

Design optimization of cold-formed steel portal frames taking into account the effect of building topology

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Cold-formed steel portal frames are a popular form of construction for low-rise commercial, light industrial and agricultural buildings with spans of up to 20 m. In this article, a real-coded genetic algorithm is described that is used to minimize the cost of the main frame of such buildings. The key decision variables considered in this proposed algorithm consist of both the spacing and pitch of the frame as continuous variables, as well as the discrete section sizes. A routine taking the structural analysis and frame design for cold-formed steel sections is embedded into a genetic algorithm. The results show that the real-coded genetic algorithm handles effectively the mixture of design variables, with high robustness and consistency in achieving the optimum solution. All wind load combinations according to Australian code are considered in this research. Results for frames with knee braces are also included, for which the optimization achieved even larger savings in cost.

Keywords: cold-formed steel; portal frames; building topology; real-coded genetic algorithm

1. Introduction

The majority of portal frames use conventional hot-rolled steel sections for the primary loadcarrying members (*i.e*₁ columns and rafters) and cold-formed steel for the secondary members (*i.e*₁purlins, side rails and cladding). Using hot-rolled steel, spans of up to 60 m can be achieved. For frames of more modest spans, the use of cold-formed steel for the primary load-carrying members (*i.e*₁ columns and rafters) should be an alternative to conventional hot-rolled steel.

However, because fabrication and erection costs for cold-formed steel are much lower than for hot-rolled steel, there is scope to vary the frame spacing and pitch. Other advantages of coldformed steel frames compared to hot-rolled steel are as follows. Pregalvanized cold-formed steel sections that do not require painting to prevent rusting are maintenance free. The transportation costs are lower owing to efficient stacking of cold-formed steel sections. Also, the acquisition costs are lower as the cold-formed steel used for the secondary members can be purchased from the same manufacturer/supplier.

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Figure 1. Cold-formed steel portal framing system.

Cold-formed steel portal frames (Figure 1) are a popular form of construction in Australia and the UK. They are commonly used for low-rise commercial, light industrial and agricultural buildings with spans of up to 20 m. For such frames, moment-resisting joints at the eaves and apex can be formed through mechanical interlock (Figure 2a). As can be seen from Figure 2(a), the joints are formed through brackets that are bolted to the webs of the channel-sections. Under moment, the matching swages in both the brackets and the webs of the channel-sections interlock, thus forming a rigid joint (Kirk 1986). With longer span frames, to reduce the section sizes of both the column and rafter members, a knee brace is often included at the eaves (Figure 2b). The effect of including a knee brace is that the bending moment that needs to be resisted by both the column and rafter around the joint is reduced, with the axial load carried through the knee brace (Rhodes and Burns 2006).

Over the past three decades, the design optimization of structures has attracted much attention from researchers (Kuhn and Tucker 1951, Moses 1964, Allwood and Chung 1984, Erbatur and Al-Hussainy 1992, Gero *et al.* 2005). Many structural engineering design problems have discrete decision variables. For instance, in the design of steel frames, the cross-sectional sizes of the columns and rafters are selected from standard tables. One of the most efficient methods for solving complex combinatorial optimization problems such as the design of steel portal frames is genetic algorithms (GAs), based on the Darwinian principle of survival of the fittest and adaptation (Holland 1975, Goldberg 1989).

Binary-coded GAs have been applied to the design of hot-rolled steel frames to find the discrete cross-sectional sizes for the members that minimize the weight of the structure (Chen and Chen 1997, Kameshki and Saka 2001, Toropov and Mahfouz 2001, Gero *et al.* 2006). However, one of the limitations of binary-coded GAs is the extra computational complexity of the algorithm associated with continuous decision variables (Deb 2001). Therefore, real-coded GAs (Wright 1991, Michalewicz 1992, Ono and Koframeashi 1997, Deb 2001, Deb and Gulati 2001) were proposed to resolve the drawback of binary-coded GAs.

In design optimization of hot-rolled steel portal frames, Saka (2003) described a binary-coded
genetic algorithm to minimize the weight of a portal frame through selecting the most appropriate
hot-rolled steel section sizes for the columns and rafters, from a catalogue of available standard
sections, based on elastic analysis and design as described in the British Standards (BS 5950-1).
More recently, Issa and Mohammad (2010) described a genetic algorithm using binary strings,
to study the same problem. They varied the length and depth of the haunched part of the rafter in

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hot-rolled steel portal frames having tapered members, according to the Chinese specification for

148 portal frames (CECS-102).

Previous research focused on the design of hot-rolled steel portal frames with fixed topology where the pitch and frame spacing were not optimized and were specified *a priori*. However, the optimization of cold-formed steel portal frames with either fixed or variable topology has not been considered previously. In this article, a genetic algorithm is proposed to minimize the cost of cold-formed steel portal frame buildings by minimizing the cost of the main structural elements per unit length of the building. Although any code of practice can be used the Australian code was adopted, rather than the UK code, since in Australia the spans of the frames can be larger as there is less snow.

