

Online Research @ Cardiff

This is an Open Access document downloaded from ORCA, Cardiff University's institutional repository: <https://orca.cardiff.ac.uk/id/eprint/136037/>

This is the author's version of a work that was submitted to / accepted for publication.

Citation for final published version:

Thompson, Jonathan A., Pisano, Giampaolo ORCID: <https://orcid.org/0000-0003-4302-5681> and Tucker, Carole ORCID: <https://orcid.org/0000-0002-1851-3918> 2020. Mesh low-pass filters for millimeter-wave applications: is the square capacitive shape optimal? *Journal of Astronomical Telescopes, Instruments, and Systems* 6 (3) , 036003. 10.1117/1.JATIS.6.3.036003 file

Publishers page: <http://dx.doi.org/10.1117/1.JATIS.6.3.036003>
<<http://dx.doi.org/10.1117/1.JATIS.6.3.036003>>

Please note:

Changes made as a result of publishing processes such as copy-editing, formatting and page numbers may not be reflected in this version. For the definitive version of this publication, please refer to the published source. You are advised to consult the publisher's version if you wish to cite this paper.

This version is being made available in accordance with publisher policies.

See

<http://orca.cf.ac.uk/policies.html> for usage policies. Copyright and moral rights for publications made available in ORCA are retained by the copyright holders.



1 Mesh material low pass filters for millimeter wave applications: is 2 the square capacitive shape optimal?

3 Jonathan A. Thompson^{a,*}, Giampaolo Pisano^a, Carole Tucker^a

4 ^aSchool of Physics and Astronomy, Cardiff University, The Parade, Cardiff UK, CF24 3AA

5 **Abstract.** The use of mesh filters for millimeter wave applications using capacitive and inductive grids is well known
6 and they are widely used in cosmic microwave background instrumentation. We report here on an investigation into
7 whether the capacitive square shape typically used in low pass filter designs, could be improved upon. The micro-
8 genetic algorithm and the finite differences, time domain, electromagnetic modelling method were used to look for
9 shape variations to the standard square shape. Any shape changes discovered were then analysed to establish which
10 variations had the most effect. We shall show that improvements found using pixelated patterns evolved by the genetic
11 algorithm were somewhat mixed.

12 **Keywords:** millimeter waves, FDTD, genetic algorithm, mesh filter.

13 *Jonathan A. Thompson, thompsonja3@cardiff.ac.uk

14 1 Introduction

15 Filters for millimeter electromagnetic waves are an essential part of many radio astronomical in-
16 struments and their implementation using metallic mesh grids is becoming common place. Their
17 design, however, can still be a somewhat involved process and tends to be restricted to a small num-
18 ber of well-known patterns, namely square plates (for capacitive grids) and square holes (inductive
19 grids).

20 The study reported here set out to discover if there were any improvements that could be made
21 to a capacitive mesh made from square patches when used for low pass filters. The investigation
22 procedure consisted of three stages; finding an initial conventional design, improving the design by
23 varying the layer pattern and then decomposing the pattern changes. A genetic algorithm coupled
24 to an electromagnetic modelling method was used in the first two stages.

25 The use of square patch and square hole metal meshes to construct low and high pass filters
26 was first described by Ulrich.¹ Pisano et al² review many device types that have been implemented

27 using such meshes in millimetre wave astronomy instrumentation, while Ade et al³ concentrates
28 on filter applications. A combination of the square patch and hole (often referred to as a cross)
29 has been used to produce a band pass response; Moallem and Sarabandi⁴ and Wang et al⁵ describe
30 examples while Melo et al⁶ reviews the history of its use and mentions other patterns that have
31 been investigated. Split ring resonators have also been used in band pass and band stop filters,
32 Navarro-Cia et al⁷ and Kundu et al⁸ are examples. Wu et al⁹ investigate a fractal pattern to provide
33 a similar response.

34 The use of a genetic algorithm as a search method was first described by Holland¹⁰ and later
35 by Goldberg.¹¹ The Micro-Genetic Algorithm (MGA), an improvement that does not require large
36 numbers of candidate solutions to be evaluated, was proposed by Krishnakumar.¹² Ge and Esselle¹³
37 described using the MGA and the finite differences, time domain (FDTD) modelling method to ex-
38 plore unconventional patterns for reflective surfaces. The patterns were constructed from 16x16
39 pixelated grids with four fold symmetry. Sui et al¹⁴ use this technique to design and construct a fre-
40 quency selective surface absorber and Thompson and Pisano¹⁵ extended the method to find designs
41 for millimeter wave transmission devices. Ranjan et al^{16,17} use a binary wind driven optimisation
42 algorithm with pixelated patterns to design frequency selective surfaces. Mohammed et al¹⁸ com-
43 pare a number of nature inspired optimisation algorithms, including the genetic algorithm, when
44 applied to antenna design. Campbell et al¹⁹ review the general field of numerical optimisation
45 applied to electromagnetic device design.

46 The first stage of the investigation consisted of searching for low pass filter solutions using
47 the genetic algorithm and the propagation matrix electromagnetic modelling method described by
48 Orfanadis.²⁰ Admittance curves were computed for various sizes of conventional square capacitive
49 plate in a standard unit cell. This information was used to derive admittance curves for the desired

50 unit cell and plate size by scaling and interpolation. The propagation matrix method used this in-
51 formation to rapidly model a solution and generate its transmittance and phase shift characteristics.
52 In this way, large numbers of generations of possible solutions can be assessed in a short time to
53 arrive at an initial low pass filter design using the MGA search algorithm.

54 In the second stage, this initial solution was used as a starting point for a further MGA search
55 using the FDTD electromagnetic modelling method, described by Taflove and Hagness²¹ and by
56 Schneider.²² The conventional square plates were converted into NxN pixelated panels, the value
57 of N being chosen so as to represent the initial solution as closely as possible while still keeping
58 the search space and FDTD model reasonably sized. The search then proceeds with the genetic
59 algorithm varying the patterns away from the initial conventional square shapes, while keeping the
60 unit cell and layer spacing fixed at the values found in stage 1. This stage is rather more processor
61 intensive than the first stage and was undertaken on Cardiff University's Hawk compute cluster.

62 The NxN patterns resulting from stage 2 usually showed the basic outline of the starting shapes
63 from stage 1 with a number of changes, for example, corners cut off, extra protrusions, holes etc. In
64 the third stage, changes were identified and tested in isolation to establish their individual effects.

