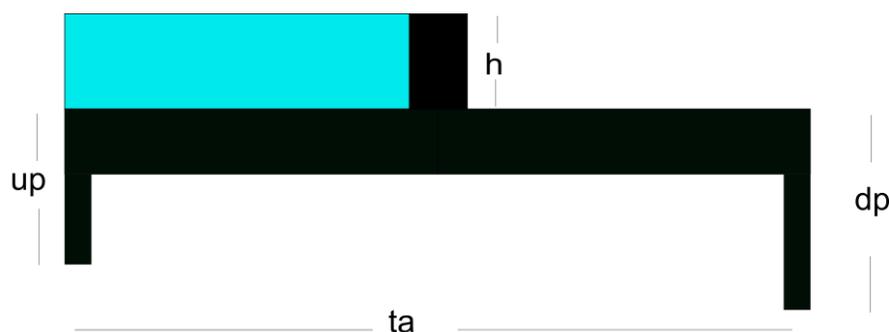
105 **2. Traditional design method for diversion structures on permeable**
106 **media**

107 The longitudinal profile of the structure is designed by considering the surface and subsurface
108 flow of water and geology of the site. Designing a longitudinal profile of the structure includes
109 setting the depths of the upstream and downstream piles, lengths and thickness at various points
110 along the structure of the upstream and downstream aprons.

111 ***Stability against scour***

112 The upstream and downstream piles are designed to guard against anticipated scouring action
113 of surface water. Lacy (1939) found the scour depth (R) depends on soil property and discharge
114 intensity as:

115 $R = 1.35 \left(\frac{q^2}{f} \right)^{\frac{1}{3}}$ where, R= scour depth, q = discharge intensity, f = Lacy's silt factor



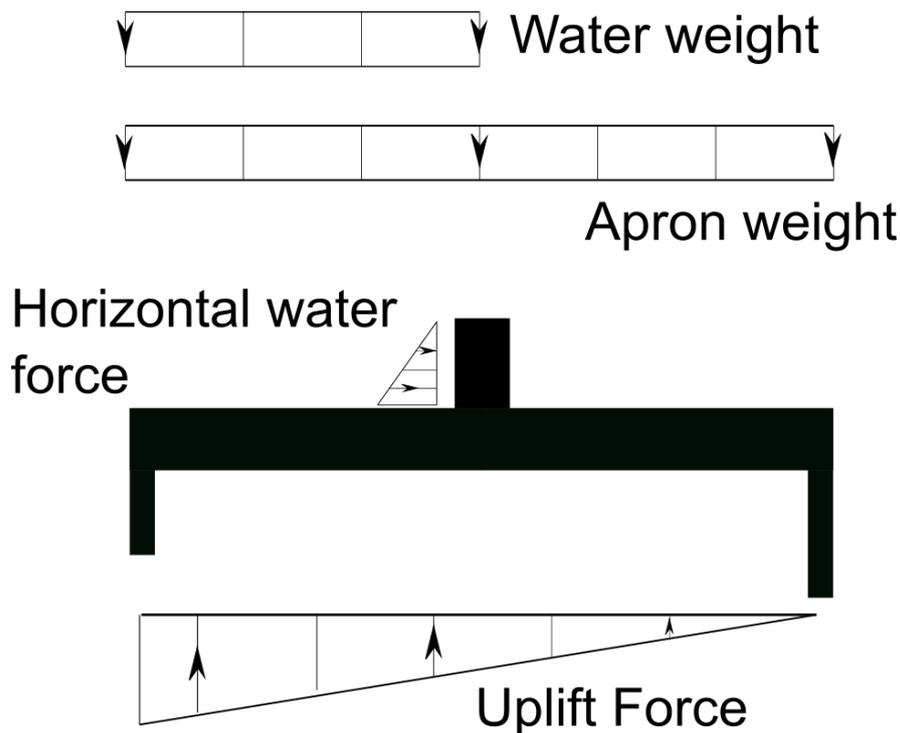
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117 *Figure 1 Simplified drawing of a longitudinal cross-section of a low head diversion structure with*
118 *upstream and downstream piles.*

119

120

121



122

123 *Figure 2 free hand diagram of forces acting on the diversion structure*

124 The downstream pile depth is traditionally set to be more than 1.5 times the scour depth @
 125 below the high flood level (HFL) and the upstream pile depth is set at 1.25 times R below the
 126 high flood level.

127 ***Stability against piping***

128 The exit gradient is the hydraulic gradient of the seepage flow under the base of the weir floor.
 129 The rate of seepage increases with the increase in exit gradient. Such an increase can wash
 130 away by the percolating water. This, commonly known as piping, can be minimized by
 131 reducing the exit gradient. The total length of apron and the downstream pile depth act together
 132 to keep the exit gradient in safe limit. Khosla gives the relation between these parameters as

133 $G_E = \frac{H}{dp\pi\sqrt{\lambda}}$ where, G_E = the exit gradient; λ = a function of the relation between the total
 134 apron length and downstream pile depth; H = is the maximum head difference anticipated from
 135 high flood flow, pond level flow and static water cases; dp = downstream pile depth

136 The value of λ for a permissible exit Gradient G_E and a downstream pile depth d_d is therefore
 137 calculated as $\lambda = \left[\frac{H}{dp\pi G_E} \right]^2$. The value of α is calculated as $\alpha = \sqrt{(2\lambda - 1)^2 - 1}$ from which

138 the total apron length is calculated as calculated as $\alpha = \sqrt{(2\lambda - 1)^2 - 1}$ from which $b_T = \alpha d_d$

139 where b_T = Total apron length, d_p = Depth of downstream pile.

140 The downstream apron length is set to contain the full length of the hydraulic jump for all flow
141 conditions. Viz. high flood flow and pond level flow with and without concentration of flow
142 and retrogression of river with time considered.

143 Design goals are finding the combination of parameters (shown with letters in the figure) that
144 will achieve stability against multiple forms of failure with the least cost

145 performance objectives are maximizing stability against uplift, sliding, overturning, and
146 minimizing cost

147 Constraints are that upstream and downstream piles are deeper than the scour depth

148 Once the pile depths and the length of the aprons are set based on the surface flow (e.g., scour
149 depth) and subsurface flow (e.g., stability against piping failure) considerations, the structure
150 should be checked for stability.

151 ***Stability against uplift***

152 Uplift force exists on the structure because of the subsurface flow of water underneath it. This
153 uplifting pressure head decreases from upstream to downstream. Designed for stability against
154 uplift is achieved by balancing the thickness of aprons at various points along the longitudinal
155 section to the uplift pressure due to the subsurface flow. The forces and moments acting on the
156 corresponding structure are then calculated and the structure is checked for its stability against
157 overturning and sliding.

158 $t = \frac{h}{G-1}$ where, t = Thickness of apron at a point, h = The unbalanced head between the uplifting
159 pressure head and surface water depth, G = Density of construction material for apron

160 ***Stability against Overturning***

161 As with design consideration against uplift, it is important to keep the stabilizing moment more
162 than the destabilizing moments.

163 Since unpredictable situation are likely to occur and cause the toppling moment to exceed the
164 balancing one, a factor of safety of 1.5-2.0 is usually applied for safety against overturning

165 (Baban, 1995). In order to avoid lifting up the structure's heel and tension occurrence at the
 166 base, the resultant force must pass through the middle third of the structure's base.

$$167 \quad e < \frac{ta}{6}$$

$$168 \quad e < \left[\frac{ta}{2} - X \right] < \frac{ta}{6}$$

169 Where $X = \frac{\sum M}{\sum V_f}$, 'ta' is the bed width, M is the moment about the toe and 'Vf' the vertical forces

170 However, if the condition is not satisfied, the tension and compression at the hill of the weir shall be checked
 171 as follows:

172 $\rho_{min} = \frac{\sum F_v}{2} x \left(1 - \frac{6e}{B} \right)$ $\rho_{min} > 0$ and $\rho_{max} = \frac{\sum F_v}{2} x \left(1 + \frac{6e}{B} \right)$ $\rho_{max} < 70$ tones/m² for masonry under
 173 stressing or tension is checked as where e = eccentricity, $\sum M$ = summation of all moments about the
 174 structures toe, $\sum V_f$ = summation of vertical forces excluding the base reaction, X= distance of the
 175 resultant of the forces from the toe, ta = width of the weir base.