The optimization method proposed addresses all the relevant combinations of the permanent and imposed loads, incorporates the full range of design constraints and considers all feasible wind load combinations. It is assumed here that full lateral restraint is applied to columns and rafters. It is also assumed that the cost of the purlins, side rails and sheeting is independent of frame spacing. The present research differs from previous work on hot-rolled steel portal frames in that the cross-sectional sizes of the columns and rafters and the topology of the building, including the pitch and frame spacing, are all jointly optimized simultaneously. The decision variables used in the design optimization are the spacing of the frames; the pitch of the roof ind the cross-sectional sizes of the main structural elements. Self-evidently, the solution space has both discrete and continuous variables. Unlike in previous research on hot-rolled steel frames that used binary coding for the genetic algorithm, real coding is used here.

2. Frame descriptions

2.1. Frame parameters

Figure 3 shows the parameters used to define both a rigid-jointed cold-formed steel portal frame (Figure 3a) and a rigid-jointed cold-formed steel portal frame having knee braces at the eaves





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Figure 4. Details of cold-formed steel channel-section: (a) single channel-section (C-section); (b) back-to-back channel-section (BBC).

Table 1. Dimensions and section properties of cold-formed steel sections.

221	Section	D (mm)	<i>B</i> (mm)	<i>t</i> (mm)	$EA(\times 10^2)$ (kN)	$EI(\times 10^6)$ (kN.mm ²)	Weight (kg/m)	Cost (A\$/m)
222	<u></u>	402			151.0	72.0	1.70	
223	C10010	102	51	1.0	451.0	/3.8	1.78	5.58
222	C10012	102	51	1.2	555.0	88.2 110.7	2.10	0.15 6.77
224	C10015	102	51	1.5	840 5	137.4	3.29	8 37
225	C15010	152	64	1.0	604.8	225.5	2.32	7.03
226	C15012	152	64	1.2	717.5	264.5	2.89	7.99
227	C15015	152	64	1.5	902.0	330.1	3.59	8.46
228	C15019	152	64	1.9	1148.0	414.1	4.51	10.52
220	C15024	152	64	2.4	1455.5	520.7	5.70	12.88
229	C20012	203	76	1.2	922.5	574.0	3.50	8.99
230	C20015	203	76	1.5	1148.0	723.7	4.49	10.04
231	C20019	203	76	1.9	1455.5	924.6	5.74	12.56
222	C20024	203	76	2.4	1845.0	1166.5	7.24	15.29
232	C25015	254	76	1.5	1312.0	1250.5	5.03	13.66
233	C25019	254	76	1.9	1660.5	1562.1	6.50	14.43
234	C25024	254	76	2.4	2091.0	1972.1	8.16	17.82
235	C30019	300	96	1.9	2070.5	2788.0	7.92	22.76
235	C30024	300	96	2.4	2583.0	3485.0	10.09	29.52
236	C30030	300	96	3.0	3280.0	4366.5	12.76	36.25
237	C35030	350	125	3.0	3915.5	7339.0	15.23	44.74

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240 (Figure 3b). These parameters are as follows: span of frame L_f , height to eaves h_f , length of rafter 241 s_f , pitch of frame θ_f , flexural rigidity of members *EI*, axial rigidity of members *EA* and frame 242 spacing b_f . For the case of portal frames having knee braces at the eaves, the positions of the knee 243 braces are fixed relative to h_f , as shown in Figure 3(b).

Figure 4 shows the dimensions of the cold-formed steel channel-sections used for the primary load-carrying members in portal frame buildings. Table 1 shows the cold-formed steel channelsections used in this article. These channel-sections can be used either singly or back to back. The swages on the web of channel-sections improve the load-carrying capacity of the members. However, it should be noted that in order to simplify the checking procedure and to obtain a conservative design, the section properties and member checks are based on plane channel-sections and therefore ignore the benefit of the swages.

251 2.2. Frame geometry

253 In this article, the design optimization of two exemplar frames with L_f of 20 m and h_f of 4 m is 254 considered: Frame A, without knee braces (Figure 3a), and Frame B, with knee braces (Figure 3b). 255 To minimize the cost per unit length of the building, the decision variables are the pitch, frame 256 spacing and cross-section sizes of the members.

It is assumed that the column bases are pinned, and that the purlins and side rails positioned within the web of the members are spaced sufficiently close each other to prevent out-of-plane buckling from occurring.

2.3. Frame loading

2.3.1. Permanent and imposed roof loads

The permanent and imposed roof loads (AS/NZS 1170-1 2002b) that will be applied to the frames are as follows:

- permanent load (G): 0.10 kN/m^2 (purlins, rails, cladding) and self-weight of the members (Table 1)
- imposed load (Q): 0.25 kN/m^2 .

