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An Improved Shape Annealing Algorithm for Truss Topology Greneration

Giridhar Reddy and Jonathan Cagan

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An Improved Shape Annealing Algorithm For Truss Topology Generation - with Detailed Examples

Giridhar Reddy and Jonathan Cagan Department of Mechanical Engineering Carnegie Mellon University Pittsburgh, PA

ABSTRACT

An improved shape annealing algorithm for truss topology generation and optimization, based on the techniques of shape grammars and simulated annealing, is introduced. The algorithm features a shape optimization method using only simulated annealing with a shape grammar move set; while no traditional gradient-based techniques are employed, the algorithm demonstrates more consistent convergence characteristics. By penalizing the objective function for violated constraints, the algorithm incorporates geometric constraints to avoid obstacles. The improved algorithm is illustrated on various structural examples taking into account stress, Euler buckling and geometric constraints, generating a variety of solutions based on a simple grammar.

1 INTRODUCTION

Structural topology optimization has become an important research area. Consistent with shape optimization, the structural topology optimization problem is to minimize an objective function, such as the weight or the cost of the structure, subject to a set of structural constraints: stress constraints, buckling constraints, geometric constraints such as location of loads, support points, and obstacles, as well as a variety of other design criteria. Beyond shape optimization (selection of cross-sectional areas and node locations) of the configuration, the problem becomes the generation the optimal topology itself; shape optimization of an inferior topology is an inefficient use of resources and so the best configuration of members to shape optimize is desired.

This paper focuses on truss topology optimization. Some limited analytical approaches go back to early in the century (Michell, 1904). In recent years, various research efforts have attempted to use numerical approaches to generate truss topologies. Three such classes of approaches can be classified as ground structures, emergent material distribution structures, and heuristically produced structures. Of the ground structures, a highly connected grid of members is assumed in which members are removed (*e.g.*, Hemp, 1973), or a grid of permissible nodal points are assumed upon which members are generated (*e.g.*, Dora, *et al.*, 1964; Pederson, 1992; Achtziger, *et al*, 1992), often, but not always, using a linear programming approach. See Kirsch /IQROY for a thnraiurh discussion of these annroaches. These efforts slilLlimiLjiermissible

used. Thus non-convex constraints can be used and geometric obstacles avoided. 2) The simulated annealing algorithm penalizes violated constraints during the optimization process until the design converges on a feasible, optimally directed shape and topology. Because a complete shape optimization is not required at each iteration the computational time is greatly reduced. 3) Within the grammar, two sets of shape rules are used: one which changes the geometry of the truss and the cross-sectional areas of the truss members for shape optimization along with one for topology modifications. The rule sets work in conjunction with each other and consistently make load bearing changes to the design. 4) The shape space and artifact space are now unified and stretching is eliminated; each design generated is already a valid truss (although constraints may still be violated). 5) The new shape grammar makes smaller changes to the design in each iteration, keeping the objective function in the same neighborhood so that the designs can more consistently converge to the same class of structures (i.e., designs have more consistent final objective functions). Each of these changes result in better efficiency, consistency, and performance of the algorithm. Note, however, that due to the remaining topology jumps and large space of possible solutions, repetitive convergence to the same configuration is still not likely. Further, as the complexity of the design constraints increase (e.g., buckling along with gross geometric obstacles) along with the topology jumps, convergence and consistency still decreases; as is typical with simulated annealing, the algorithm can still get stuck in a local optimum or not converge. Yet even for complex geometric obstacles, shape annealing is able to eventually find a feasible solution. The approach generates a variety of design topologies of relatively equal quality (based on the objective function evaluation) each of which could solve a given problem.

2 IMPROVED SHAPE ANNEALING ALGORITHM FOR TRUSS GENERATION

This section describes the improved shape annealing algorithm for truss generation. Two sets of rules are used; one permits small moves within a simulated annealing algorithm for shape optimization while the other makes larger moves by modifying the topology; each rule set works in conjunction with the other forming a grammar which describes the language of permissable trusses. Simulated annealing is used to optimize both the topology and shape. Constraint violations are penalized in the objective function; as the solution converges to the optimum, it also converges to a feasible design and the penalties go to zero. This approach to optimization allows the algorithm to avoid geometric obstacles. It also leads to more consistent convergence properties.