65 A number of constraints were applied to the designs produced by the first stage. They were
66 restricted to 4 layers, two instances of each of two patterns arranged symmetrically. The sizes of
67 the square shapes for the initial conventional designs were required to fit on the FDTD grid used
68 for the second stage pattern search. The unit cell used on each layer was restricted to a simple set of
69 ratios, 1:1, 2:1, to allow the designs to be easily modelled. Since possible improvements due to the
70 layer patterns are being investigated, it does not matter that the initial designs are not necessarily
71 optimal due to these constraints, any improvements found due to shape variation should still be
72 applicable.

73 **2 The Electromagnetic Modelling Methods**

74 The propagation matrix method, used in stage 1, is described in detail by Orfanidis.²⁰ The method
75 is a one-dimensional solution of Maxwell's equations, separating waves into forward and reverse
76 components. Matrices were derived that represent waves travelling through dielectrics (propaga-
77 tion matrices) and across boundaries between different dielectrics (matching matrices). *The treat-
78 ment is easily extensible to cover the metallic meshes that are the subject of this study by using
79 their complex admittance* either from a formula, for example one those collected together by Lee
80 et al,²³ or by using modelling tools to measure it, the approach taken here. Thompson and Pisano¹⁵
81 provide a summary of the modelling method as implemented.

82 The second stage utilised FDTD, a three-dimensional modelling method that calculates the
83 solutions to Maxwell's equations using second order finite differences. Good descriptions of the
84 method are given by Taflove and Hagness²¹ and Schneider.²² The implementation used here is
85 described in Thompson and Pisano¹⁵ and was verified against Ansys HFSS.²⁴

86 **3 The Micro-Genetic Algorithm**

87 The MGA used to perform the searches is described by Krishnakumar¹² and the implementation
88 utilised here is a development of that described in Thompson and Pisano.¹⁵ It proceeds as shown
89 in Fig. 1.

90 To calculate the fitness of each individual, the results from the electromagnetic model are as-
91 sessed against a fitness function. In this study, the transfer functions of two common low pass
92 filter types, Butterworth and Chebyshev, were used. The area between the modelled curve and the
93 fitness function was used as the measure of unfitness.

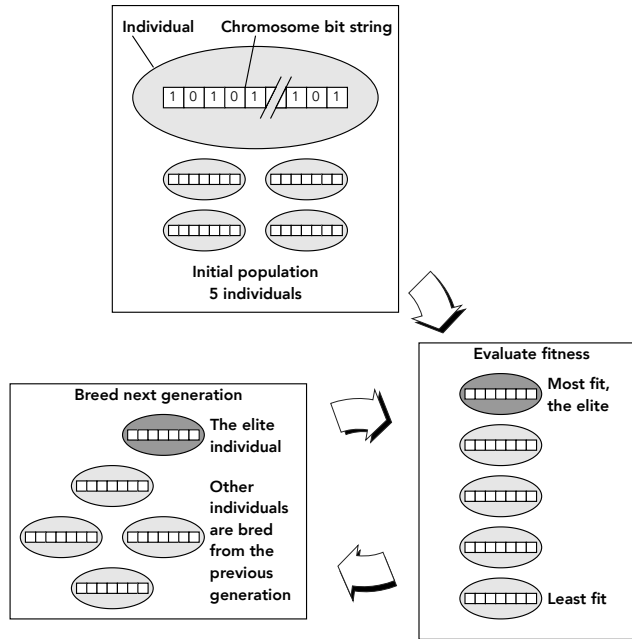


Fig 1 The micro-genetic algorithm. The starting point is an initial population of 5 individuals. Their chromosome bit strings are normally generated randomly, but in stage 2 of the methodology one of the individuals is initialized from the best solution of stage 1. The fitness of each individual is then assessed by running the electromagnetic model on the structure represented by its chromosome and comparing the results with the fitness function. The best individual is marked as the elite and is automatically included in the next generation. Four new members are then bred from the previous generation to bring the numbers back to 5. A check is performed to maintain the genetic diversity of the generation; if all the individuals have more than 95% of their chromosomes identical, the four bred individuals are replaced with completely new random individuals. The loop then continues with fitness evaluation.

94 The patterns used by the MGA to improve the fit of a solution to the desired transfer function
 95 were N by N (N is always even), four-fold symmetric, pixelated plates. The 16 by 16 pixel plate
 96 was used by Ge and Esselle[1] in their study. Here the resolution of the plate is chosen such that
 97 the starting point for stage 2 can be reasonably accurately represented. The disadvantage of higher
 98 resolution plates is that they increase the size of the solution space being searched. To maintain the
 99 four-fold symmetry that all the filter designs described here require, only one triangle of a plate is
 100 actually specified, as shown in Fig. 2.

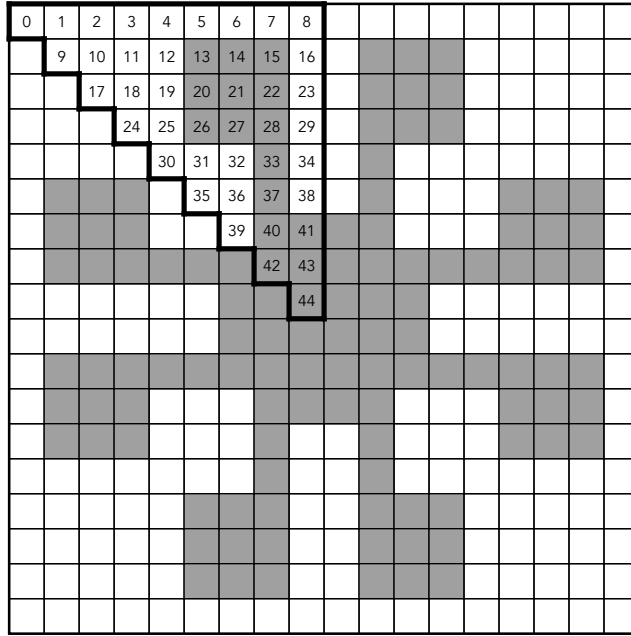


Fig 2 A pixelated plate and its binary encoding. As described by Ge and Esselle¹³ and reiterated in Thompson and Pisano,¹⁵ a four-fold pixelated $N \times N$ plate can be represented by the binary code covering one triangle of the plate as shown in this example of an 18×18 plate. The numbers in the pixels indicate the bit number in the gene encoding.

101 **4 Results**

102 The methodology outlined in the previous sections was used to investigate two low pass filter
 103 designs; a seventh order Chebyshev and a fifth order Butterworth.

104 *4.1 Seventh order Chebyshev low pass filter*

105 A seventh order low pass Chebyshev filter with a pass-band ripple of 0.5dB was specified as the
 106 fitness function for the first investigation. The conventional squares solution found by stage 1 is
 107 shown in Fig. 3. The stage 2 search was then initialised with this information and the layer patterns
 108 found are shown in Fig. 4. The transmittance curves are shown in Fig. 5, showing how the pattern
 109 changes made by stage 2 have brought the response closer to the ideal.