2.3.2. Wind loads

From the Australian Standard code of practice on wind loading for the design of buildings (AS/NZS 1170-2 2002c), the basic wind pressure q_u for the ultimate limit state is calculated from a design wind speed V_{des} , which in turn is calculated from the regional wind speed V_R multiplied by factors M_d (wind direction multiplier), $M_{z,cat}$ (terrain/height multiplier), M_s (shielding multiplier) and M_t (topographic multiplier). It is worth noting that $M_{z,cat}$ depends on both the terrain category and the average height of the building.

For example, Frame A, with dimensions as shown in Figure 5, built in the wind region W in Australia with V_R of 49.4 m/s, the multiplier factors of M_d , M_t and M_s are taken as 1.0 and $M_{z,cat}$ of 0.87. The design wind speed is calculated as follows:

$$V_{des} = V_{site} = V_R.M_d.(M_t.M_s.M_{z,cat}) = 42.98 \text{ m/s}$$
(1)



300 Figure 5. Dimensions of frame.

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301 where V_{site} is the site wind speed and V_{des} is the design wind speed; and the basic wind pressure:

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 $q_u = \frac{0.6 \times V_{des}^2}{1000} = 1.1 \text{ kN/m}^2$ (2)

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where q_u is the ultimate design wind pressure. 306

The design wind pressure acting on each of the four faces of the frame (AB, BC, CD and 307 DE) is obtained by multiplying q_{μ} by a coefficient of pressure and other related factors. The 308 coefficient of pressure acting on each face is obtained from a combination of the external pressure 309 coefficient C_{pe} and the internal pressure coefficient C_{pi} . The external pressure coefficients C_{pe} 310 should be calculated for wind acting on the side and on the end. These values are shown in 311 Table 2, calculated based on AS/NZS 1170-2 (2002c). 312

For buildings of normal permeability without dominant openings, C_{pi} has a minimum value of 313 -0.3 for suction and a maximum value of 0.2 for pressure. 314

The eight wind load combinations for Frame A (WLC1 to WLC8), and their corresponding 315 coefficients for both side wind and end wind, are shown in Table 3. The coefficients of pressures 316 C_{pe} given by WLC1 are illustrated in Figure 6. As can be seen, the frame will be checked for all 317 eight wind load combinations in the design procedure to be described in Section 2.5. 318

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Table 2. Coefficients of external pressure C_{pe} .

	Coefficient C_{pe} on face						
Description	AB	BC	CD	DE			
Wind acting on side of frame (WT1)	0.7	-0.3	-0.3	-0.3			
Wind acting on side of frame (WT2)	0.7	-0.7	-0.3	-0.3			
Wind acting on end of frame (WL1)	-0.65	-0.9	-0.9	-0.65			
Wind acting on end of frame (WL2)	-0.2	0.2	0.2	-0.2			

Table 3. Coefficients of pressure $(C_{pe} + C_{pi})$ corresponding to different wind load cases.

		Coefficient on face					
Wind load combination	Description	AB	BC	CD	DE		
WLC1	Wind on side + internal pressure	0.7 + 0.2	-0.3 + 0.2	-0.3 + 0.2	-0.3 + 0.2		
WLC2	Wind on side + internal suction	0.7 - 0.3	-0.3 - 0.3	-0.3 - 0.3	-0.3 - 0.3		
WLC3	Wind on side + internal pressure	0.7 + 0.2	-0.7 + 0.2	-0.3 + 0.2	-0.3 + 0.2		
WLC4	Wind on side + internal suction	0.7 - 0.3	-0.7 - 0.3	-0.3 - 0.3	-0.3 - 0.3		
WLC5	Wind on end + internal pressure	-0.65 + 0.2	-0.9 + 0.2	-0.9 + 0.2	-0.65 + 0.		
WLC6	Wind on end + internal suction	-0.65 - 0.3	-0.9 - 0.3	-0.9 - 0.3	-0.65 - 0.		
WLC7	Wind on end + internal pressure	-0.2 + 0.2	0.2 + 0.2	0.2 + 0.2	-0.2 + 0.2		
WLC8	Wind on end + internal suction	-0.2 - 0.3	0.2 - 0.3	0.2 - 0.3	-0.2 - 0.3		
	B 0.3 C	0.3	31	D			
0.7				0.3			
_	→			→ 0.0			
	.						

L E

350 Figure 6. Coefficients of wind pressure for wind load combination 1 (WLC1) for standard frame.

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2.3.3. Limit state design

In accordance with AS/NZS 1170-0 (2002a), the frame will be checked at the ultimate limit state for the following three ultimate load combinations:

$$ULC1 = 1.2G + 1.5Q$$
$$ULC2 = 1.2G + WLC$$
$$ULC3 = 0.9G + WLC$$
(3)

It should be noted that ULC3 is used for the uplift wind load combination.

