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SMS2D designs as starting points for lens optimization

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Abstract: Traditional imaging design methods can often be ineffective when designing aspheric systems because of the large number of optimization parameters and lack of a good starting point. They are often trapped in a poor local minimum and it can be highly time-consuming to find a good solution in a bumpy design landscape. The simultaneous multiple surface (SMS) method can significantly shorten the time and effort needed to find a desired solution by providing a starting point to optimize close to a good local minimum. We investigate here two design examples and compare them with similar designs obtained via traditional design approaches, as well as global optimization. In the examples considered here, the SMS method combined with a shorter optimization leads to an optimal design.

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1. Introduction

Classic imaging design methods are based on aberration theory and multi-parametric optimization techniques [1–3]. With the development of precise manufacturing techniques, aspheric surfaces are no longer a luxury and therefore can be encountered quite often in modern designs. For optical systems with few mild aspheric surfaces described by a small polynomial order, the traditional techniques may provide many useful solutions. However, when more complex systems with many aspheric surfaces with a large number of free parameters are considered, the optimization approach can easily get trapped in a poor local minimum (e.g., M_2 in Fig. 1(a)) when the starting point (S_2 in Fig. 1(a)) is far from the good local minimum of interest (M_1 in Fig. 1(a)). In such cases, the designer's experience is often a crucial factor when trying to get out of a poor minimum in search for a good solution. For instance, optimization is initially constrained to a lower dimension subspace, by optimizing only the low order coefficients. When a local minimum in the constrained subspace is reached (e.g., M in Fig. 1(b)), the subspace dimension is increased by unfreezing some higher-order coefficients. In the larger dimension subspace, that design is no longer a minimum and optimization can progress again, as illustrated in Fig. 1(b). When the number of variables is large, this change of the parameter subspace dimension can be done in various ways. Depending on the experience of the designer, which and how many variables are chosen in each step to optimize the system can be very different. Different design strategies usually end up in different local minima. Several global optimization methods such as genetic algorithms [4], evolution strategies [5], particle swarm [6] and simulated annealing [7] have been adapted to different optical design problems to avoid getting trapped in a local minimum. However, these methods are based on generally applicable mathematical models, and designers understand little about the optical systems behind them. Another method known as saddle point construction (SPC) method uses a systematical approach to find solutions in the design landscape and helps to get out of bad local minima [8,9].

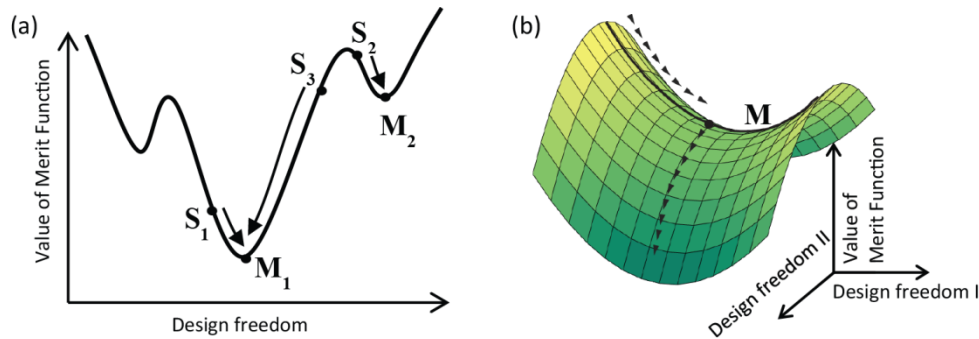


Fig. 1. An illustration of an optical design landscape in 2D (a) and 3D (b).

Alternatively, the Simultaneous Multiple Surfaces (SMS) method is a direct method that allows to find a better starting point (like S_1 in Fig. 1(a)) [10,11] and progress with it with a single step optimization (varying all parameters). Even though SMS is capable of designing good aspheric imaging systems, there is no study comparing it directly with other design strategies and describing the characteristics of different methods.