110 The losses incurred by the signal passing through the filter were estimated by measuring both
 111 the transmittance and the reflectance. These are shown in Fig. 6. The losses for both the capacitive

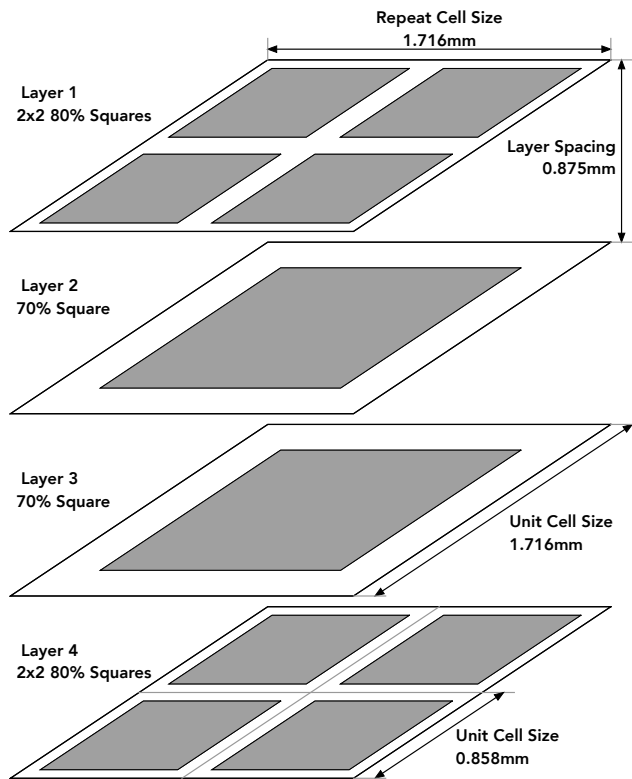


Fig 3 The best solution from stage 1 for the seventh order Chebyshev low pass filter example. It consists of two layers repeated in reverse order with a repeat cell size of 1.716mm and a layer spacing of 0.875mm. Layers 1 and 4 consist of a 2x2 layout of 80% squares each in a unit cell of 0.858mm. Layers 2 and 3 are single 70% squares in a unit cell of 1.716mm covering the entire repeat cell.

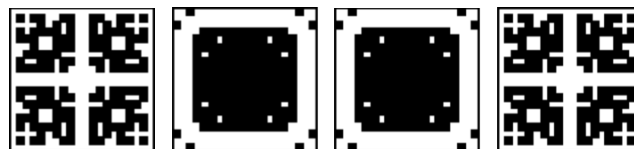


Fig 4 The best solution from stage 2 for the seventh order Chebyshev low pass filter example. These are the 4-fold 20x20 binary patterns settled on for the four layers.

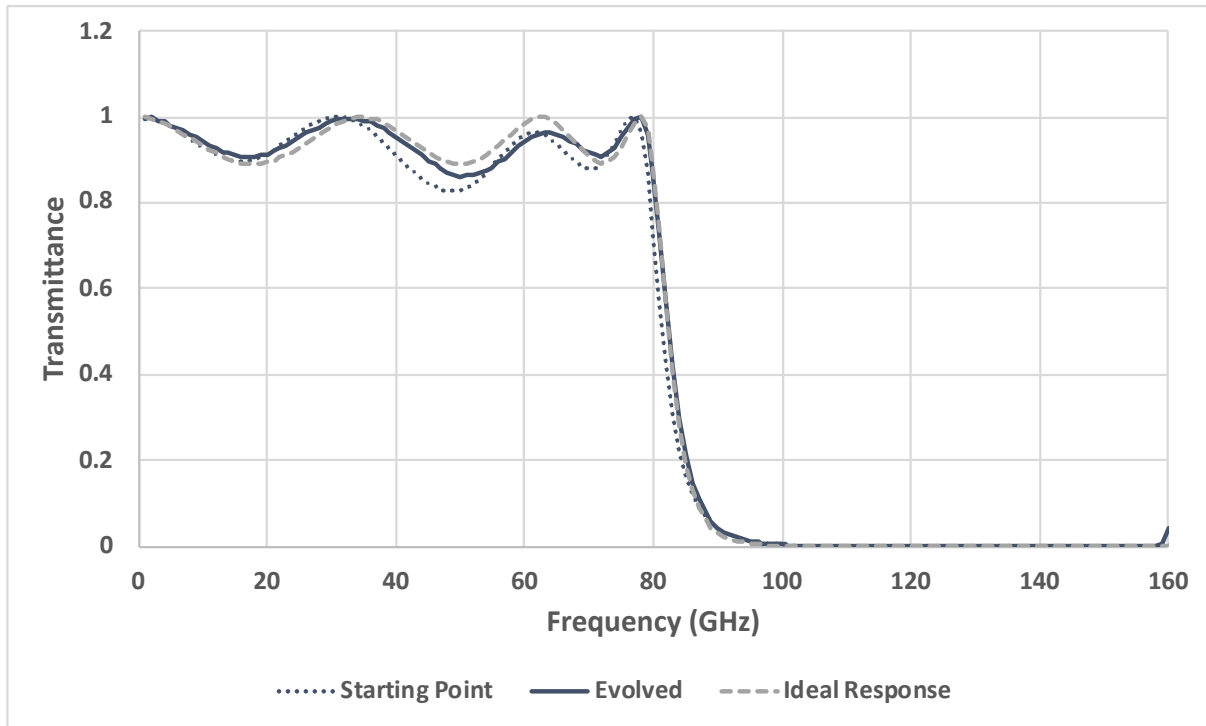


Fig 5 The transmittance curves of the starting point (the best solution from stage 1), the best evolved solution from stage 2 and the ideal filter response. It can be seen that the plate pattern changes from the stage 2 evolution have caused the transmittance characteristic to move closer to the target; most significantly, the pass-band ripple has been reduced.

112 squares pattern produced by stage 1 and the binary pattern produced by stage 2 are pretty much
 113 identical within the passband of the filter and below the diffraction limit, within the accuracy limits
 114 of the FDTD modelling technique. The losses of the evolved design do increase in the stop band.