2.1 The Shape Grammar

2.1.1 Shape Optimization Rules

In this algorithm shape optimization occurs through simulated annealing as opposed to

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moves are required; over time the move size decreases along with the neighborhood. We introduce a shape rule set, illustrated in Figure 2, to shape optimize 2-dimensional trusses through shape annealing. Size modification rules, Figure 2a, increase or decrease the cross-sectional area of a member. Shape modification rules, Figure 2b, move a node in the truss to a different location. If a size modification rule is applied, one of the truss members is selected and its cross-sectional area is either increased or decreased, with equal probability, by a default amount which starts from a finite preset value for the initial iteration and reduces to a value near zero as the iterations progress to the final iteration; this decrease in step size allows the annealer to decrease the neighborhood over time. If a shape modification rule, Figure 2b, is applied one of the nodes in the truss is selected and moved in a randomly selected direction by a default value which again starts from a finite preset value for the initial iteration, and reduces to a value close to zero by the final iteration. reducing the neighborhood as well. Size and shape modifications move the truss to an optimum geometry through the annealer. For both modifications, the magnitude of the change distance decreases with the annealing temperature; thus the initial modifications can be quite large and random and, as the annealer becomes more deterministic, the moves become much smaller. Note that these rules are continuous rules in that, over a successive number of moves, the design state can repeat a previous state; thus no reversal rules are required.

3.1.2 Topology Rules

The topology modification rules take a different form than the shape and sizing rules. Topology modifications may create disturbances in the objective function by changing the basic configuration of the truss. Here the neighborhood does not decrease over time. Although the grammar presented in Reddy and Cagan (1994a) produces good truss designs, there are configurations which cannot be produced through that grammar; further, that grammar only modifies the shape in a serial manner off the last point in the chain of triangles. Each time a modification takes place the shape has to be reconnected to the load and anchor points and completely shape optimized. A better approach would have topology modifications occur between shape modifications based on the current shape of the truss; from that point shape optimization continues on the modified topology.

The current grammar takes this approach. Dividing and adding rules are used as shown in the left side of Figures 2c and 2d. A dividing rule modifies the topology by dividing an existing triangle in the design into two new triangles as shown in the left side of Figure 2c. If a dividing rule is to be applied, a random number generator is used to select one of the three sides of the triangle on which to modify the topology. An additive rule is applied at any of the fixed nodes; a fixed node is a node whose geometric location can not be changed (such as a load or anchor point) as illustrated by the label ''•'' in Figure 2d. If an additive rule is applied to the truss, a fixed node is randomly selected (node 1 in the left of Figure 2d) and then another node is randomly selected from the adjacent nodes to that fixed node (node 2 in the figure). An additive rule modifies the truss topology by adding a new triangle at the fixed node. Node 1 is disconnected from the fixed node location and is moved by a preset distance. A new node, node 4 in the figure, is connected to the fixed node location and a new triangle, 1-2-4, is formed. Figure 3 shows a truss to which a dividing rule is applied and one to which an additive rule is applied.

In the exploration of the design space, inferior designs may be pursued, with the expectation that they may lead to superior designs. In order to reverse the exploration when the solution path is abandoned, reversal rules are defined for both the dividing and adding rules as shown in the right side of Figures 2c and 2d; *i.e.*, a rule can be applied from left-to-right or from right-to-left. Thus the effect of any topology modification rule can be nullified by applying the partner rule for the reverse direction³. Note that there is no requirement that a rule be applied only at the last point of rule application; thus the grammar is not limited to serial shape designs as required in the 1994 paper and Cagan and Mitchell's original paper. Although there is no guarantee for completeness of the truss solutions generated from this grammar, the grammar can generate any feasible truss that the original grammar was able to generate.

3.1.3 Grammar Probabilities

The topology modifications may result in large disturbances in the design and evaluation of the objective function. Over time, the shape and sizing modifications result in much smaller disturbances in the objective function. If an annealer spends effort on shape optimization and then makes a change in topology, much of the effort from the shape optimization in the area where the topology changes is lost; yet, any given topology can have a large range of objective function values for various sizes and dimensions and some indication of the final dimensions must be used to accurately evaluate a topology. Once the optimal topology is determined then the structure need only be shape optimized and no further topology modifications should be applied. Thus the probability of selecting shape and sizing rules versus topology rules should change as the algorithm progresses; toward the start there should be a high probability of selecting a topology modification rule but as time progresses that probability should drop off considerably. At the same time the probability of selecting a shape and sizing rule should increase as the algorithm progresses. An adjustable probability is associated with each rule. Presently, a starting probability of 0.1 is used for topology modifications so that roughly 1 in 10 moves will attempt to modify the topology; the probability is reduced with every iteration so as to reach zero by the final iteration. A probability of 0.45 is used for shape modifications. The probability of a size modification starts at a value of 0.45 and increases linearly to 0.55 at the final iteration. Note that the sum of the probabilities equals 1.0 at any time; our experience finds the current distribution to be effective in both generating topologies and shape optimizing the trusses.