2.4. Frame analysis

A first order elastic frame analysis program, written by the authors for cold-formed steel sections, referred as LGSFrame, is used to analyse and design the portal frame. To take into account second order effects, an amplification factor is applied as described in the Australian code of practice. For each load combination, the bending moment, shear force and axial force are determined. LGSFrame is called to analyse each candidate solution in each generation as shown in Figure 7.

2.5. Member checks

2.5.1. Columns and rafters

In accordance with AS/NZS 4600 (2005), the columns and rafters are checked for combined axial compression and bending, distortional buckling, and combined bending and shear.

The combined axial compression and bending check is:

$$\frac{N_k^*}{\phi_c N_s^k} + \frac{M_{xk}^*}{\phi_b M_{bx}^k} \le 1$$
(4)

where N_s^k is the nominal section capacity of member k in compression, M_{bx}^k is the nominal member moment capacity of member k about the x-axis, N_k^* is the design axial compression in member k, M_{xk}^* is the design bending moment in member k about the x-axis of the effective cross-section, ϕ_c is the capacity reduction factor for compression, and ϕ_b is the capacity reduction factor for bending. The distortional buckling check is:

$$M_{xk}^* \le \phi_b M_{bx}^k \tag{5}$$

with $M_{bx}^{k} = Z_{c}f_{c}$ and $f_{c} = M_{c}/Z_{f}$; where M_{xk}^{*} is the design bending moment in member k about the x-axis of the effective cross-section, ϕ_{b} is the capacity reduction factor for bending, M_{bx}^{k} is the nominal member moment capacity of member k, Z_{c} is the effective modulus at a stress f_{c} in the extreme compression fibre, M_{c} is the critical moment, and Z_{f} is the full unreduced section modulus for the extreme compression fibre.

The combined bending and shear check is:

$$\frac{M_{xk}^*}{\phi_b M_s^k} + \frac{V_k^*}{\phi_v V_{vk}} \le 1 \tag{6}$$

where M_{sk}^* is the design bending moment in member k about the x-axis of the effective crosssection, M_s^* is the nominal section moment capacity of member k about the x-axis, V_k^* is the design shear force in member k, V_{vk} is the nominal shear capacity of the web of member k, ϕ_b is the capacity reduction factor for bending, and ϕ_v is the capacity reduction factor for shear.



435 Figure 7. Flowchart of the real-coded GA.

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2.5.2. Eaves knee braces

439 The knee brace is a pin-ended member and is checked for both compression and tension. The
440 compression check is:
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$$N_k^* \le \phi_c N_{ck} \tag{7}$$

where N_k^* is the design compressive axial force of member k, N_{ck} is the nominal member capacity of the member k in compression, and ϕ_c is the capacity reduction factor for compression.

The tension check is:

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$$N_k^* \le \phi_t N_{tk} \tag{8}$$

449 where N_k^* is the design tensile force of member k, N_{tk} is the nominal section capacity of the 450 member k in tension, and ϕ_t is the capacity reduction factor for tension. D.T. Phan et al.

3. Optimization formulation

453 (The objective of the design optimization is to determine the portal frame building having the 454 minimum cost, while satisfying the design requirements. The cost of the main frame depends on 455 frame spacing, pitch and cross-section sizes. The objective function can be expressed in terms of 456 the cost per unit length of the building as follows:

Minimize W =
$$\frac{1}{b_f} \sum_{i=1}^{n} w_i l_i$$
 (9)

where W is the cost of main frame per unit length of building, b_f is the frame spacing, w_i are the cost per unit length of cold-formed steel sections (Table 1), l_i are the lengths of cold-formed steel structural members, and n is the number of members.

The normalized forms of the design constraints or unity factors given in Equations (4)–(8) are expressed as follows:

$$g_1 = \frac{N_k^*}{\phi_c N_s^k} + \frac{M_{xk}^*}{\phi_b M_{bx}^k} - 1 \le 0$$
 (10a)

$$g_2 = \frac{M_{sk}^*}{\phi_b M_b^k} - 1 \le 0$$
(10b)

$$g_3 = \frac{M_k^*}{\phi_b M_s^k} + \frac{V_k^*}{\phi_v V_{vk}} - 1 \le 0$$
(10c)

$$g_4 = \frac{N_k^*}{\phi_c N_{ck}} - 1 \le 0 \tag{10d}$$

$$g_5 = \frac{N_k^*}{\phi_t N_{tk}} - 1 \le 0 \tag{10e}$$

The design specification is the ultimate limit state (ULS) that constitutes the constraints for the optimization problem. Penalty functions are required to define the relationship between the objective function and constraints and to transform a constrained problem to an unconstrained one (Camp et al. 1998, Pezeshk *et al.* 2000). The relationship known as the fitness function often has the form:

$$F = W[1+C] \tag{11}$$

490 where F is the fitness function, W is the objective function, being the cost of frame per unit length 491 of building, and C is the constraint violation penalty. 492 In this research the penalty value is assigned through the maximum level of violation of the