115 The effect of the various changes made by the genetic algorithm were then investigated in
 116 stage 3. An error value was calculated for the stage 2 result by measuring the area between its
 117 transmittance curve and the target curve. Pattern features introduced by the genetic algorithm were
 118 then removed one by one and the FDTD model run to allow the calculation of an error for each
 119 case. The effect of each pattern feature could then be assessed. All the results of this are shown
 120 in Fig. 7. The conventional capacitive squares solution produced by stage one, returned an error
 121 of 3.19 (the units are GHz, the transmittance being a power ratio). The evolved solution of stage
 122 2 showed an error of 1.68. The evolved solution reduced the error by 47%, most of the change

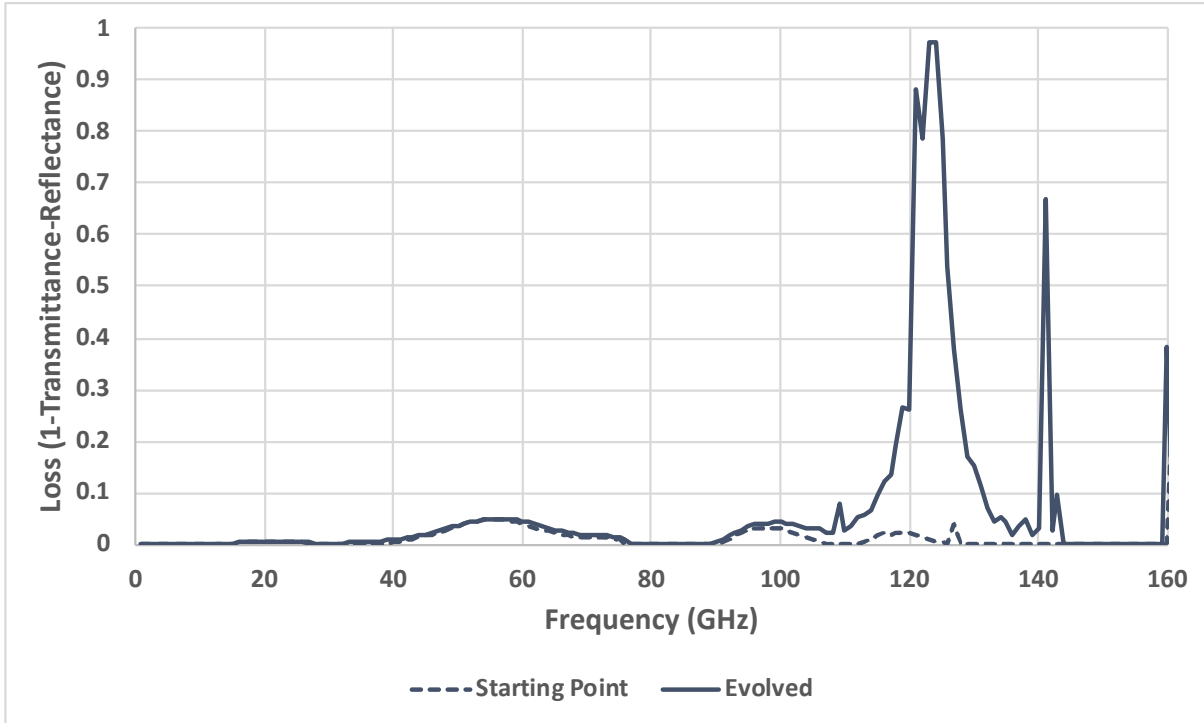


Fig 6 The losses of the starting point (the best solution from stage 1) and the best evolved solution from stage 2 for the Chebyshev filter. The losses of the two solutions are minimal and similar within the passband. In the stop band, however, the evolved solution showed significant loss around 125GHz. The diffraction limit is at 160GHz.

123 coming in the ripple in the passband.

124 4.2 Butterworth low pass filter

125 For the second example, a fifth order low pass Butterworth filter was specified as the fitness func-
 126 tion. To obtain a reasonable Butterworth response, it proved necessary to reduce the upper limit
 127 to 125GHz, thus easing the requirements on the diffraction zone. The result of stage 1 is shown
 128 in Fig. 8. The stage 2 search was initialised with this result, the layer patterns found are shown in
 129 Fig. 9 and the transmittance curves of the stage 1 solution, stage 2 solution and the ideal are shown
 130 in Fig. 10.

131 Using the same error estimating method as for the Chebyshev case, an error reduction of just
 132 8% is achieved by the evolved design over the conventional squares. However, unlike with the

Description	Layers 1&4	Layers 2&3	Comments
Stage 1 result Error: 3.19			The conventional square capacitive plates result
Stage 2 result Error: 1.68 Improvement: 47%			The solution evolved from the conventional result
Layers 2&3 corner dots Error: 1.90 Improvement: 40%			The dots only make a small difference to the error
Layers 2&3 plate holes Error: 1.71 Improvement: 46%			These plate holes make even less difference
Layers 2&3 plate corners not cut off Error: 2.74 Improvement: 14%			This change makes the biggest difference of all the changes made to layers 2&3
Layers 1&4 plate center holes Error: 1.84 Improvement: 42%			The holes in the center of the layer 1&4 plates make only a small difference
Layers 1&4 middle corners Error: 1.97 Improvement: 38%			Only a small difference made by this feature
Layers 1&4 outer corners Error: 2.20 Improvement: 31%			The rounding of the outer corners of layers 1&4 do make a difference

Fig 7 Fitness factors were calculated for the result of stage 2 and for patterns with various stage 2 changes removed. The conclusion is that the corners of layers 2&3 and the outer corners of layers 1&4 make the most difference. The other changes appear to make only a marginal difference.

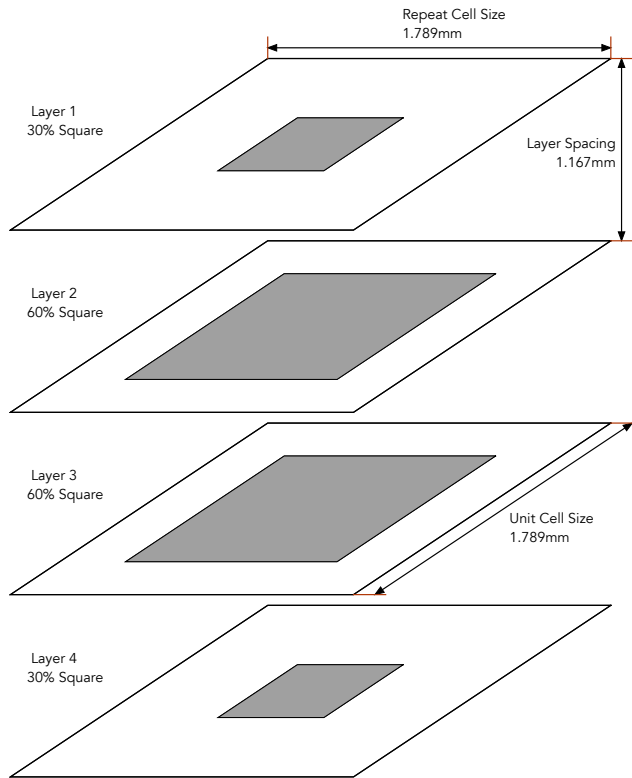


Fig 8 The best solution from stage 1 for the fifth order Butterworth low pass filter example. It consists of two layers repeated in reverse order with a repeat cell size of 1.789mm and a layer spacing of 1.167mm. Layers 1 and 4 consist of a single 30% square. Layers 2 and 3 are single 60%.