3.2 The Algorithm

The improved shape annealing algorithm requires an initial connected design which provides a path between the loads and the anchor points; the minimum such design of exhaustive non-

³The shape grammar is required to have a corresponding reversal rule for every discrete shape rule; continuous rules, or rules that discretize a continuum, can generate any solution along its continuum of allowable moves and require no reversal rules.

intersecting connections is currently input by the user (*e.g.*, left side of Figure 3). Note that the process of stretching is eliminated from this algorithm and the algorithm starts with a fully connected initial shape. All the truss members in the initial design are assigned a default cross-sectional area; any new member added to the design during topology modifications is currently also assigned the same default value.

The shape annealing algorithm, like the simulated annealing algorithm, involves a large number of iterations; every iteration involves a shape rule application and design evaluation as follows: A shape rule is randomly selected according to the probabilities discussed in Section 3.1.3. If a topology modification rule is selected, then the choice between dividing and adding rules as well as the direction of application (left-to-right or right-to-left) is randomly determined. Once a shape rule has been applied, the design is analyzed and the objective function evaluated. As is typical with simulated annealing, at each iteration if the modified design is better in its objective function than the previous design it is accepted as the new design; if it is worse in the objective function then it is accepted with a probability that decreases as the algorithm progresses. This process iterates until convergence is achieved or a limit in number of iterations is reached. The best design generated during the process is also saved.

All analysis is performed through a finite element analysis of the structure; currently only stresses, Euler buckling, and geometric constraints are imposed, although others can easily be added. Note that buckling constraints impose a minimum area on a member; to allow an element in the truss to disappear from the final design through optimization, the buckling constraint is not applied to a member with a cross-sectional area less than a predefined limit. In the final design all remaining elements with area less than that limit are removed, unless they are both required for stability and do not buckle. Obstacles are geometric constraints tested through intersection of the members with the obstruction.

The objective function is evaluated based on the FEM results. The design objective function is calculated. The total objective function is increased based on violations in the design constraints. A penalty is added to the objective function for each constraint violation; the greater the violation of the constraint the larger the penalty added:

totaLobjective = design_objective + £(constraint_violation)*penalty.

As the annealing process progresses, the design is pushed from an infeasible state to a feasible state. For this implementation, the mass of the structure is selected as the design objective function, and the penalty is defined as:

penalty = design^ objective
$$\left(1 - e^{-\frac{-\frac{-iteration}{total_iterations}}{}}\right)$$
.

constraint_violation is the sum of the values of all the violated stress, buckling and geometric constraints. The penalty is set dynamically as a function of the design objective of the current design and the proportion of iterations completed. The penalty is designed so that the constraint violations have little effect during the initial iterations and increasing effect as the iterations progress until it is of the same order of magnitude as the design objective. For this implementation a value of 10 is presently used for K; the larger the value of K selected the larger the rate of increase of the penalty. Any other constraints such as displacement constraints, constraints limiting the cross-sectional area, and frequency constraints could also be incorporated at this stage by assigning appropriate penalties for constraint violation.

Note the effect of this new approach to performing shape annealing. In the 1994 paper, after any iteration, a feasible, shape optimized structure was generated. In the current approach, after each iteration, the structure is not guaranteed to be feasible or shape optimized; it is possible that the only design that will be both feasible and shape optimized is the final design. It is through the simulated annealing optimization that the structure pushes out of the infeasible region. This approach is similar to that used in VLSI layout with simulated annealing: As components are laid out they are allowed to overlap, causing a penalty on the objective function; the annealing process pushes the components away from each other until the overlap disappears (Jepsen and Gelatt, 1983).

4. EXAMPLES

The success of the method is now demonstrated through a series of examples. In the first example, a bridge problem, the algorithm determines a solution for a fairly common truss problem. In the second example, a single load truss problem, the results demonstrate the ability for shape annealing to laterally explore the design space with good convergence. In the third example, the same as the single load truss formulation but with obstacles, the ability of the method to use geometric constraints is shown. All examples consider stress and Euler buckling constraints. In each example note the variety of solutions generated and the ability of the algorithm to mold the design based on the constraints. Each example runs for up to 100,000 iterations; convergent designs run in fewer iterations. All structures are optimized with simulated annealing; solutions without geometric obstacles have been verified by optimizing the topology generated by shape annealing with traditional gradient-based shape optimization. Note that in those problems that were also solved using the 1994 algorithm, the current algorithm solves the problem, with generally better solution objective function values, in one fifth the time.

4.1 Bridge Problem

The problem is to construct a bridge to support three loads anchored at the two support points shown in Figure 4a. The material properties are: Young's modulus of 30×10^6 psi, allowable stress of 35000 psi, and density of 0.286 lb/in³. As the problem input is symmetric, a symmetric

design is expected. The algorithm when applied to the problem produces the designs in Figure 4b and 4c with masses of 2911 lb and 2477 lb, respectively, when buckling is ignored. Note that the design of Figure 4b is expected for a bridge (with the structure above the loads) and was found through one run of the algorithm, however the design of Figure 4c (with the structure below the loads) is a lighter design found by the algorithm. When Euler buckling is included, the design of Figure 4d is found with mass of 3374 lb. In these examples, the thickness of the lines is proportional to the thickness of the members. Node locations and bar cross sectional areas are found in Tables la - lc, correpsonding to Figures 4b - 4d, in the Appendix.