In this research, the penalty value is assigned through the maximum level of violation of the unity-factor constraints in Equations (10a)–(10e) as follows:

$$g = \max\left[g_1, g_2, g_3, g_4, g_5\right] \tag{12}$$

In this article, penalty values are imposed empirically, in proportion to the severity of constraint violation. Through a numbers of trials, it is observed that two levels of violated constraints with the magnitudes as shown in Equation(13) are suitable to eliminate the violated solutions through

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515 Figure 8. Progress of the GA for Frame A with fixed pitch and frame spacing.

517 the evolutionary process, as follows:

$$C = \begin{cases} 0 & \text{if } g \le 0\\ g & \text{if } 0 < g \le 0.5\\ 10g & \text{if } g > 0.5 \end{cases}$$
(13)

523 The proposed optimization procedure aims to minimize the value of the fitness function F 524 (Equation 11). This is achieved by minimizing the cost W and reducing the penalty C to zero. The 525 procedure involves a real-coded GA, frame analysis and cold-formed steel design. As can be seen 526 from Figure 8, the evaluation process computes the fitness function values using the objective 527 function (Equation 9) along with the corresponding penalty values defined in Equation (13). 528 Better (i.e. cheaper) solutions will yield smaller fitness values, and consequently are selected 529 preferentially by the tournament selection operator. The criterion for terminating the program 530 is a predefined total number of function evaluations. This criterion is suitable to investigate the 531 convergence history of the algorithm.

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4. Real-coded genetic algorithm

536 537 The characteristic of real-coded GAs is that genetic operators are directly applied to the design 538 variables without coding and decoding as with binary GAs. Solving optimization problems using 539 real-coded GAs is therefore less cumbersome when compared to the binary-coded GAs. The 540 algorithm used in this article randomly generates a set of solutions known as the initial population. 541 From this population, the next generation of solutions is evolved by conducting three genetic 542 operations: selection, crossover and mutation. The flowchart of the real-coded GA used in this 543 article is shown in Figure 7.

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545546**4.1.** Selection and elitism strategy

547 In tournament selection operator, the process is conducted by picking at random two solutions 548 from the current population to compare their fitness values. The solution with a better fitness 549 value is selected for the next operation. The process of random selection ensures that the best 550 solutions in the population will not dominate the mating pool, as in the proportional selection

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method. The diversity of the population is thus preserved to increase the exploration component of the algorithm.

The best individuals in the population, depending on the adopted percentage of population, are retained and carried forward unchanged to the next generation. The rest of the new population is created by the three genetic operators of selection, crossover and mutation applied to the entire current population including elite individuals.

4.2. Real-coded crossover and mutation operators

With real coding, the difficulty is how to use a pair of real-coded decision variable vectors to produce a new pair of offspring vectors or how to mutate a real decision variable vector in a meaningful manner (Deb 2001). In this article, the simulated binary crossover (SBX) (Deb and Agrawal 1995) and polynomial mutation (Deb 1997, Deb and Gulati 2001) are applied to create the new individuals for the next generation.

The SBX operator picks at random two solutions in the current population, known as parents, to create two offspring symmetrically to avoid a bias toward any particular parent solution in a single crossover operation. The formulation used for SBX is as follows:

$$x_i^{(1,t+1)} = 0.5[(1+\beta)x_i^{(1,t)} + (1-\beta)x_i^{(2,t)}]$$

$$x_i^{(2,t+1)} = 0.5[(1-\beta)x_i^{(1,t)} + (1+\beta)x_i^{(2,t)}]$$
(14)

where β is the probability distribution function for crossover, $x_i^{(1,t)}$ and $x_i^{(2,t)}$ are the parent solutions, and $x_i^{(1,t+1)}$ and $x_i^{(2,t+1)}$ are the children created for the next generation.

To ensure that the new values of the decision variable remain within the range $[x_i^l, x_i^u]$, where x_i^l and x_i^u are the lower and upper bounds, respectively, the probability distribution for the crossover operator has the form:

$$\beta(\eta_c) = \begin{cases} [\alpha u]^{1/(\eta_c+1)} & \text{if } u \le 1/\alpha, \\ [1/2 - \alpha u]^{1/(\eta_c+1)} & \text{if } 1/\alpha < u \le 1 \end{cases}$$
(15)

where u is a random number between 0 and 1, η_c is the distribution index for crossover, $\alpha =$ $2 - \chi^{-(\eta_c+1)}$, and χ is calculated as follows:

$$\chi = 1 + \frac{2}{x_i^{(2,t)} - x_i^{(1,t)}} \min\left[(x_i^{(1,t)} - x_i^l), (x_i^u - x_i^{(2,t)}) \right]; \text{ assuming } x_i^{(1,t)} < x_i^{(2,t)}$$