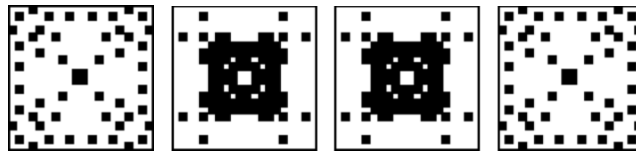


Fig 9 The best solution from stage 2 for the fifth order Butterworth low pass filter example. These are the 4-fold 20x20 binary patterns settled on for the four layers.

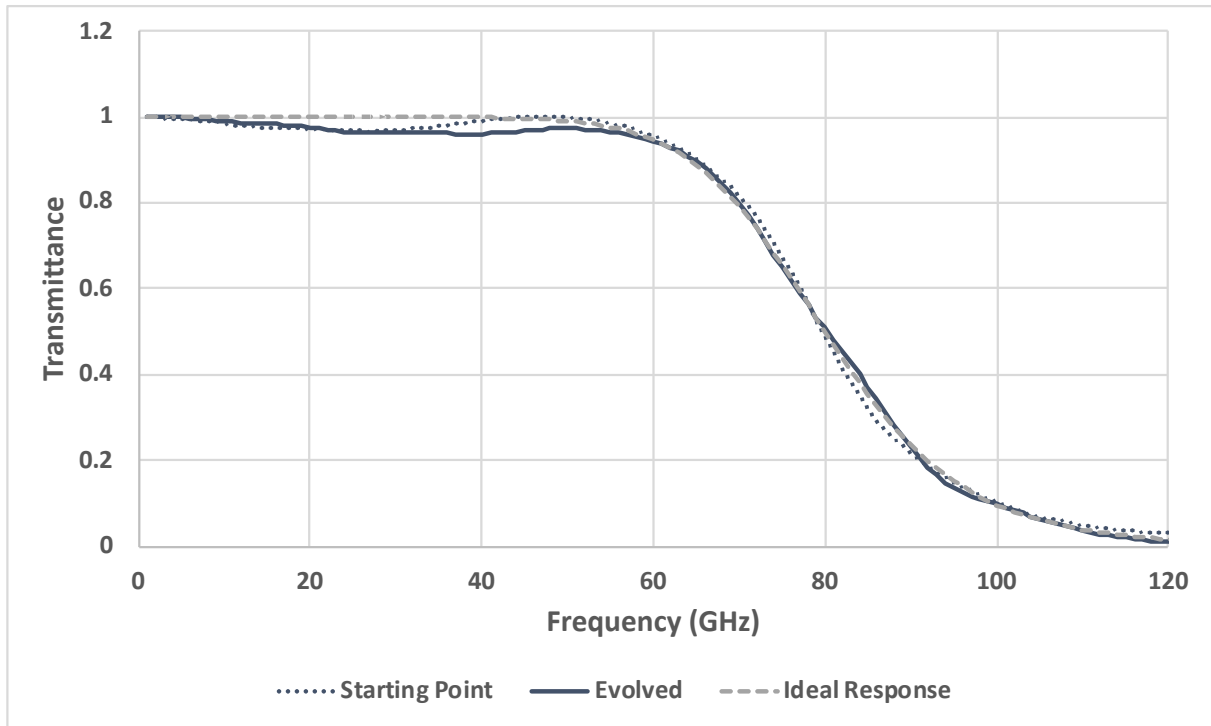


Fig 10 The best solution from stage 2 for the fifth order Butterworth low pass filter example. These are the 4-fold 20x20 binary patterns settled on for the four layers.

133 Chebyshev filter, three quarters of this comes about through a better conformance to the roll off on
 134 the approach to the stop band, arguably an unimportant improvement.

135 The losses incurred by the signal passing through the filter were again estimated, shown in Fig.
 136 11. The losses for both the capacitive squares pattern produced by stage 1 and the binary pattern
 137 produced by stage 2 are very similar and minimal up to the 125GHz search limit.

138 4.3 An existing filter

139 A six layer mesh filter design that has been used on a number of millimeter wave astronomy
 140 instruments, including the Atacama Cosmology Telescope,²⁵ was used to demonstrate the effects
 141 of the optimisation suggested. The original filter showed a Chebyshev-like response at the lower
 142 frequencies in its pass band. In accordance with the Chebyshev filter recommendation, the corners
 143 of the square patches used were cut off. The results of HFSS modelling is shown in Fig. 12. The