4.2 Single Load/10-Bar Truss Specifications

The classic 10-bar truss, single load problem specifications are as shown in Figure 5a. The material properties are: Young's modulus of 10. X 10⁶ psi, allowable stress of 25000 psi, and density of 0.1 lb/in³. The shape optimized 10-bar truss topology produces designs of mass 1117 lb without buckling and 4693 lb when Euler buckling is included. The original algorithm-when applied to the same specifications produces a 16 bar topology mass of 1077 lb without buckling and a 12 bar topology mass of 2711 lb with buckling criteria. The new algorithm presented in this paper, when applied to the same specifications, produces the, designs in Figure 5b & 5c without buckling, both 8 bar trusses with masses of 1068 lb and 1084 lb respectively, and the designs in Figure 5d & 5e with Euler buckling, a 12 bar truss of mass 2028 lb and an 14 bar truss of mass 2128 lb respectively. The new algorithm produces designs with similar masses, but different topology, to the one produced by the original algorithm when buckling is ignored; however, it produces clearly superior designs when Euler buckling is included. Note that although Figure 5e shows a topology with similar characteristics to that generated by the original algorithm, the topology of the truss in Figure 5d is quite different with compression bars held by string-like tension bars to the upper support point; the design is lighter as well. Note that the shape annealing algorithm proposes significantly different and superior topologies to the 10 bar topology proposed by the designer in this scenario. Node locations and bar cross sectional areas are found in Tables 2a - 2d, correpsonding to Figures 5b - 5e, in the Appendix.

<u>4.3 Single Load Specifications with Obstacles</u>

Three different obstacles are now placed within the specifications for the single load truss problem given above. The first, a single rectangular object, is placed as shown in Figure 6a; the algorithm finds solutions shown in Figures 6b and 6c with masse&of 1215 lb and 1255 lb without buckling, and Figure 6d with buckling having a mass of 4635 lb⁴. A larger obstacle shown in

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⁴Obstacle intersection in the figures is due to the exaggerated line thickness; the actual member does not intersect the obstacle.

Figure 7a, making the lower support point difficult to reach, leads to shape annealed solutions shown in Figures 7b without buckling having a mass of 1985 lb, and Figure 7c with buckling having a mass of 5803 lb. Note how the truss in Figure 7b moves behind the supports to gain more space to work with. Figure 8a demonstrates a tall obstacle, making the path between the load and either support difficult to connect; shape annealing negotiates the obstacle, generating the crane-like design shown in Figure 8b without buckling having a mass of 1833 lb. For the obstacle of Figure 8a with buckling included, shape annealing, after several tries, is able to generate the truss shown in Figure 8c having a mass of 6427 lb. Simulated annealing starts by violating the obstacles and pushes the solution into the feasible space. Node locations and bar cross sectional areas are found in Tables 3a-3c, 4a-4b, and 5a-5b, correpsonding to the solutions in Figures 6-8, in the Appendix.

CONCLUDING REMARKS

Both the original shape annealing algorithm for truss generation and the algorithm introduced in this paper demonstrate the feasibility of shape annealing to generate truss structures. Of interest is the large variety of feasible structures the algorithm is able to generate from the simple shape grammar for classes of objective function values. There are three major directions this research is taking:

1. One direction is the development of efficient grammars. By investigating existing truss configurations (indicating designers¹ preferences) and the optimal designs generated by shape annealing, we expect to further refine the current grammar. In addition, some understanding of the completeness of the grammar would be useful. Due to the generality of the method, other elements (frames and plates) and three-dimensional structures can be modeled with the shape grammars and used within shape annealing to generate more interesting structures.

2. Convergence of the algorithm is still incomplete. For designs without buckling and geometric obstacles, the convergence is excellent; when buckling or simple obstacles are included convergence is good; when both buckling and simple obstacles are included together, convergence is fair; when buckling and gross geometric obstacles are included, convergence and consistency of the algorithm is poor, although a solution can be found. In almost every run a feasible solution is found even though the algorithm does not always converge on the best solution. By examining the best solutions found these convergent characteristics can be seen. Table 1 shows the mean and standard deviation of the best solutions for 20 runs of the algorithm for the single load specification with and without obstacles; the medium obstacle (Figure 6) and 2-block obstacle (Figure 7) with buckling did not find a feasible solution each run resulting in only 16 data points each. Note that the medium obstacle converges with less consistency due to the tight space near the bottom support.