176 ***Stability against Shear and Sliding***

177 The structure may slide in the flow direction if there is not enough grip between the base and
 178 the foundation. To prevent this happening, the vertical forces are checked to be adequate,
 179 compared to the horizontal forces, to supply static friction that would keep the structure intact
 180 in its place. The US bureau of reclamation, as quoted by Baban (1995) suggests 0.35 for
 181 concrete structures on common soils. $\frac{\sum V}{\sum H} > 0.35$ Where, $\sum V$ = Sum of external vertical forces,
 182 $\sum H$ = Sum of external horizontal forces

183 **3. Method**

184 Approaches that link simulation models with heuristic global search methods such as evolutionary
 185 algorithms (Deb *et al.*, 2002; Coello Coello, Lamont and Veldhuizen, 2007) are well suited to handle
 186 non-linearity present in diversion structure design. We develop a computer program that calculates
 187 multiple stability performance metrics and cost for a set of input parameters. The simulator is capable
 188 of replicating a design with the traditional method (Khosla's method of independent variable) where
 189 the component parameters are set empirically. The simulator calculates measures of stability
 190 (Table 1).

191 The simulator, which accepts decision variables that alter the minimum design parameters for the
 192 structure as input parameters, is then linked to a multi objective evolutionary algorithm to reveal a set
 193 of Pareto-optimal designs that best balance multiple performance objectives. We validate the proposed
 194 multi objective optimization approach by comparing design results with those based on the traditional
 195 method, and optimization based on parametric grid search. The multi objective design problem is
 196 formulated as:

197 Minimize $F = f(\mathbf{f_u}, \mathbf{f_o}, \mathbf{f_s}, \mathbf{f_e}, \mathbf{f_c})$

198 Where $\mathbf{f_u}$ = stability against uplift, $\mathbf{f_o}$ = stability against overturning, $\mathbf{f_s}$ = stability against sliding,
 199 $\mathbf{f_e}$ = eccentricity, $\mathbf{f_c}$ = cost. Decision variables are the depths of upstream and downstream cut-off
 200 piles, length and thickness of horizontal aprons.

201

202 *Table 1 Performance consideration in traditional method and metrics used in the suggested method*

Design Criteria	Traditional checks	Suggested metric transformation
Stability against Uplift	$\frac{\sum Vg}{\sum Vu} > 1$	Maximize $f_u = \frac{\sum Vg}{\sum Vu}$
Stability against Overturning	$\frac{\sum Ms}{\sum Md} > 2$	Maximize $f_o = \frac{\sum Ms}{\sum Md}$ Where $\sum Ms$ moments leading to stability $\sum Md$ destabilizing moments
Reduce eccentricity	$e < \left[\frac{ta}{2} - X \right] < \frac{ta}{6}$	Minimize $f_e = \left[X - \frac{ta}{2} \right]$ where $X = \frac{\sum M}{\sum Vf}$

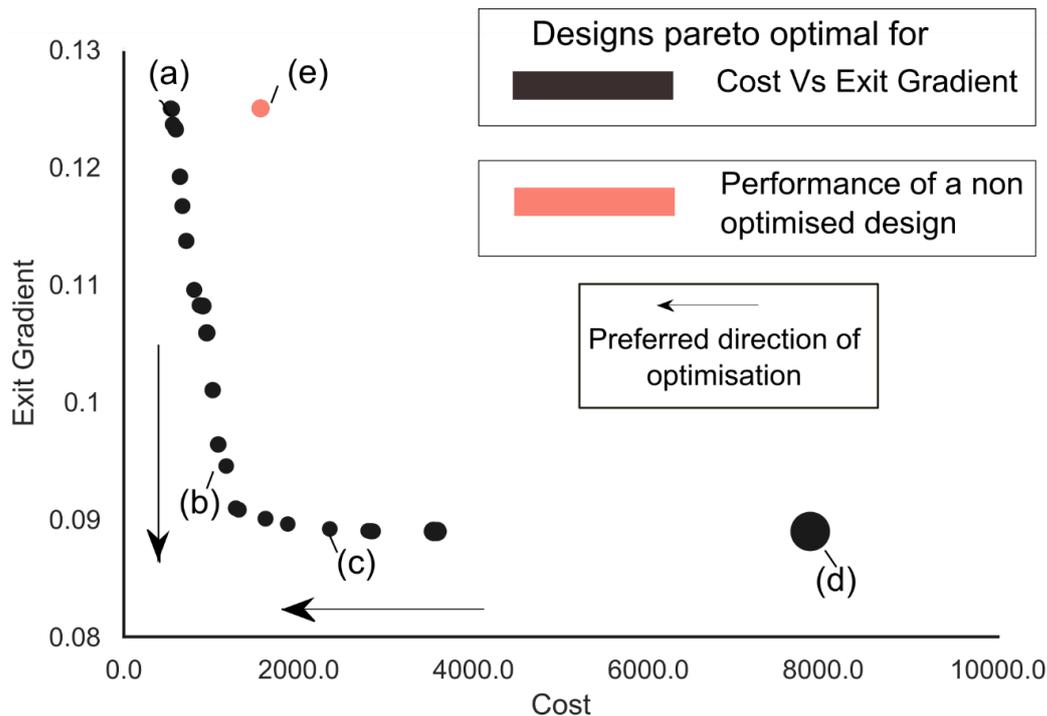
Stability against Shear and Sliding	$\frac{\sum V}{\sum H} > 0.35$	Maximize f_s Where $f_s = \frac{\sum V}{\sum H}$
Cost	Cost was calculated post design	Minimize f_c Where f_c is the sum of costs of apron, piles and excavation and depends on the length, thickness and depth of the components

203 4. Results

204 For this proof of concept paper, we use a synthetic problem with simplified structure. This is done to
 205 ease communication and facilitate the reproducibility of results.

206 For this example we will assume the following site parameters Head = 7.12, Safe exit gradient =
 207 0.125, Discharge intensity = 2.5, Lacy silt factor = 0.75. Density of concrete material (with which
 208 the apron and the cut-off piles will be built) = 2200 kg/m³. Density of water = 1000 kg/m³

209 This section describes results of the proposed approach to the stylized diversion structure design
 210 problem. We start by discussing the stopping criteria for the MOEA computations. We then discuss the
 211 trade-offs between 2 first, followed by 3 and then all 6 performance objectives and discuss the
 212 contribution of various component parameters to the Pareto-optimal designs.