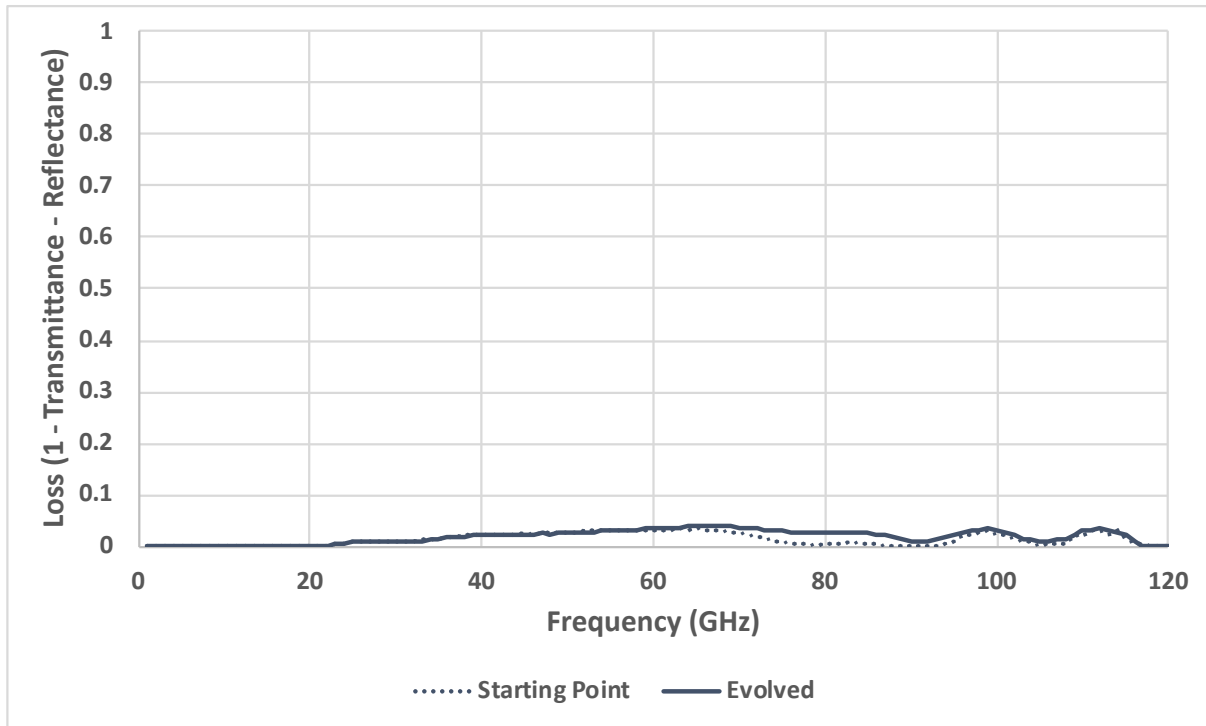


Fig 11 The losses of the starting point (the best solution from stage 1) and the best evolved solution from stage 2 for the Butterworth filter. The losses of the two solutions are minimal and very similar (the evolved pattern having a slightly higher loss around 80GHz) up to 125GHz limit of the search.

144 effect of removing the corners does reduce the pass band ripples slightly. In addition, the transition
 145 from pass band to the cut off is a little sharper. Both effects are quite small but may be useful.

146 5 Conclusions

147 The search for alternative shapes to conventional capacitive squares for the implementation of
 148 millimeter wave filters returned mixed results. The search concentrated on designs that conformed
 149 to a set of constraints to make their modelling possible. The aim, though, was to find shape changes
 150 that might be more generally applicable.

151 For Chebyshev-like filters, where ripple in the pass band is traded for a steeper cut-off, remov-
 152 ing the corners from the conventional squares, can lead to a 35% improvement in the deviation
 153 from the theoretical transmittance. Most of this improvement is achieved in the pass band ripple

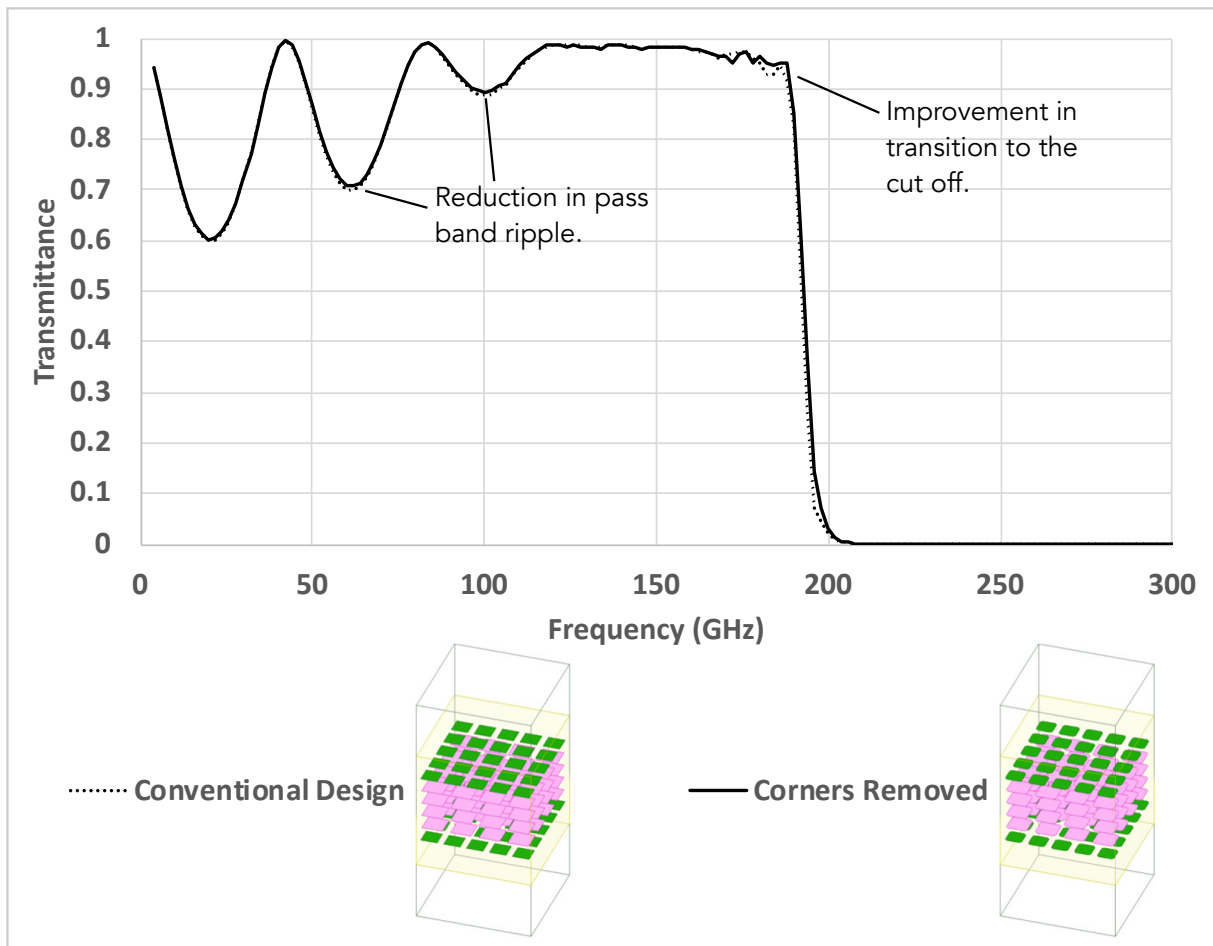


Fig 12 The response of a common low pass filter design used by the Cardiff group in various millimeter wave astronomy instruments showing the effect of removing the corners of the square patches.

154 which is significant. This can be achieved with minimal effect on losses. When applied to an
155 existing filter, there was also a small sharpening of the pass band to cut off corner.

156 The Butterworth style filter tells a different story though. Although the evolved pattern in this
157 case did show an 8% improvement in deviation from the theoretical transmittance curve, most of
158 this improvement came about through better conformance of the cut-off near the stop band. A part
159 of the curve where it can be argued small changes like this are not significant.

160 The experimental testing of the results reported was, unfortunately, halted due to the Covid-19
161 pandemic and remains outstanding for the foreseeable future. The modelling, manufacturing and
162 testing of the mesh technology used in this study is well understood by the group at Cardiff and
163 confidence is high that the performance of manufactured devices would be close to the modelling
164 reported here. This has been demonstrated through a variety of conceptually different devices, half
165 wave plates,^{26,27} Toraldo pupils,²⁸ Magnetic mirrors,²⁹ Mesh lenses.^{30,31}

166 *Disclosures*

167 The authors declare no conflicts of interest.

168 *Acknowledgments*

169 We acknowledge the support of the Supercomputing Wales project, which is part-funded by the
170 European Regional Development Fund (ERDF) via Welsh Government.