Solution of the more complex problems is difficult; however, we believe consistent convergence can be obtained. One approach to improving the algorithm is improving the annealing

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schedule, as discussed below. Another aspect of the problem is the jump in the objective function due to the topology modifications; simulated annealing works best with smooth changes in the objective function as the solution is perturbed. We are exploring topology modification rules which further decrease the jump in objective function value. Completeness of the algorithm also requires additional analysis as well as more thorough structural stability calculations.

3. The other aspect of the algorithm which greatly affects the quality of the solutions is the annealing schedule. This implementation uses a simple schedule where the temperature is reduced by a fixed amount at each iteration. In related work we have found that dynamic annealing schedules (Huang, *et aL*₉1986) and move sets from the VLSI community can greatly improve the quality and consistency of the annealing solutions (Szykman and Cagan, 1994). However, due to the jumps in the objective function as the topology changes, this problem may not be as well suited for such standard schedules. An annealing schedule better suited for this problem is being investigated.

Table 1. Means (top) and standard deviations (bottom), in lb, of runs over single load specifications.

no	no	small	small	medium	medium	tall	tall
obstacles-	obstacles-	obstacle-	obstacle-	obstacle-	obstacle-	obstacle-	obstacle-
stress	stress &	stress	stress &	stress	stress &	stress	stress &
only	buckling	only	buckling	only	buckling	only	buckling
1104	2923	1358	4870	2270	15171	2013	12599
4	477	66	1146	411	6848	228	5197

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REFERENCES

Achtziger, W., M. Bends0e, A. Ben-Tal, and J. Zowe (1992), "Equivalent Displacement Based Formulations for Maximum Strength Truss Topology Design¹¹, Impact of Computing in Science and Engineering, 4:315-345.

Anagnoustou, G., E.M. R0nquist, and A.T. Patera (1992), "A Computational Procedure for Part Design", *Computer Methods in Applied Mechanics*, 97:33-48.

- Bends0e, M, and N. Kikuchi (1988]), "Generating Optimal Topologies in Structural Design using Homogenization Method", *Computer Methods in Applied Mechanics and Engineering*, 71:197-224.
- Bremicker, M, Chirehdast, M, Kikuchi, M., and P.Y. Papalambros (1991), "Integrated Topology and Shape Optimization in Structural Design", *Mechanics of Structures and Machines Int. J.*, 19(4).
- Cagan, J., and W. J. Mitchell (1993), "Optimally Directed Shape Generation by Shape Annealing", *Environment and Planning B*, 20:5-12.
- Chapman, C, K. Saitou, and M.J. Jakiela (1993), "Genetic Algorithms as an Approach to Configuration and Topology Design", published in proceedings: DE-Vol 65-1, *Advances in Design Automation*, ASME, Albuquerque, NM, September 19-22,1993, Vol 1, pp. 485-498.
- **Diaz, A.R., and B. Belding (1993), 'On Optimum Truss Layout by a Homogenization Method'',** *Transactions of the ASME Journal of Mechanical Design,* 115:367-373:
- Dorn, W.S., R.E. Gomory, and H.J. Greenberg (1964), "Automatic Design of Optimal Structures", *Journal de Mécanique*, 3(1):25-52.
- Hemp, W.S. (1973), Optimum Structures, Clarendon, Oxford.
- Huang, M. D., F. Romeo, and A. Sangiovanni-Vincentilli (1986), "An Efficient Cooling Schedule for Simulated Annealing", *Proceedings of 1986 IEEE International Conference in CAD*, Nov. 1986, pp. 381-384.
- Jepsen, D.W. and CD. Gelatt, Jr. (1983), "Macro Placement by Monte Carlo Annealing," Proceedings of the IEEE International Conference on Computer Design, November, pp. 495-498.
- Kirkpatrick, S., C. D. Gelatt, Jr., and M. P. Vecchi (1983), "Optimization by Simulated Annealing", *Science*, 220(4598):671-679.
- Kirsch, U. (1989), "Optimal Topologies of Structures", Applied Mechanics Review, 42(8):223-238.
- Lakmazaheri, S., and W.J. Rasdorf (1990), "The Analysis and Partial Synthesis of Truss Structures via Theorem Proving", *Engineering with Computers*, 6:31-45.
- Michell, A.G.M. (1904), "The Limits of Economy of Materials in Frame Structures", *Philosophical Magazine*, S.6, 8(47):589-597.
- Papalambros, P. Y., and M. Chirehdast (1990), "An Integrated Environment for Structural Configuration Design", *Journal of Engineering Design*, 1(1):73-96.
- Pedersen, P. (1992), "Topology Optimization of Three Dimensional Trusses", *Topology Design* of Structures, NATO ASI Series - NATO Advanced Research Workshop (Bends0e, M.P. and C.A. Mota Soares, eds.), Sesimbra, Portugal, June 20-26, Kluwer Academic Publishers, Dordrecht.
- Reddy, G., and J. Cagan (1994a), "Optimally Directed Truss Topology Generation Using Shape Annealing", to be published: *ASME Journal of Mechanical Design*.
- Reddy, G., and J. Cagan (1994b), "An Improved Shape Annealing Method For Truss Topology Generation," proceedings of: ASME Design Theory and Methodology Conference, Minneapolis, MN, September 11-14.
- Reddy, G., and J. Cagan (1994c), "An Improved Shape Annealing Method For Truss Topology Generation," *EDRC Report 24~***-94*, Engineering Design Research Center, Carnegie Mellon University, Pittsburgh, PA 15213.
- Rodrigues, H.C., and P.A. Fernandes (1993), "Generalized Topology Optimization of Linear Elastic Structures Subjected to Thermal Loads", published in proceedings: DE-Vol 65-1, *Advances in Design Automation*, ASME, Albuquerque, NM, September 19-22,1:769-777.
- Rogers, J., S. Feycock, and J. Sobieszczanski-Sobieski (1988), "STRUTEX- A Prototype Knowledge Based System for Initially Configuring a Structure to Support Point Loads in Two Dimensions", in Artificial Intelligence in Engineering: Design (Gero, J.S., ed.), Elsevier publications, Amsterdam, pp. 315-335.
- Shah, J. J. (1988), "Synthesis of Initial Form for Structural Shape Optimization", Journal of Vibration, Acoustics, Stress, and Reliability in Design, 110:564-570.