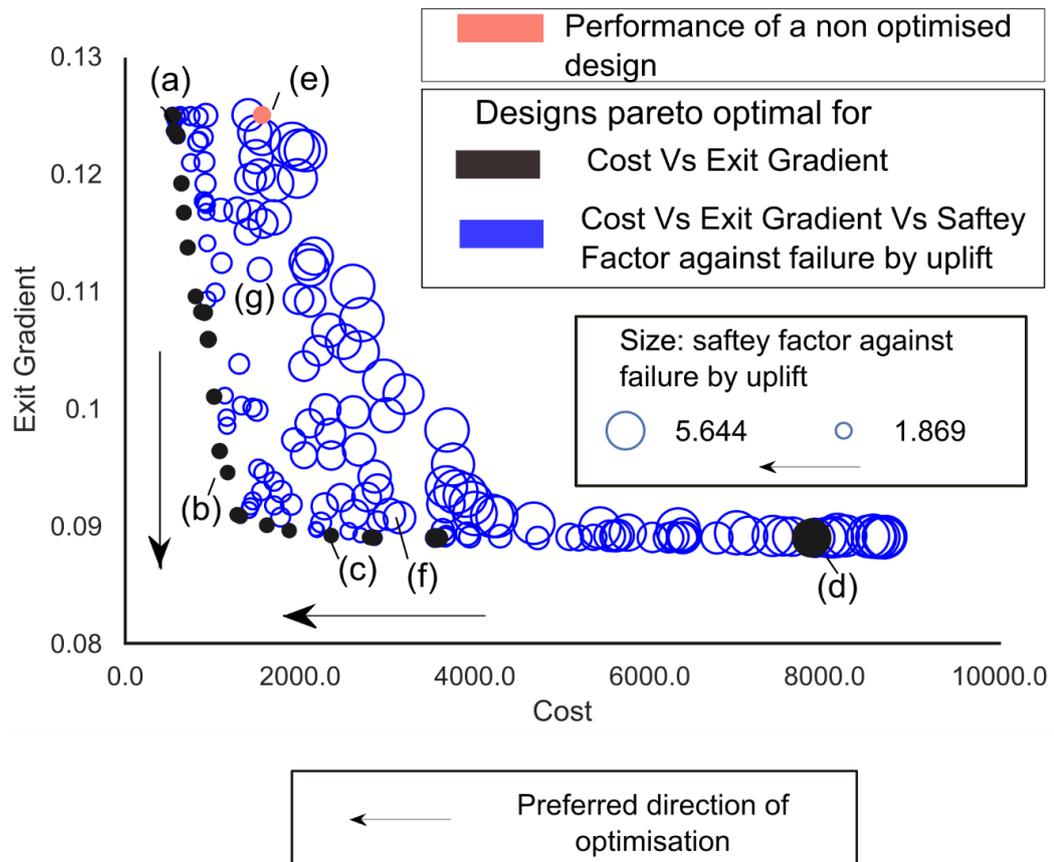
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214 Figure 3 trade-offs between cost and exit gradient. Improving the exit gradient from 0.12 to 0.1 costs
 215 around 50% more (shown with label 'b'). However a further improvement of the exit gradient to 0.09
 216 or less can cost from 300 to 600% (shown with labels 'c' and 'd' respectively).

217

218 Figure 3 shows the performance of the optimised designs and trade-offs between cost and exit gradient,
 219 both of which are preferred to be minimized. Designs labelled 'a' and 'b' show the possible
 220 simultaneous gain compared to the traditional design (shown with label 'e') both in reducing cost and
 221 increased safety against failure by piping (represented by the exit gradient) when optimisation is
 222 applied. Performance comparison of designs 'a' and 'b' show how with a relatively small increase in
 223 cost, a large improvement in the exit gradient may be achieved. However, if the exit gradient were to be
 224 decreased further from label 'b' say to point 'd' a substantial cost increase will need to be sustained (by
 225 more than 6 fold).

226 The best designs when only cost and exit gradient are considered may not meet other performance
 227 criteria such as stability against uplift, overturning and sliding. Blue markers in Figure 4 show designs
 228 that are dominated (hence undiscovered) if only cost and exit gradient were the decision criteria but
 229 would be relevant when stability against uplift (represented by the size of the markers) is also
 230 considered.



231

232 *Figure 4 trade-off between cost, exit gradient and stability against uplift failure. Black circles show the*
 233 *same designs shown in Figure 3. Performance on stability against uplift is represented with the size of*
 234 *the circles. Blue circles (in addition to the black circles) show designs that are Pareto optimal when*
 235 *considering safety factor against uplift in addition to the cost and exit gradient.*

236

237 Figure 4 shows the designs that are Pareto optimal for cost and exit gradient only perform poorly in
 238 their stability against uplift (shown with the size of the marker). However, design with higher
 239 performance in stability against uplift (e.g., 'f' and 'g') can be found with close performance to those
 240 that are Pareto-optimal when considering only cost and exit gradient (e.g., 'b' and 'c'). Although higher
 241 cost is associated with more construction material, designs with higher cost do not necessarily perform
 242 higher in stability against failure by uplift because the larger portion of the material is going into cut-
 243 off pile depths (to reduce the exit gradient) in some of the designs.

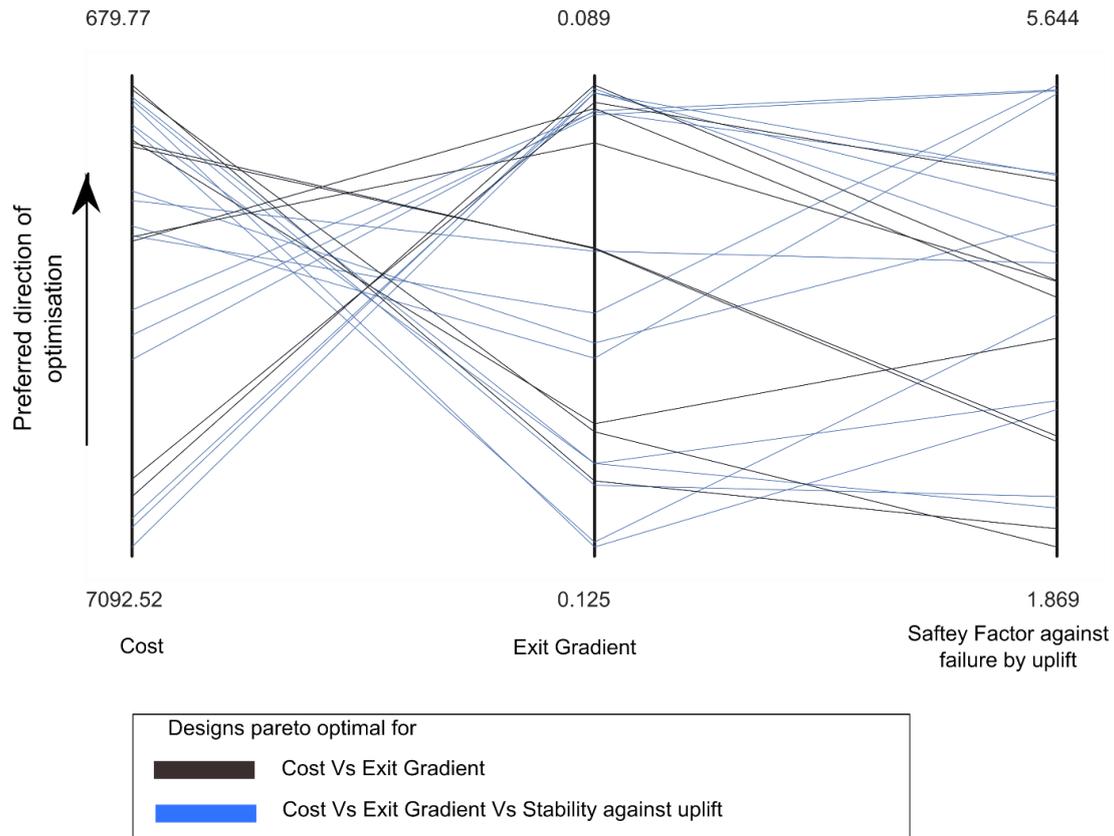
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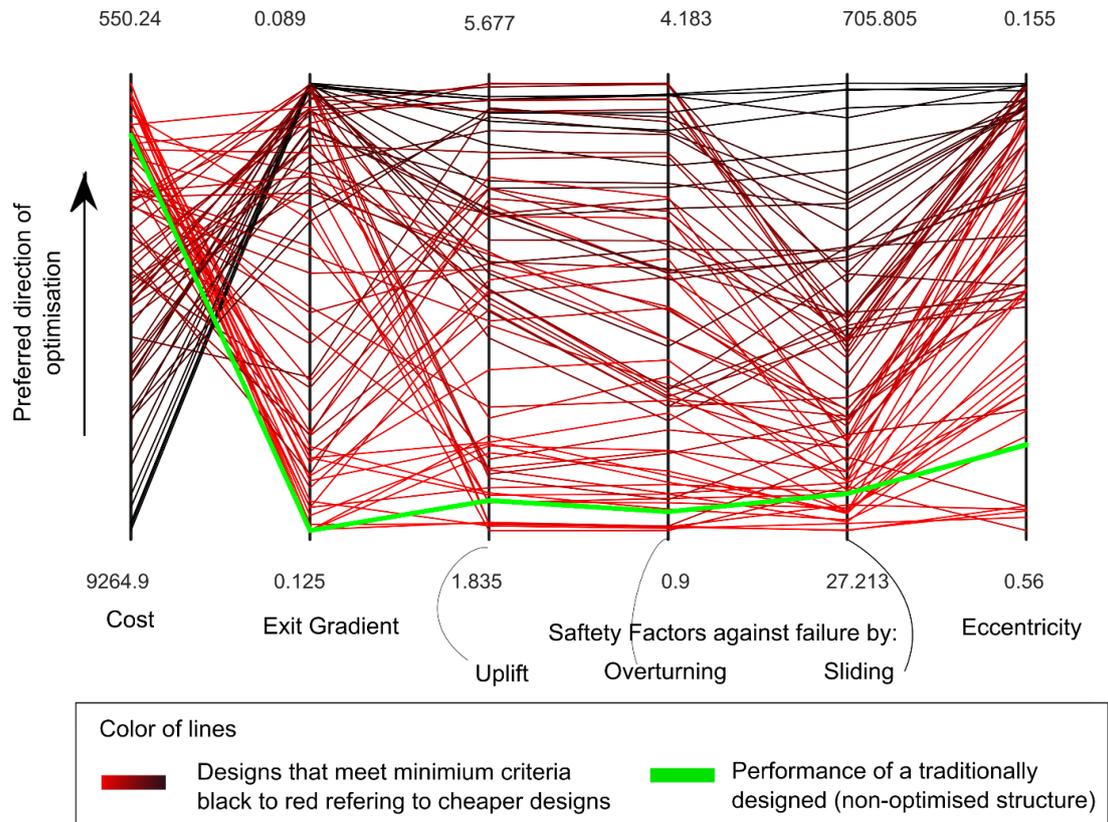
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250 *Figure 5 trade-off between cost, exit gradient and safety factor against uplift failure shown with parallel*
 251 *axis plots. This figure shows the same information as Figure 4 and shown here to introduce parallel*
 252 *axis plots for further analysis of the Pareto optimal designs. An ideal design would have been a*
 253 *horizontal line touching the vertical axes at the top. The largest trade-off is shown to be between*
 254 *minimizing cost and minimizing exit gradient objectives (indicated by the steepness of the line*
 255 *crossings).*