171 *References*

172 1 R. Ulrich, “Far-infrared properties of metallic mesh and its complementary structure,” *In-*
173 *frared Physics* **7**, 37–55 (1967).

- 174 2 G. Pisano, C. Tucker, P. A. R. Ade, *et al.*, “Metal mesh based metamaterials for millimetre
175 wave and THz astronomy applications,” in *2015 8th UK, Europe, China Millimeter Waves
176 and THz Technology Workshop (UCMMT)*, 1–4 (2015).
- 177 3 P. A. R. Ade, G. Pisano, C. Tucker, *et al.*, “A review of metal mesh filters,” in *Millimeter
178 and Submillimeter Detectors and Instrumentation for Astronomy III*, J. Zmuidzinas, W. S.
179 Holland, S. Withington, *et al.*, Eds., SPIE, (Bellingham, WA) (2006).
- 180 4 M. Moallem and K. Sarabandi, “A single-layer metamaterial-based polarizer and bandpass
181 frequency selective surface with an adjacent transmission zero,” in *2011 IEEE International
182 Symposium on Antennas and Propagation (APSURSI)*, 2649–2652 (2011).
- 183 5 Y. Wang, B. Yang, Y. Tian, *et al.*, “Micromachined Thick Mesh Filters for Millimeter-Wave
184 and Terahertz Applications,” *IEEE Transactions on Terahertz Science and Technology* **4**,
185 247–253 (2014).
- 186 6 A. M. Melo, A. L. Gobbi, M. H. O. Piazzetta, *et al.*, “Cross-Shaped Terahertz Metal Mesh
187 Filters: Historical Review and Results,” *Advances in Optical Technologies* **2012**, 530512
188 (2012).
- 189 7 M. Navarro-Cía, S. A. Kuznetsov, M. Aznabet, *et al.*, “Route for Bulk Millimeter Wave and
190 Terahertz Metamaterial Design,” *IEEE Journal of Quantum Electronics* **47**, 375–385 (2011).
- 191 8 A. Kundu, S. Das, S. Maity, *et al.*, “A tunable band-stop filter using a metamaterial structure
192 and MEMS bridges on a silicon substrate,” *Journal of Micromechanics and Microengineering*
193 **22**, 045004 (2012).
- 194 9 T. Wu, “Novel metamaterial absorber with fractal elements,” in *2015 IEEE International*

- 195 *Symposium on Antennas and Propagation USNC/URSI National Radio Science Meeting,*
196 244–245 (2015).
- 197 10 J. Holland, *Adaptation in Natural and Artificial Systems: An Introductory Analysis with Ap-*
198 *plications to Biology, Control, and Artificial Intelligence*, MIT Press (1992).
- 199 11 David E Goldberg, *Genetic algorithms in search, optimization and machine learning*,
200 Addison-Wesley, Reading, Mass. (1988).
- 201 12 K. Krishnakumar, “Micro-Genetic Algorithms For Stationary And Non-Stationary Function
202 Optimization,” in *Intelligent Control and Adaptive Systems*, **1196**, 289–297, International
203 Society for Optics and Photonics (1990).
- 204 13 Y. Ge and K. P. Esselle, “GA/FDTD technique for the design and optimisation of periodic
205 metamaterials,” *Antennas Propagation IET Microwaves* **1**, 158–164 (2007).
- 206 14 S. Sui, H. Ma, J. Wang, *et al.*, “Topology optimization design of a lightweight ultra-broadband
207 wide-angle resistance frequency selective surface absorber,” *Journal of Physics D: Applied*
208 *Physics* **48**, 215101 (2015).
- 209 15 J. A. Thompson and G. Pisano, “Use of evolutionary computing algorithms in the design
210 of millimetre-wave metamaterial devices,” in *Millimeter, Submillimeter, and Far-Infrared*
211 *Detectors and Instrumentation for Astronomy IX*, **10708**, 107083H, International Society for
212 Optics and Photonics (2018).
- 213 16 P. Ranjan, A. Choubey, S. K. Mahto, *et al.*, “A six-band ultra-thin polarization-insensitive
214 pixelated metamaterial absorber using a novel binary wind driven optimization algorithm,”
215 *Journal of Electromagnetic Waves and Applications* **32**, 2367–2385 (2018).
- 216 17 P. Ranjan, S. K. Mahto, and A. Choubey, “BWDO algorithm and its application in antenna

- 217 array and pixelated metasurface synthesis,” *Antennas Propagation IET Microwaves* **13**(9),
218 1263–1270 (2019).
- 219 18 H. J. Mohammed, F. Abdulsalam, A. S. Abdulla, *et al.*, “Evaluation of genetic algorithms,
220 particle swarm optimisation, and firefly algorithms in antenna design,” in *2016 13th Inter-*
221 *national Conference on Synthesis, Modeling, Analysis and Simulation Methods and Applica-*
222 *tions to Circuit Design (SMACD)*, 1–4 (2016).
- 223 19 S. D. Campbell, D. Sell, R. P. Jenkins, *et al.*, “Review of numerical optimization techniques
224 for meta-device design [Invited],” *Optical Materials Express* **9**, 1842–1863 (2019).
- 225 20 S. Orfanidis, “Electromagnetic Waves and Antennas.” 2016 (accessed 7 November 2017)
226 [<http://www.ece.rutgers.edu/~orfanidi/ewa/>].
- 227 21 A. Taflove and S. C. Hagness, *Computational electrodynamics : the finite-difference time-*
228 *domain method*, Artech House (2005).
- 229 22 J. B. Schneider, “Understanding the Finite-Difference Time-Domain Method.” 2010 (ac-
230 cessed 23 March 2020) [www.eecs.wsu.edu/~schneidj/ufdtd].
- 231 23 S.-W. Lee, G. Zarrillo, and C.-L. Law, “Simple formulas for transmission through periodic
232 metal grids or plates,” *IEEE Transactions on Antennas and Propagation* **30**, 904–909 (1982).
- 233 24 “ANSYS HFSS: High Frequency Electromagnetic Field Simulation Software.” (accessed 18
234 December 2019) [<https://www.ansys.com/en-gb/products/electronics/ansys-hfss>].
- 235 25 J. W. Fowler, M. D. Niemack, S. R. Dicker, *et al.*, “Optical design of the Atacama Cosmology
236 Telescope and the Millimeter Bolometric Array Camera,” *Applied Optics* **46**, 3444–3454
237 (2007).