Spillers, W. (1985), "Shape Optimization of Structures", in: *Design Optimization* (Gero, J.S., ed.), Academic Press Inc., Orlando, pp. 41-70.

Stiny, G. (1980), "Introduction to Shape and Shape Grammars", *Environment and Planning* 5, 7:343-351.

Szykman, S., and J. Cagan (1994), "A Simulated Annealing Approach to Three Dimensional Component Packing", accepted in: *ASME Journal of Mechanical Design*.



Figure 1: Example shape space (la) and analogous artifact space (lb) trass generation from original 1994 algorithm.

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Figure 2: Shape grammar for sizing (2a) shape (2b) and topology (2c: dividing and 2d: adding) topology modifications.

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Figure 3: Application of dividing and adding topology modification rules.





Figure 4: Bridge example: input (4a), solutions without buckling (4b and 4c), solution with Euler buckling (4d).

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(a)



(b)



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Figure 5: Single load truss example: input (5a), solutions without buckling (5b and 5c), solutions with Euler buckling (5d and 5e).

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Figure 6: Single load truss with obstacle 1: input (6a), solutions without buckling (6b and 6c), solution with Euler buckling (6d).

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(a)



Figure 7: Single load truss with obstacle 2: input (7a), solution without buckling (7b), solution with Euler buckling (7c).

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(a)



Figure 8: Single load truss with obstacle 3: input (8a), solution without buckling (8b), solution with Euler buckling (8c).

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APPENDIX

Table 1a: Bridge with mass of 2911 lb (No Buckling) corresponding to Figure-4b

Youngs Modulus = $3.0e^7$ psi Yield stress = 35000 psi Density = 0.2861b/in³

Node No.	X Coordinate (in)	Y Coordinate (in)
0	250.0	339.9
1	500.0	0.0
2	750.0	0.0
3	750.0	339.0
4	0.0	50.0
5	1000.0	50.0
6	250.0	0.0

Element No.	Starting Node	Ending Node	Area(iin ²)
0	0	1	1.78
1	1	3	1.78
2	0	3	4.76
3	2	3	2.86
4	0	4	5.67
5	3	5	5.67
6	0	6	2.86
7	4	6	0.01
8	1	6	0.01
9	1	2	0.01
10	2	5	0.01

Table 1b: Bridge with mass of 2477 lb (No Buckling) corresponding to Figure-4c

Youngs Modulus = $3.0e^7$ psi Yield stress = 35000 psi Density = 0.286 lb/in³

Node No.	X Coordinate (in.)	Y Coordina te (in)
0	250.0	-200.0
1	250.0	0.0
2	500.0	-286.8
3	750.0	0.0
4	750.0	-200.0
5	500.0	50.0
6	0.0	0.0
7	1000.0	50.0

Element No.	Starting Node	Ending Node	Area (in ²)
0	0	1	2.86
1	1	2	0.01
2	0	2	4.36
3	2	4	4.36
. 4	0	4	0.17
5	2	3	0.01
6	3	4	2.86
7	1	5	0.01
8	2	5	2.86
9	1	6	0.01
10	0	6	6.06
11	3	7	0.01
12	3	5	0.01
13	4	7	6.06

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Table 1c: Bridge with mass of 3374 lb (With Buckling) corresponding to Figure-4d