256

257 Figure 5 presents the same information as Figure 4. The values at the top of the parallel axes represent
 258 the highest achievable performance if that particular objective were to be prioritized. The non-vertical
 259 lines represent efficient (Pareto-optimal) designs. Lines that cross between two adjacent axes signal a
 260 trade-off between those measures; the steeper the angle the stronger the trade-off between the two
 261 performance indicators.

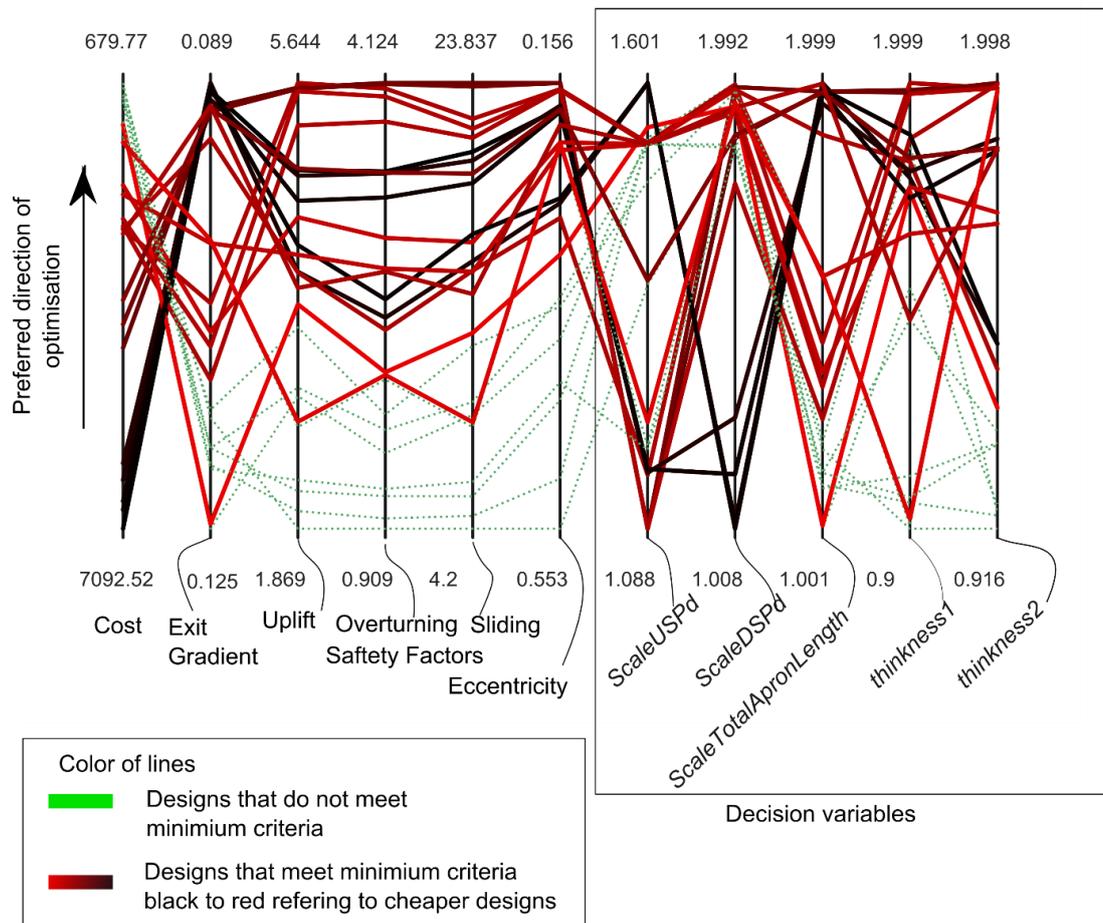
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263

264 Figure 6 A Parallel axis plot showing designs that are Pareto-optimal when all 6 performance objective
 265 are considered. Color of lines show the cost gradient used to visually track how designs performing
 266 well in one of the objective does in the rest of the objectives. The designs that are the cheapest (bright
 267 red lines) are shown to have the lowest score in all other performance measures.

268 Figure 6 demonstrate the multi objective optimisation identifies superior designs than with the
 269 traditional engineering design approach (all lines that consistently score higher in the six performance
 270 objectives than the green line are better). The safety factors against uplift, overturning, sliding and
 271 lowering of eccentricity objectives have relatively small trade-off amongst themselves but exhibit
 272 strong trade-offs with both cost and exit gradient minimizing objectives.



273

274 Figure 7 parallel axis plot showing the design parameters corresponding to the different Pareto-
 275 optimal designs. The solid lines show designs having a score of more than or equal to 2 in the safety
 276 factors against uplift, overturning and sliding and an eccentricity of less than 1/6; the minimum criteria
 277 for acceptable designs. The dashed green lines show designs that are Pareto optimal but that do not
 278 meet the minimum stability criteria. The 5 axes on the right show the design parameters for the Pareto
 279 – optimal designs. Scaling the downstream pile depth is shown to be critical to achieve lower overall
 280 cost.

281 Solid lines in Figure 7 show designs that would be acceptable based on commonly practiced stability
 282 criteria. Among those designs cost is most sensitive to downstream pile depth and to a lesser extent the
 283 downstream thickness. Various combinations of values can lead to acceptable results. However, there
 284 is a strong trade-off between the exit gradient and cost. Some of the least cost design that are cheaper
 285 than the designs with the traditional (non-optimised) approach (Shown with green line in Figure 6) fail
 286 the commonly used minimum stability criteria; showing that low dimensional optimisation (which
 287 doesn't include the stability performance criteria) can be misleading.

288 This figures shows how a decision maker acquainted with traditional design criteria would be able to
289 reduce the choice set. Multiple designs meet the criteria against the 5 failure modes. Line A in the figure
290 is shown for comparing a design made with the traditional method against the optimized ones. And
291 show the multi objective optimization produces designs that are superior in all performance objectives.