- 238 26 G. Pisano, B. Maffei, P. A. R. Ade, *et al.*, “Multi-octave metamaterial reflective half-wave
239 plate for millimeter and sub-millimeter wave applications,” *Applied Optics* **55**, 10255–10262
240 (2016).
- 241 27 G. Pisano, B. Maffei, M. W. Ng, *et al.*, “Development of large radii half-wave plates for CMB
242 satellite missions,” in *Millimeter, Submillimeter, and Far-Infrared Detectors and Instrumenta-*
243 *tion for Astronomy VII*, **9153**, 915317, International Society for Optics and Photonics (2014).
- 244 28 G. Pisano, A. Shitvov, C. Tucker, *et al.*, “Metamaterial-based Toraldo pupils for super-
245 resolution at millimetre wavelengths,” in *Millimeter, Submillimeter, and Far-Infrared De-*
246 *tectors and Instrumentation for Astronomy IX*, **10708**, 107080G, International Society for
247 Optics and Photonics (2018).
- 248 29 G. Pisano, P. A. R. Ade, and C. Tucker, “Experimental realization of an achromatic magnetic
249 mirror based on metamaterials,” *Applied Optics* **55**, 4814–4819 (2016).
- 250 30 G. Pisano, A. Shitvov, P. Moseley, *et al.*, “Development of large-diameter flat mesh-lenses for
251 millimetre wave instrumentation,” in *Millimeter, Submillimeter, and Far-Infrared Detectors*
252 *and Instrumentation for Astronomy IX*, **10708**, 107080D, International Society for Optics and
253 Photonics (2018).
- 254 31 G. Pisano, J. Austermann, J. Beall, *et al.*, “Development of Flat Silicon-Based Mesh Lens
255 Arrays for Millimeter and Sub-millimeter Wave Astronomy,” *Journal of Low Temperature*
256 *Physics* **199**, 923–934 (2020).

257 **Jonathan A. Thompson** is a postgraduate student working towards a PhD in the School of Physics
258 and Astronomy at Cardiff University, UK. His interests are in the use of meta-materials for the
259 implementation of filters and lenses in the field of millimeter wave astronomy.

260 **Giampaolo Pisano** is a Reader in the School of Physics and Astronomy at Cardiff University, UK.

261 His primary interest is new instrumentation focused on the detection of B-modes in the Cosmic

262 Microwave Background. He has developed high performance RF devices such as phase shifters,

263 orthomode transducers, polarizers and polarisation modulators along with novel quasi-optical com-

264 ponents based on metamaterials such as mesh half-wave plates, mesh lenses and magnetic mirrors.

265 **Carole Tucker** is a Professor in the School of Physics and Astronomy at Cardiff University, UK.

266 Her particular expertise lies with the quasi-optical filter technology for which Cardiff is the sole-

267 provider worldwide. She has interests in the fields of FIR astronomical instrumentation for pho-

268 tometric and spectroscopic studies, Quasi Optics and metamaterials for FIR Astronomy and THz

269 applications, IR to THz spectroscopy of optical components and materials and Cryogenic instru-

270 mentation.

271 **List of Figures**

- 272 1 The micro-genetic algorithm. The starting point is an initial population of 5 in-
273 dividuals. Their chromosome bit strings are normally generated randomly, but in
274 stage 2 of the methodology one of the individuals is initialized from the best so-
275 lution of stage 1. The fitness of each individual is then assessed by running the
276 electromagnetic model on the structure represented by its chromosome and com-
277 paring the results with the fitness function. The best individual is marked as the
278 elite and is automatically included in the next generation. Four new members are
279 then bred from the previous generation to bring the numbers back to 5. A check is
280 performed to maintain the genetic diversity of the generation; if all the individuals
281 have more than 95% of their chromosomes identical, the four bred individuals are
282 replaced with completely new random individuals. The loop then continues with
283 fitness evaluation.
- 284 2 A pixelated plate and its binary encoding. As described by Ge and Esselle¹³ and
285 reiterated in Thompson and Pisano,¹⁵ a four-fold pixelated NxN plate can be rep-
286 resented by the binary code covering one triangle of the plate as shown in this
287 example of an 18x18 plate. The numbers in the pixels indicate the bit number in
288 the gene encoding.

- 289 3 The best solution from stage 1 for the seventh order Chebyshev low pass filter
290 example. It consists of two layers repeated in reverse order with a repeat cell size
291 of 1.716mm and a layer spacing of 0.875mm. Layers 1 and 4 consist of a 2x2
292 layout of 80% squares each in a unit cell of 0.858mm. Layers 2 and 3 are single
293 70% squares in a unit cell of 1.716mm covering the entire repeat cell.
- 294 4 The best solution from stage 2 for the seventh order Chebyshev low pass filter
295 example. These are the 4-fold 20x20 binary patterns settled on for the four layers.
- 296 5 The transmittance curves of the starting point (the best solution from stage 1), the
297 best evolved solution from stage 2 and the ideal filter response. It can be seen that
298 the plate pattern changes from the stage 2 evolution have caused the transmittance
299 characteristic to move closer to the target; most significantly, the pass-band ripple
300 has been reduced.
- 301 6 The losses of the starting point (the best solution from stage 1) and the best evolved
302 solution from stage 2 for the Chebyshev filter. The losses of the two solutions
303 are minimal and similar within the passband. In the stop band, however, the
304 evolved solution showed significant loss around 125GHz. The diffraction limit
305 is at 160GHz.
- 306 7 Fitness factors were calculated for the result of stage 2 and for patterns with various
307 stage 2 changes removed. The conclusion is that the corners of layers 2&3 and the
308 outer corners of layers 1&4 make the most difference. The other changes appear
309 to make only a marginal difference.

- 310 8 The best solution from stage 1 for the fifth order Butterworth low pass filter ex-
311 ample. It consists of two layers repeated in reverse order with a repeat cell size of
312 1.789mm and a layer spacing of 1.167mm. Layers 1 and 4 consist of a single 30%
313 square. Layers 2 and 3 are single 60%.
- 314 9 The best solution from stage 2 for the fifth order Butterworth low pass filter exam-
315 ple. These are the 4-fold 20x20 binary patterns settled on for the four layers.
- 316 10 The best solution from stage 2 for the fifth order Butterworth low pass filter exam-
317 ple. These are the 4-fold 20x20 binary patterns settled on for the four layers.
- 318 11 The losses of the starting point (the best solution from stage 1) and the best evolved
319 solution from stage 2 for the Butterworth filter. The losses of the two solutions are
320 minimal and very similar (the evolved pattern having a slightly higher loss around
321 80GHz) up to 125GHz limit of the search.
- 322 12 The response of a common low pass filter design used by the Cardiff group in
323 various millimeter wave astronomy instruments showing the effect of removing
324 the corners of the square patches.