Like in the SBX operator, polynomial mutation also uses probability distribution $\overline{\delta}(\eta_m)$, being a polynomial function to create the child solution in the vicinity of a parent solution. The formulation for the mutation operator (Deb 1997, Deb and Gulati 2001) has the form:

$$y_i^{(1,t+1)} = x_i^{(1,t+1)} + (x_i^u - x_i^l)\bar{\delta}$$
(16)

where x_i^u and x_i^l are the boundaries of decision variables, and $y_i^{(1,t+1)}$ is a new solution obtained from the mutation operator.

To ensure that no solution would be created outside the range of x_i^u and x_i^l (Deb and Gulati 2001), the parameter $\bar{\delta}(\eta_m)$ has the following form:

$$\bar{\delta} = \begin{cases} \left[2u + (1 - 2u)(1 - \delta)^{\eta_m + 1} \right]^{1/(\eta_m + 1)} - 1 & \text{if } u \le 0.5, \\ 1 - \left[2(1 - u) + 2(u - 0.5)(1 - \delta)^{\eta_m + 1} \right]^{1/(\eta_m + 1)} & \text{if } 0.5 < u \le 1 \end{cases}$$
(17)

where *u* is a random number between 0 and 1, η_m is the distribution index for mutation, and $\delta = \min\left[(x^{(1,t+1)} - x_i^l), (x_i^u - x^{(1,t+1)})\right]/(x_i^u - x_i^l)$.

601 In this article, $\eta_m = \eta_c = 1$ is used. A technique that rounds off the number in dealing with 602 discrete design variables is used in case SBX crossover or polynomial mutation create decimal numbers. Constant probabilities are assigned to both crossover and mutation operators to reduce 603 the possibility of destroying the good solutions. Based on a number of trials, a crossover probability 604 P_c of 0.9 was used throughout in this study. It was observed that premature convergence happened 605 606 with a low mutation probability. To increase the GA's exploration capacity in the solution space to increase the chance of locating the optimum solution, the mutation probability P_m is tuned 607 empirically as high as 0.1. 608 609

611 612 **5. Design examples**

- 613 614 **5.1.** Frame A with fixed topology
- 615 616 5.1.1. Exhaustive enumeration

Frame A, without knee braces, in which the pitch is 10° and frame spacing is 4 m, is considered. Such a typical pitch and frame spacing are commonly used for cold-formed steel portal frames. This design problem has two discrete decision variables. The optimum cross-sections for the columns and rafters can be determined from the cross-sections shown in Table 1 by exhaustive enumeration using the LGSFrame as mentioned in Section 2.4. As can be seen from the list of cold-formed steel channel-sections, there are 40 options of cold-formed cross-sections used for members, including both single sections (C) and back-to-back channel-sections (BBC).

Four designs satisfied the member checks. These design options are shown in Table 4. As can be seen, the smallest unit cost obtained is A\$513/m with back-to-back channel-section BBC30030 for both columns and rafters. This result will be used to validate the proposed real-coded GA for searching the optimum solution. It is observed that all design constraints, *i.e.* the member checks, are satisfied. The critical constraint relating to combined axial compression and bending moment for rafters governs the design in the case of ultimate load combination ULC3, with the unity factor being 0.9 against the upper limit of 1.0.

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5.1.2. Real-coded GA

634 The real-coded GA is used to determine the optimum cross-sections for the members of Frame 635 A as discrete variables, from the 40 alternative cross-sections for members as mentioned above. 636 The design process (Figure 7) is repeated with three different population sizes to investigate the 637 possibility of reaching the optimum solution. The convergence histories obtained from the GA 638 are depicted in Figure 8. As can be seen, the fitness function converged within 5500 function 639 evaluations. The three runs in the optimization process lead to the same minimum-cost solution 640 as the exhaustive enumeration process. This result proves the reliability of the proposed GA with 641 real-coded parameters. 642

Table 4. Optimum cross-sections for Frame A with fixed topology.

645	Calaran anation	Defterrestion						
646	Column section	Ratter section	g_1	g 2	<i>g</i> ₃	<i>g</i> 4	<i>8</i> 5	W (A\$/m)
647	BBC30030	BBC30030	-0.30	-0.28	-0.32	-1.0	-1.0	513
648	BBC35030	BBC30030	-0.32	-0.30	-0.35	-1.0	-1.0	547
649	BBC30030	BBC35030	-0.35	-0.45	-0.59	-1.0	-1.0	599
650	BBC35030	BBC35030	-0.44	-0.41	-0.58	-1.0	-1.0	633

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Figure 9. Effect of pitch on unit cost of Frame A with fixed frame spacing.