Youngs **Modulus** = $3.0e^7$ psi **Yield** stress = 35000 psi Density = **0.286 lb/in³**

Node No.	X Coordinate (in)	Y (Coordinate (in)
0	0.0	0.0
1	1000.0	0.0
2	250.0	-50.0
3	500.0	-208.0
4	750.0	-166.0
5	750.0	-50.0
6	250.0	-166.0
7	500.0	-50.0

Element No.	Starting Node	Ending Node	Area(i n²)
0	0	6	6.92
1	6	3	5.84
2	3	0	1.20
3	2	6	7.56
4	3	4	5.84
5	4	1	6.92
6	1	3	1.20
7	4	5	7.56
8	7	3	10.29
9	3	2	0.01
10	3	5	0.01
11	7	2	0.01

Table 2a: Single Load with mass of 1068 lb (No Buckling) corresponding to Figure-5b

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = 0.1 lb/in³

Node No.	X Coordinate (in)	Y <coordinate (in)<="" th=""></coordinate>
0	311.0	152.8
1	389.4	-77.8
2	720.0	0.0
3	0.0	360.0
4	0.0	0.0
5	423.4	292.1

Element No.	Starting Node	Ending Node	Area (ii 1²)
0	0	1	1.61
1	1	2	3.37
2	2	5	4.60
3	0	5	3.08
4	0	3	3.35
5	3	5	5.28
6	0	4	4.68
7	1	4	3.87

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Table 2b: Single Load with mass of 1084 lb (No Buckling) corresponding to Figure-5c

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = 0.1 lb/in^3

Node No.	X Coordinate (in)	Y Coordin ate (in)
0	0.0	360.0
1	0.0	0.0
2	427.3	285.3
3	303.7	141.4
4	578.1	-41.8
5	720.0	0.0

Element No.	Starting Node	Ending Node	Area (in ²)
0	2	3	4.15
2	0	3 2	6.50
3	0	3	1.97
4	1	4	3.71
5	2	4	1.31
6	2	5	4.41
7	4	5	3.29

Table 2c: Single Load with mass of 2028 lb (With Buckling) corresponding to Figure-5d

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = $0.1ib/in^3$

Node No.	X Coordinate (in)	Y Coordinate (in)
0	0.0	360.0
1	116.5	-20.6
2	232.3	-40.5
3	347.2	-55.4
4	0.0	0.0
5	472.6	-50.0
6	594.4	-29.9
7	720.0	0.0

Element 10. Starting 1000 Enung 1000 Are	a (m)
0 0 1 0	0.04
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$).42 3.87
3 0 3	1.69
4 2 3 18 5 1 4 19	3.30
6 0 5	1.05
7 3 5 18	3.33
9 5 6 17	7.03
10 0 7 6	5.06 7 10

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Table 2d: Single Load with mass of 2128 lb (With Buckling) corresponding to Figure-5e

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = $0.1ib/in^3$

Node No.	X Coordinate (in)	Y Coordinate (in)
0 ·	0.0	360.0
1	258.8	-15.6
2	397.7	-5.5
3	135.3	-8.5
4	582.7	114.8
5	0.0	0.0
6	619.3	8.0
7	527.4	20.3
8	720.0	0.0

Element No.	Starting Node	Ending Node	Area(ii n²)
0	0	1	1.20
1	1	2	21.26
2	0	2	1.08
3	0	3	0.05
4	1	3	19.74
5	. 0	4	7.06
6	2	4	0.08
7	3	5	21.65
8	2	7	19.32
9	4	7	9.48
10	4	6	0.32
11	6	7	12.16
12	4	8	6.89
- 13	6	8	13.16

Table 3a: Single Load with obstacle with mass of 1215 lb corresponding to Figure-6b (No Buckling; Rectangular Constraint at location: ((200,-100);(200,100);(500,100);(500,-100)).)

Youngs Modulus ==1.Oe'psi Yield stress = 25000 psi Density = 0.1 lb/in³

Node No.	X Coordinate (in)	Y Coordinate (in)
0	557.4	292.4
1	201.9	100.9
2	720.2	101.1
3	0.0	0.0
4	355.1	361.9
5	0.0	360.0
6	720.0	0.0

Element No.	Starting Node	Ending Node	Area(ii a²)
0	0	1	3.74
1	1	2	3.40
2	0	2	5.23
3	1	3	8.94
4	1	4 .	2.58
5	0	4	6.72
6	0	5	0.36
7	4	5	7.66
8	0	6	0.02
9	2	6	3.99

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Table 3b: Single Load with obstacle with mass of 1255 lbs corresponding to Figure-6c (No Buckling; Rectangular Constraint at location: ((200,-100);(200,100);(500,100);(500,-100)).)