292 **5. Discussion**

293 A comparison of the results in Figure 3 and Figure 4 demonstrate how by using the traditional empirical
294 methods or single objective optimization, infrastructure design engineers could inadvertently ignore
295 important decision alternatives and substantial potential to improve performance. The advantage of
296 optimising all performamnce criteria explicitly, rather than treating them as constraints is that design
297 engineer may not know what is possible before seeing the full set of possibilities. This, that considering
298 many-objectives explicitly and simultaneously can help avoid such cognitive myopia was recognised
299 early on by (Hogarth, 1981). Visual interaction with results allows engineers and stakeholders to
300 introduce minimum performance requirements (by filtering or ‘brushing’ results) (Reed and Kollat,
301 2013). The minimum criteria for structural stability or cost can be set post optimisation as shown in
302 Figure 7. Interactive multi-criteria performance plots can help in understanding the implications of
303 choice of balance of sometimes conflicting performance goals on the design parameterization (Reed &
304 Kollat, 2013; Woodruff, Reed, & Simpson, 2013; Geressu & Harou, 2019). By adopting a generate-
305 first-choose-later approach (Herman, Zeff, Reed, & Characklis, 2014; Geressu & Harou, 2015) enabled
306 by many objective optimization and visual analytics, assessments are not restricted by a lone designer’s
307 assumptions of acceptable performance tradeoffs and safety factors against failure.

308 Considering multiple social, economic and environmental metrics simultaneously along with potential
309 impacts of new infrastructure on existing and future use should be preferred to avoid decision biases
310 (Giuliani, Herman, Castelletti, & Reed, 2014; Hogarth, 1981; Geressu et al., 2020). Applying state of
311 the art system design approaches could be crucial both for efficient use of the limited resources (Jeuland
312 *et al.*, 2014) and for creating consensus and cooperation among various stakeholders (Berger *et al.*,
313 2007; Wu *et al.*, 2016). This requires building human resources and institutional capacity along with a
314 culture of consultative decision making.

315 A larger number of low head diversion structures are designed and implemented at more local levels by
316 various engineers and government offices. The design of diversion structures is complex; requiring
317 extensive experience along with a number of tests to craft the right design for a specific project. These
318 requirements are lacking in in many developing countries because of limited human, capital and test
319 and design equipment resources (Baban, 1995). Commonly, higher factors of safety are applied
320 resulting in unnecessarily high cost. In economies with high inflation, the cost of material and
321 construction vary with time; making an optimum design at one point in time (while in design stage) to

322 be obsolete at other (during construction). Computer programs developed with this article allows for
323 the optimum design to easily be revised with short notice (including at the construction stage), reducing
324 overall expenses. A wider application of the design optimization code could also help standardize the
325 designs, lower costs by avoiding large often unwarranted safety factors and ensure designs that are
326 robust to various uncertainties (e.g., high flood magnitude, inflation) are considered. Moreover, the
327 design of low head diversion structures, by its nature is multi objective (e.g., minimizing cost,
328 maximizing safety against uplift, overturning , sliding, etc.); making it ideal to introduce the wider
329 practicing engineering community to the use of trade-off analysis in decision making.

330 This study optimised a simplified diversion structure problem for brevity and as a proof of concept. The
331 design large low-head diversion structures particularly on natural rivers involves extensive analysis of
332 surface flow condition at high floods. We used Khosla's method of independent variable for the
333 calculation of seepage. Other analytical or computational methods such as finite difference method
334 may achieve better accuracy in calculating the seepage head loss and its distribution under the horizontal
335 apron (Al-Juboori and Datta, 2018; Tilaye and Hailu, 2020).

336 The optimum combination different parameters of the components of a diversion structure is dependent
337 on the cost of construction of the components at a particular site. The optimal design would vary based
338 on the relative cost of piles and apron construction cost. This would depend on specific site condition
339 and availability and choice of material for construction. Hence generalizing the sensitivity of cost to
340 either the pile depths or length and thickness of apron could be misleading.

341 Unlike many water resources system problems, where the possible performance in energy, irrigation
342 and other water demand supply, the performance trade-offs for the design of low-head diversion
343 structures can be continuous to infinity. This is because there is unlimited safety factors against uplift,
344 and sliding do not have a limit if they are to be maximized. In the current formulation, the possible
345 maximum value are limited only by the range of the decision variables that the optimisation run is
346 allowed to explore. Although the simulation model is small and relatively small enough to be optimised
347 even using widely available personal computer, future studies should explore transforming the
348 performance metrics so that the unnecessarily computations related to the exploration of likely
349 unwanted designs can be avoided. This may be achieved by limiting the cost of any optimised design
350 to be within a limited range relative to the cost of a design with the traditional design procedures.

351 **6. Conclusion**

352 Traditionally, the components of a diversion structure are calculated with empirical recommendations
353 derived from experience such as the downstream pile depth being least 150 % of the expected scour
354 depth at high flood. The overall apron length is set based on the downstream pile depth, head difference

355 between upstream and downstream points to keep the safe exit gradient within acceptable limit to
356 prevent progressive erosion commonly called piping failure. The thickness of the aprons is set to
357 balance the uplift pressure due to residual seepage pressure as it is dissipated from upstream to
358 downstream. These and other parameters have complex non-linear relationship; multiple parameter
359 combination leading to safe designs but with probably higher cost than the minimum necessary.
360 However, currently practiced design approaches such as Khosla's method of independent variable do
361 not explore the trade-offs between the many relevant design objectives, failing to reveal possibly
362 superior designs.

363 In this article we propose and demonstrate a multi objective optimisation design approach for diversion
364 structures. The problem formulation transforms the traditional constraints in the design of diversion
365 structures, such as safety factor against uplift, overturning and sliding as performance objectives. The
366 simulation-optimisation method explores alternative designs by scaling the parameters of the diversion
367 structure's different component from their minimum recommended values. This generate first choose
368 later approach is used with visual analytics to show the sensitivity of the performance metrics (e.g., cost
369 and exit gradient) to the design parameters (e.g., the depth of cut-off piles and the length and thickness
370 of the horizontal apron). Simultaneously considering the various stability performance objectives reveal
371 the trade-offs between themselves and with cost.

372 The results show designs optimised for single of few of performance objectives (e.g., to minimize cost
373 and exit gradient) can perform poorly in their stability criteria such as safety again uplift. However,
374 design with higher performance in stability can be found with only small increase to the cost or exit
375 gradient if the stability performance metrics are considered simultaneously in a multi objective
376 optimisation. Multi objective optimisation identifies superior designs than with the traditional
377 engineering design approach. The safety factors against uplift, overturning, sliding and lowering of
378 eccentricity objectives have relatively small trade-off amongst themselves but exhibit strong trade-offs
379 with both cost and exit gradient minimizing objectives.

380 A larger number of low head diversion structures are designed and implemented at more local levels by
381 various engineers and government offices. The design of diversion structures is complex; requiring
382 extensive experience along with a number of tests to craft the right design for a specific project. These
383 requirements are lacking in in many developing countries because of limited human, capital and test
384 and design equipment resources. Hence the wide application of the freely available and easy to
385 understand design optimization code could help standardize the designs and lower costs by avoiding
386 large often unwarranted safety factors.

387

388 **7. Appendix: Computational details**

389 We employ a heuristic optimization approach where a search algorithm (Hadka & Reed, 2012; Kollat
390 & Reed, 2006) is coupled with a simulation model of forces on a diversion structure built on permeable
391 foundation. The ϵ -NSGAI (Deb et al., 2000) generates its initial random population of decision
392 variables by exploiting uniform random sampling within the user-specified ranges. These variables are
393 then passed as input variables to the water resources simulator that evaluates the performance of the
394 system. The performance information is passed back to the ϵ -NSGAI algorithm, which evaluates the
395 fitness of the decision variables to produce the next generation of decision variables. Over successive
396 generations of the optimization run, the high quality solution are passed into the epsilon-
397 dominance archive and injected into the population at the beginning of the next run and to
398 automatically adjust the search population size (Kollat & Reed, 2006). The ϵ -dominance archive
399 sorts solutions based on the user specified levels of significant precision (i.e., the minimum
400 change in performance level between solutions for a user to identify them as significantly
401 different).