5.2. Frame A with variable pitch

5.2.1. Exhaustive enumeration

670 To validate the capacity of the proposed GA in solving a more complicated problem as the effect 671 of pitch is taken into account, the process of exhaustive enumeration is carried out manually. 672 In this case, the pitch is varied from 5° to 30° in increments of 5° to investigate its effect. The 673 frame spacing is fixed at 4 m. For each pitch, the optimum cross-sections for members of Frame A 674 are determined through exhaustive enumeration by using the LGSFrame program manually. The 675 influence of pitch on the cost per unit length of building is shown in Figure 9. It can be seen that the 676 pitch has a clear influence on the cost of portal frame building. The minimum unit cost obtained 677 is A\$432/m, with a pitch of 20° and cross-sections of BBC30024. However, the constraint of 678 combined axial compression and bending for the rafter check was violated by $g_1 = 0.006$. The 679 logical inference from this result is that the cheapest feasible design can be expected to have a 680 pitch close to 20°.

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5.2.2. Real-coded GA with fixed frame spacing and variable pitch

685 The optimum pitch and cross-sections for the members of Frame A are determined again using the real-coded GA with a fixed frame spacing of 4 m. In this case, the pitch is processed as a 686 687 continuous variable varied in the range of $(5^{\circ}, 90^{\circ})$, while cross-sections are discrete, selected 688 from the list (Table 1). The design process is also repeated with three different population sizes. 689 The progress of the GA is shown in Figure 10. As can be seen, the minimum unit cost achieved 690 from GA is found to be A\$433/m with a pitch of 20.5° and cross-section of BBC30024 for both 691 columns and rafters. As expected, the unit cost obtained from the GA is a bit larger than the infeasible solution with a pitch of 20° from exhaustive enumeration because the GA solution is 692 693 feasible. The design constraint for the rafter becomes active in the case of the combined actions of 694 axial compression and moment. Again, this demonstrates that real-coded GA is an effective and reliable method to find the optimum pitch and cross-sections of members in portal frame design. 695

In addition to the exhaustive enumeration carried out in Subsection 5.2.1, the GA is verified further here by generating the maximum cost feasible design based on a fixed frame spacing of 4 m and variable pitch. To maximize the cost, it is sufficient to multiply the objective function (Equation 9) by -1.0. A population of 100 was used and a total of five random runs of the GA was carried out, with a maximum of 15,000 function evaluations per run.

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Figure 10. GA progress for Frame A with fixed frame spacing.716

The maximum cost obtained from RC-GA is A\$728.5/m with a pitch of 35.5° . The size of both the columns and rafters is BBC35030, which is the largest and most expensive. Only the distortional bucking constraint on the rafter is binding ($g_2 = 0$) under load combination ULC2. The other constraints are slack. This solution was achieved with three GA runs out of five.

The second most expensive solution also has the largest section of BBC35030 for both the columns and rafters. The cost and pitch are A\$725/m and 35°, respectively. The critical constraint is also distortional buckling for the rafter with g_2 equal to 0.01 which is slightly higher than the best solution. Taken together with the cost and pitch, this suggests that this solution is a near optimum. This solution was achieved in one GA run out of five.

One GA run out of the five attempts failed to generate a good maximum cost solution. It is observed that the cost, pitch and section size for members obtained are A\$564/m, 30° and section of BBC30030, respectively. Also, the distortional buckling constraint ($g_2 = 0.02$) is active for rafter under load combination ULC2. This solution is self-evidently not a maximum, but it is still 30% more expensive than the lowest cost solution of A\$433/m.

Therefore, it can be seen that four attempts out of five in total were successful in the sense that the two solutions found in these four attempts are both feasible and satisfactory in terms of the section sizes (*i.e.* the largest) with at least one binding constraint in each case. The maximum cost solution was obtained within 7200 function evaluations. The CPU time was 8.3 hours per GA run using a machine with a 1.86 GHz processor and 1 GB memory. It may be noted that the maximum-cost solution is 68% more expensive than the corresponding minimum-cost solution.

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739 5.2.3. *Real-coded GA with variable frame spacing and variable pitch* 740

741 In this example, a design optimization for Frame A that accounts for effects of both pitch and 742 frame spacing is conducted, using real-coded GA. The pitch and frame spacing are processed 743 as continuous variables. The progress of the optimization process is shown in Figure 11. As 744 can be seen, the three runs of the optimization process converged to the same unit cost, within 745 7500 function evaluations. The optimum pitch and frame spacing for Frame A obtained from 746 the algorithm are 21° and 3 m, respectively. The optimum cross-section for both the columns and rafters is BBC25024 and the unit cost is A\$355/m. The design constraint for the combined actions 747 748 of axial compression and bending on rafter is active, *i.e.* $g_1 = 0$.