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = $0.1ib/in^3$

Node No.	X Coordinate (in)	\: CoordinateKin)
0	513.1	257.3
1	0.0	0.0
2	509.0	-258.5
3	0.0	360.0
4	720.0	0.0

Element No.	Starting Node.	Ending Node	4.rea(iin²)
0	0	1	7.16
1	1	2	1.79
2	0	2	2.80
3	0	3	8.16
4	0	4	2.58
5	2	4	2.56

Table 3c: Single Load with obstacle with mass of 4635 lbs corresponding to Figure-6d (With Buckling; Rectangular Constraint at location: ((200,-100);(200,100);(500,100);(500,-100)).)

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = $0.1ib/in^3$

X Coordinate (in)	Y Coordinate (in)
0.0	360.0
720.0	355.3
720.0	0.0
464.8	228.1
0.0	0.0
223.5	112.7
	X Coordinate (in) 0.0 720.0 720.0 464.8 0.0 223.5

Element No.	Starting Node.	Ending Node	Area(ii n²)
0	1	2	4.00
1	1	0	7.92
2	0	3	0.22
3	3	1	47.86
4	0	5	8.32
5	5	3	45.31
6	4	5	42.27
7	2	3	0.01

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Table 4a: Single Load with obstacle with mass of 1985 lbs corresponding to Figure-7b (No Buckling; Rectangular Constraint at location: ((100,-100);(100,200);(500,200);(500,-100)).)

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = 0.1 lb/in^3

Node No.	X Coordinate (in)	Y Coordinate (in)
0	407.3	383.7
1	0.0	0.0
2	157.2	-476.4
3	720.0	0.0
4	-283.8	0.0
5	0.0	360.0

Element No.	Starting Node	Ending]Vode	Area(ii n²)
0	2	4	4.13
1	1	4	8.00
2	0	4	5.71
3	0	3	3.05
4	2	3	2.53
5	2	5	4.74
6	0	5	6.93
7	4	5	0.33

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Table 4b: Single Load with obstacle with mass of 5803 lbs corresponding to Figure-7c (With Buckling; Rectangular Constraint at location: ((100,-100);(100,200);(500,200);(500,-100)).)

Youngs Modulus = 1.0e'psi Yield stress = 25000 psi Density = $0.1 \text{ lb/in}^{\cdot J}$

Node No.	X Coordinate (in)	Y Coordinate (in)
0	159.1	409.0
1	508.8	212.7
2	720.0	289.4
3	720.0	0.0
4	299.3	200.0
- 5	100.4	200.0
6	0.0	360.0
7	0.0	0.0

Element No.	Stating Node	Ending Node	Area(ii n²)
0	0	1	3.87
1	1	2	34.45
2	0	2	7.10
3	2	3	4.00
4	0	4	0.75
5	1	4	38.07
6	0	5	35.30
7	4	5	36.77
8	5	6	32.62
9	0	6	13.58
10	5	7	53.32
11	1	3	0.01

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Table 5a: Single Load with obstacle with mass of 1833 lbs corresponding to Figure-8b (No Buckling; Rectangular Constraints at locations: ((200,-100);(200,100);(500,100);(500,-100)).& ((400,100);(400,400);(500,400);(500,100)).)

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = 0.1 lb/in^3

X Coordinate (in)	Y Coordinate (in)	
462.4	711.7	
391.4	400.1	
720.0	492.0	
0.0	360.0	
0.0	0.0	
720.0	0.0	
	X Coordinate (in) 462.4 391.4 720.0 0.0 0.0 720.0	

Element No.	Starting Node	Ending Node	Area(ii n²)
0	0	1	7.07
1	1	2	3.67
2	0	2	4.64
3	1	3	2.91
4	0	3	6.41
5	1	4	11.44
6	2	5	4.00
7	0	5	0.01

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Table 5b: Single Load with obstacle with mass of 6427 lb corresponding to Figure-8c (With Buckling; Rectangular Constraints at locations: ((200,-100);(200,100);(500,100);(500,-100)).& ((400,100);(400,400);(500,400);(500,100)).)

Youngs Modulus = $1.0e^7$ psi Yield stress = 25000 psi Density = 0.1 lb/in^3

Node No.	X Coordinate (in)	$\sqrt{7}$ Coordinate (in)	
0	0.0	360.0	
1	0.0	0.0	
2	405.0	429.7	
3	378.8	360.0	
4	532.8	537.6	
5	720.0	0.0	
6	720.0	408.2	
7	208.5	266.3	
8	115.9	194.1	

Element No.	Starting Node	Ending Node	Area(ij n²)
0	1	8	50.38
1	8	0	27.15
2	7	0	0.50
3	8	7	25.00
4	2	0	3.15
5	7	2	54.00
6	0	3	8.86
7	3	2	22.43
8	2	4	16.90
9	4	3	9.18
10	2	6	45.19
11	6	4	7.81
12	6	5	4.00
13	4	5	0.01

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