402 The ϵ -NSGAI was chosen for its search effectiveness for its availability (Kollat & Reed, 2007).
403 An Initial population size of 25 is used with a 0.01 epsilon value was used for all the performance
404 metrics in the Platypus framework (Hadka, 2018) on a personal laptop computer. Other recent
405 algorithms such as Borg Hadka & Reed (2013) may lead to better computational savings of using the
406 MOPO formulation.

407 Performance of many objective evolutionary algorithms is stochastic, with no guarantee that a particular
408 single optimization run will achieve close performance levels as the true (but unknown) Pareto-front.
409 Pareto-sorting of solutions from different random seeded runs (with different initial points) could better
410 approximate the extent of the true-Pareto front [Salazar et al., 2017]. A lack of improvement in the
411 hypervolume score with more evolution (number of generations) is taken as the stopping criteria.

412 The many objective optimization is conducted with up to 30 runs with different initial points (random
413 seeds) where each is allowed to last for up to 2000 function evaluations. The results from each run are
414 then sorted together to provide the best overall reference set (Kollat, Reed and Kasprzyk, 2008).

415 **8. References**

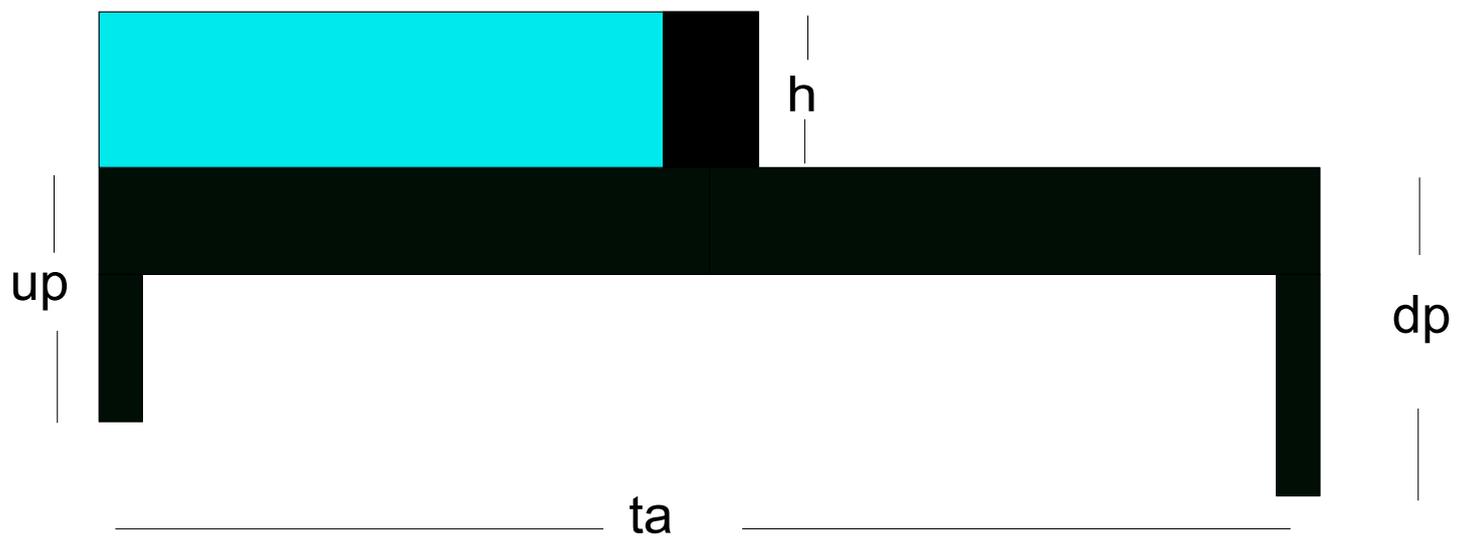
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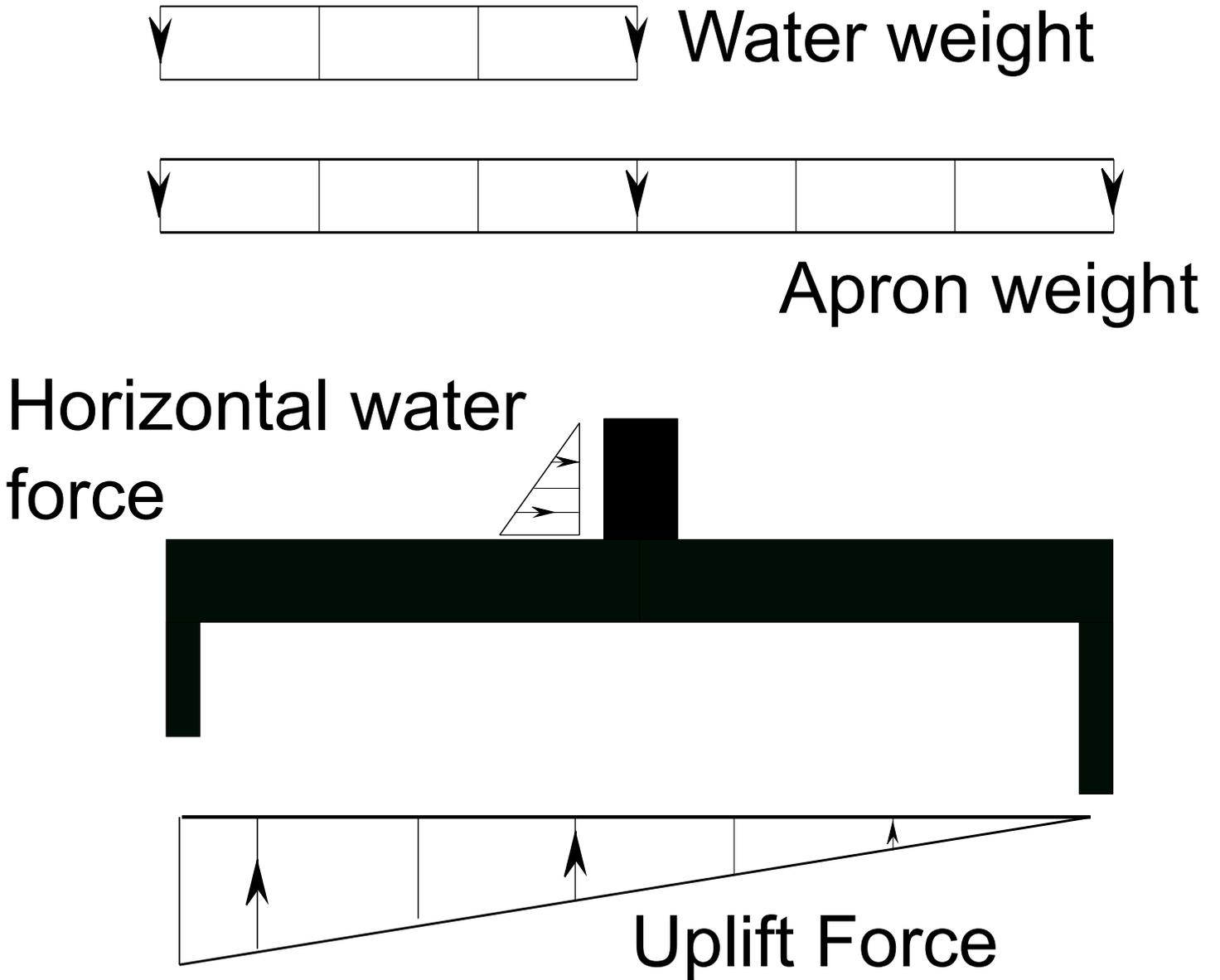
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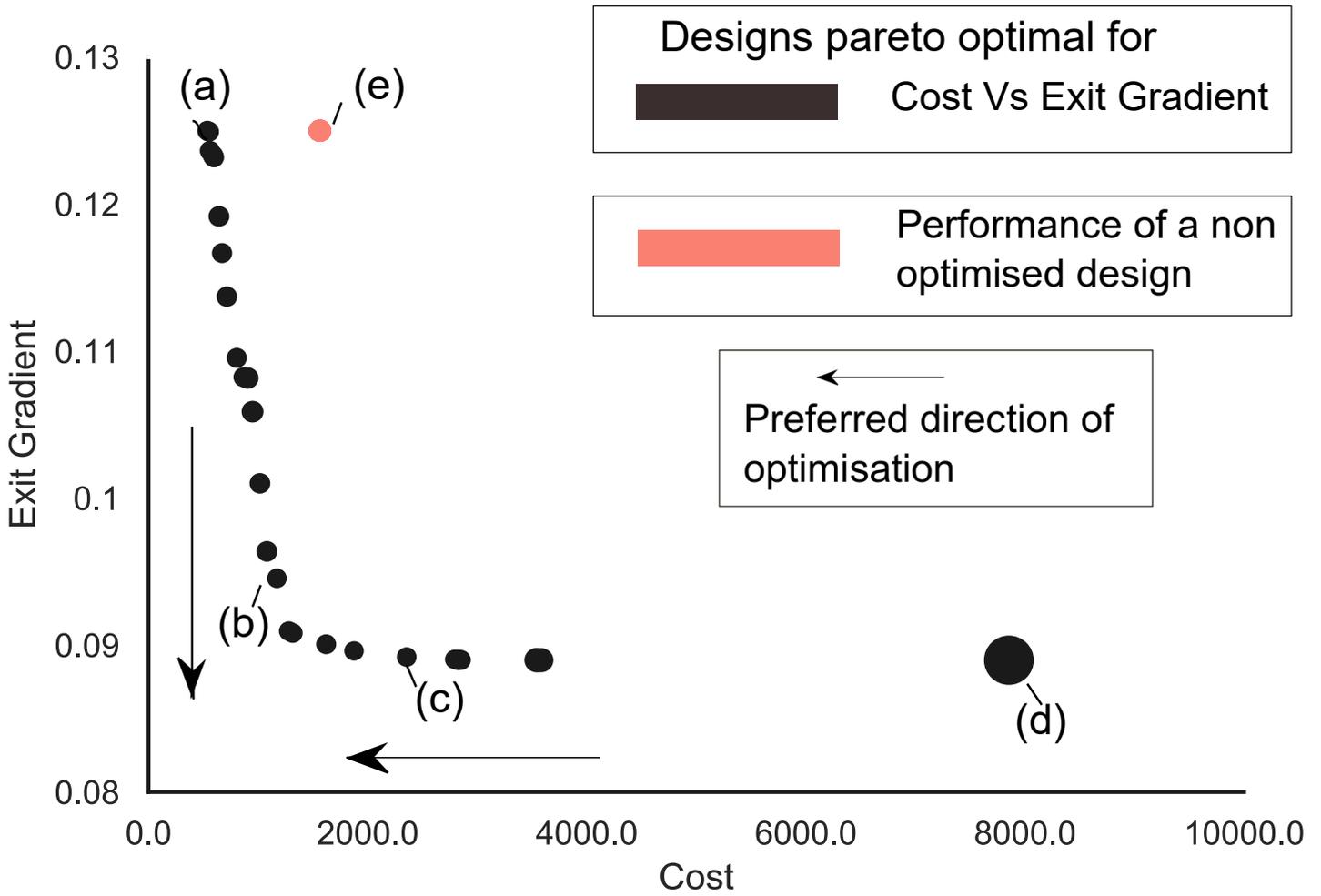
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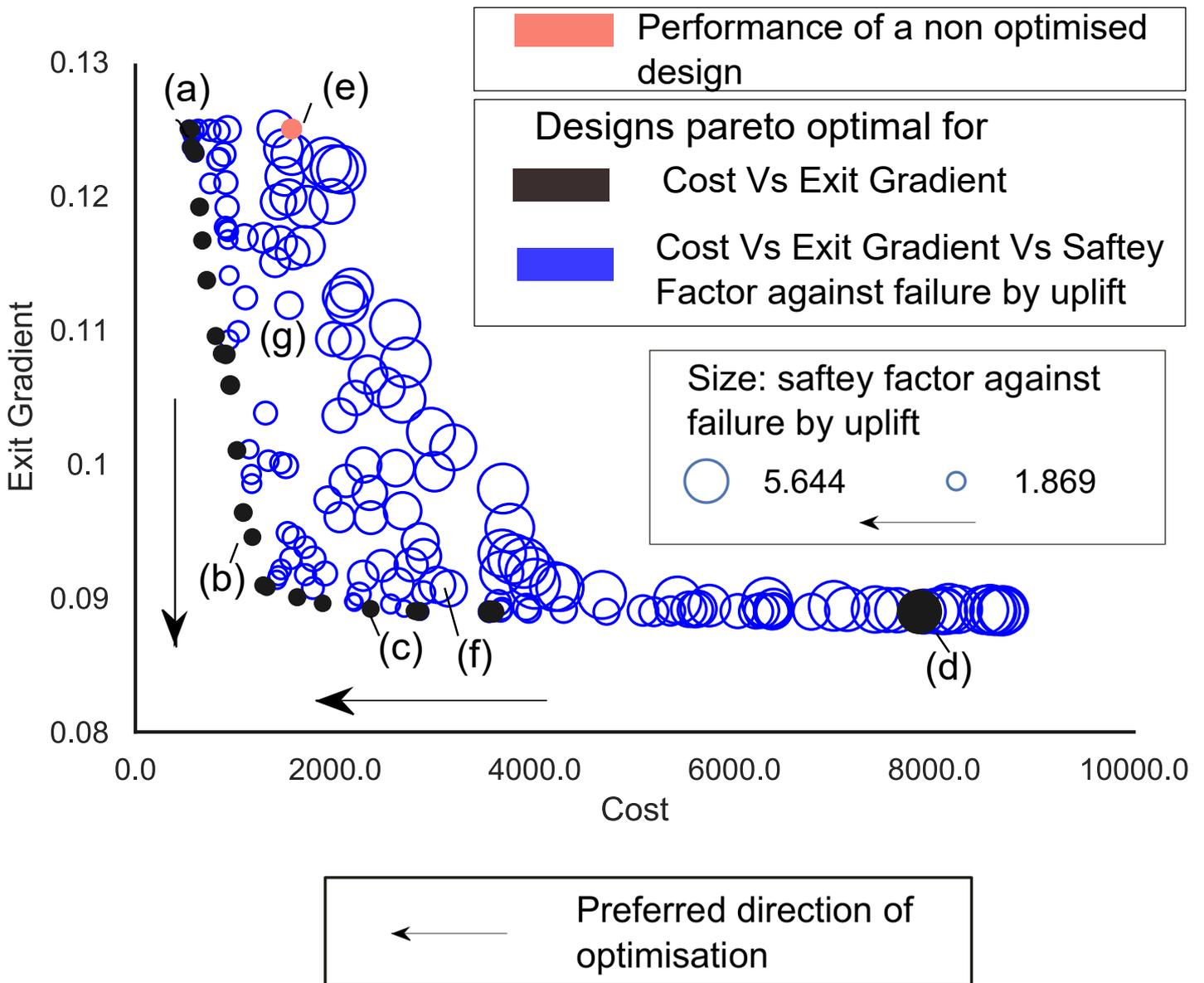
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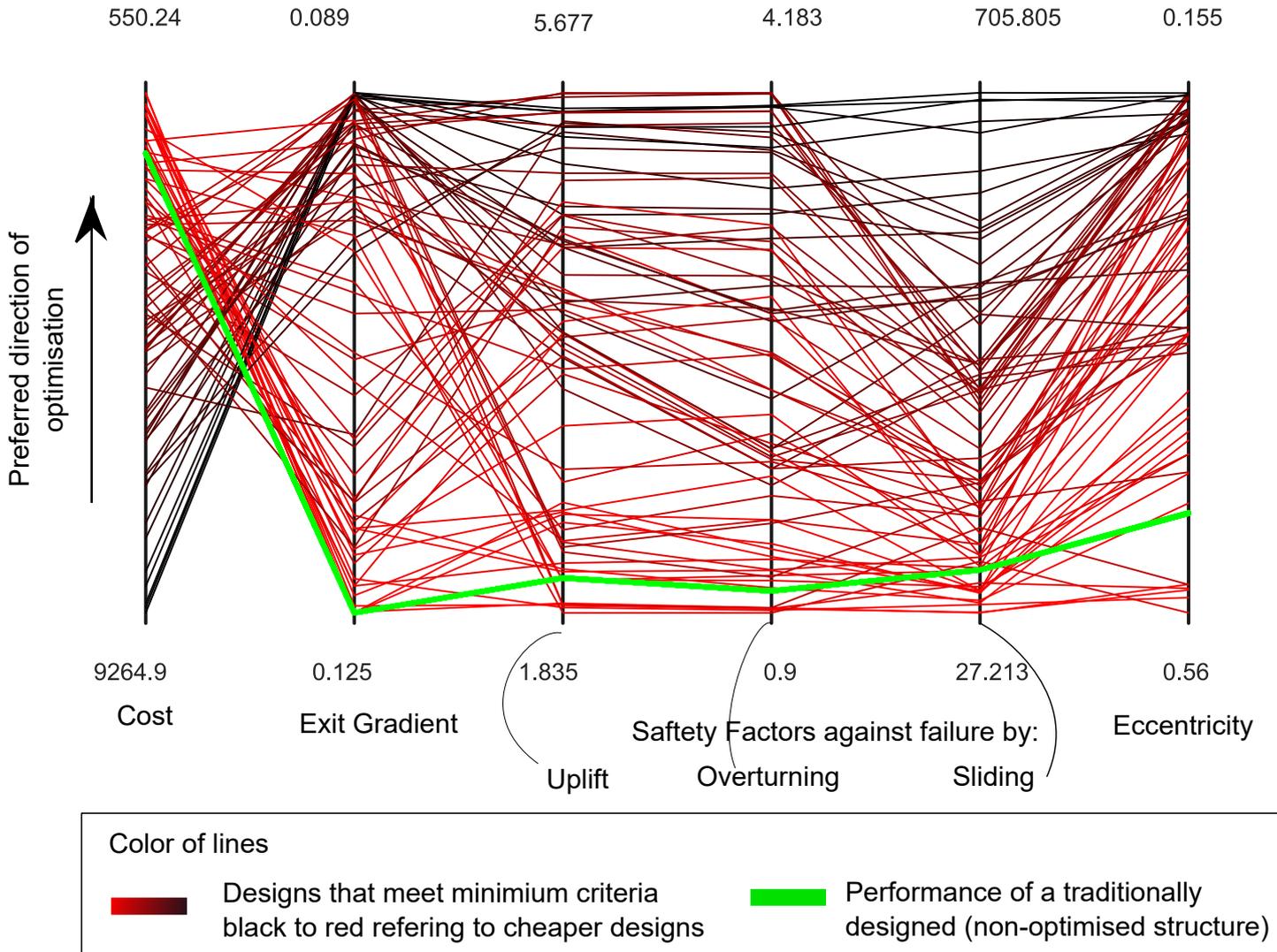
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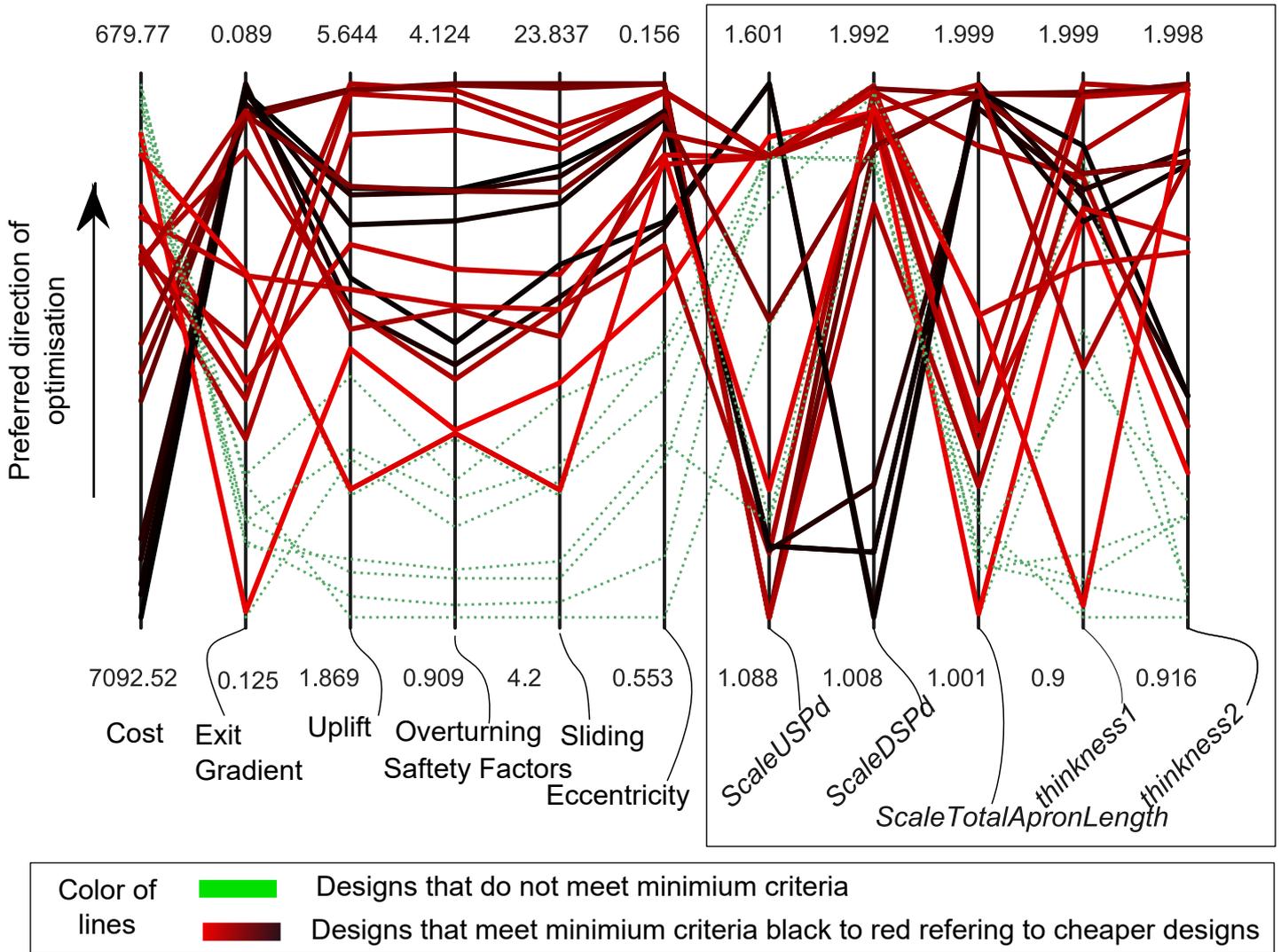


Table 1 Performance consideration in traditional method and metrics used in the suggested method

Design Criteria	Traditional checks	Suggested metric transformation
Stability against Uplift	$\frac{\sum Vg}{\sum Vu} > 1$	Maximize $f_u = \frac{\sum Vg}{\sum Vu}$
Stability against Overturning	$\frac{\sum Ms}{\sum Md} > 2$	Minimize $f_o = \frac{\sum Ms}{\sum Md}$ Where $\sum Ms$ moments leading to stability $\sum Md$ destabilizing moments
Reduce eccentricity	$e < \left[\frac{ta}{2} - X \right] < \frac{ta}{6}$	Minimize $f_e = \left[X - \frac{ta}{2} \right]$ where $X = \frac{\sum M}{\sum Vf}$
Stability against Shear and Sliding	$\frac{\sum V}{\sum H} > 0.35$	Maximize f_s Where $f_s = \frac{\sum V}{\sum H}$
Cost	Cost was calculated post design	Minimize f_c Where f_c is the sum of costs of apron, piles and excavation and depends on the length, thickness and depth of the components