749 It can be seen that when the portal frame design includes the pitch and frame spacing as decision 750 variables, the optimum frame obtained is cheaper than an optimum frame with fixed frame spacing

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Figure 11. GA progress for Frame A with variable pitch and frame spacing.

(18%) or fixed pitch (31%). This demonstrates the benefit of topology optimization covering both pitch and frame spacing. It should be emphasized that the above-mentioned cost savings relate to the typical frame geometry, and not to previously optimized results described in the literature; the latter are not available, to the authors' knowledge.

5.3. Frame B with fixed pitch and fixed frame spacing

In this example, the design optimization of Frame B having knee braces with a fixed pitch is considered using the real-coded GA. The pitch is fixed as 10° and the frame spacing is prespecified as 4 m. In this problem, the optimum cross-sections for the columns, rafters and knee braces are again to be selected from the aforementioned list of 40 cold-formed cross-sections that include both single and back-to-back channel-sections (Table 1).

The population size of 160 individuals is used for determining the optimum cross-sections for members, since it has been proven in the previous sections to be the appropriate size to search for the optimum solution. The maximum permitted number of function evaluations is set at 8000. The optimum cross-sections and corresponding unit costs are given in Table 5. The design constraint of knee brace buckling becomes active from load combination ULC2.

Thus, for the specified pitch of 10° and frame spacing of 4 m, Frame A, which does not have knee braces at the eaves, is 19.8% more expensive than Frame B, which has knee braces.

5.4. Frame B with variable pitch and variable frame spacing

The optimum design of Frame B, taking into account the effect of pitch and frame spacing simultaneously, is considered in this example. There are five design decision variables in this problem: pitch and frame spacing, considered as continuous variables; and column, rafter and knee brace cross-sections as discrete variables. This optimization problem is more complicated

Table 5. Optimum cross-sections for Frame B with fixed topology.

Member type	Cold-formed steel sections	g_1	<i>g</i> ₂	<i>g</i> ₃	g_4	85	W (A\$/m)
Column	BBC 30024	-0.18	-0.31	-0.12	-1.00	-1.00	428
Rafter	BBC 30024	-0.12	-0.13	-0.10	-1.00	-1.00	
Knee brace	C 20015	-1.00	-1.00	-1.00	0	-0.34	

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Column Rafter Knee brace 'àble 7. Nea	BBC 25024 BBC 25024 C 20015		82	<i>g</i> ₃	g_4	85	θ_{f}	<i>oj</i> (m)	())
Rafter Knee brace Pable 7. Net	BBC 25024 C 20015	-0.01	-0.03	-0.09	-1.00	-1.00	17.5°	4.0	270
Xnee brace	C 20015	0	-0.01	-0.07	-1.00	-1.00			
able 7. Ne		-1.00	-1.00	-1.00	-0.02	-0.38			
	ar-optimum solution for Frame	e B with v	variable to	opology.					
Aember type	Cold-formed steel sections	g_1	<i>g</i> ₂	<i>g</i> ₃	<i>g</i> 4	85	$ heta_{f}$	b_f (m)	W (A\$/m)
Column	BBC 30030	0	-0.06	-0.04	-1.00	-1.00	17°	7.6	283
latter	BBC 30030	-0.02	-0.08	-0.07	-1.00	-1.00			
thee brace	BBC 10015	-1.00	-1.00	-1.00	-0.05	-0.22			
Tame D Ha	e effect of knee braces r	reduces ces.	the opti	mum pi	itch by	3° com	pared	with the	מחות הומות

801 Table 6. Optimum solution for Frame B with variable topology.

837 The real-coded GA was developed to minimize the cost per unit length of a cold-formed steel 838 portal frame building with different frame spacings and topologies, based on the Australian code of practice for cold-formed steel. The influence of a knee brace at the eaves on the optimum topology 839 and unit cost was also considered. The real-coded GA program has been developed to determine 840 the optimum topology and the most suitable cross-sections for members simultaneously. The 841 842 frame obtained from the proposed algorithm can be considered as the most economical design 843 in each case, since the critical design constraint in all examples becomes active. The reliability 844 and robustness of the algorithm have been demonstrated. In addition, the high consistency of the optimum results was achieved through a number of trials for minimizing the objective function. 845 846 However, the time for solving the optimization problem is still large.

847 In the case of optimization problems having many design variables, both optimum and near848 optimum solutions were obtained. Through five design examples, it was shown that the real-coded
849 GA is very efficient in handling the optimization problems having both continuous and discrete
850 variables. Although the cost is calculated based on material used for the main frames, the reduction

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851 is very remarkable when the optimum topology is reached. It is also shown that frames having 852 knee braces result in the most optimum design with the least cost of material.

853 In further research, the position of the purlins and side rails and their cost will be consid-854 ered, taking into account buckling of the column and rafter members between points of lateral restraint. Techniques to enhance real-coded GA to reduce the computing time will also be 855 856 considered.

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