In this paper, we consider rotationally symmetric designs, for which SMS 2D is used [12]. SMS 2D may utilize rays in the meridian plane or skew rays [13], but in any case, obtaining a 2D profile of the surfaces. The method involves simultaneous calculation of N optical surfaces using N one-parameter bundles of rays for which specific conditions are imposed. The number of ray bundles can be greater than the number of surfaces to design when the footprints of the design bundles do not occupy the full SMS surfaces, as demonstrated in [14–16]. SMS also has a freeform version called SMS 3D. However, these cases are not discussed in this paper.

We start by briefly describing the SMS method applied in this paper. Then the design obtained with SMS is used as a starting point, followed by a single-step optimization of all parameters. The final design is compared with other designs obtained using spherical starting point and single-step, two-steps, stepwise and global optimizations. Two designs are considered, a simple one and a slightly more difficult one, with a larger aperture and wider field.

2. Starting point with the SMS method

The standard SMS procedure for designing aspheric surfaces involves only meridian rays. It consists of two steps: selection of central segments of the surfaces and recursive generalized Cartesian oval calculation [13,17]. We will focus on SMS designs using only meridian rays. Skew rays are then controlled in a subsequent optimization.

Here we study two designs that consist of four aspheric surfaces and use the SMS 2D using four meridian bundles. In this case, the four meridian ray bundles selected for the SMS design correspond to the rays emitted from four object points placed symmetrically about the optical axis at infinity. The image points are located in the same way at a finite distance to match the specified focal length. The field angles associated with the object points are selected so that their root mean square 2D (RMS2D) distribution curves (defined as the RMS spot diameter calculated using only meridian rays in the field) present a constant ripple over the field of view as shown in the right plot in Fig. 2 and explained in detail in [11].

Applying a standard SMS 2D procedure, for the design directions we simultaneously calculate the set of points and normals. SMS profiles are then fitted into a Forbes Qcon polynomial [18], which will be introduced into the CODE V software. This system is now ready to be optimized for the whole field of view, as in the example shown in Fig. 2.

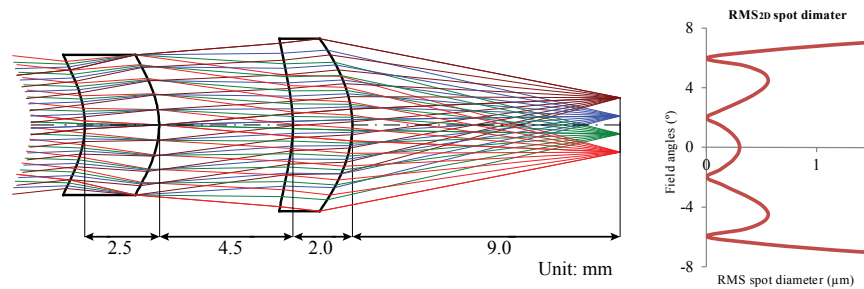


Fig. 2. SMS 2D design introduced to CODE V after fitting the surfaces to Forbes Qcon polynomials. The change of the spot diameter through the fields is shown on the right side.

Aperture stop in system 1 is positioned at the surface 2 of the first lens element (when counting from left to right, in Fig. 2). Since this study focuses on designing four surfaces using four ray bundles, in system 2 the aperture stop is placed in between the two lens elements (because other aperture stop positions lead to the use of more than four ray bundles, as mentioned in the introduction and given in [14–16]).

3. Comparing the results of different design approaches

Two lens systems were designed and optimized for monochromatic light ($\lambda = 587.56$ nm) using SMS method and other approaches mentioned in the introduction. Both lenses are made of PMMA ($n = 1.4918$ at the used wavelength) and have four aspheric surfaces in total. Their specifications are given in Table 1. The object is located at infinity. System 1 has both a smaller field of view (FOV) and aperture, and system 2 has both a larger FOV and aperture.

Table 1. System specifications

	Image F/Number	Effective Focal Length (EFL)	Maximum Half Field of View (MHFOV)	Position of the Aperture Stop
System 1	2.24	8.60 mm	7.50°	At surface 2
System 2	1.77	10.60 mm	11.25°	Air-space between the two lenses

Three design approaches are used for the comparison: 1) SMS design with a subsequent one-step optimization; 2) optimization from a spherical system and adding aspheric coefficients, either in one-step, two-steps or one-by-one (stepwise); 3) global optimization with Global Synthesis (GS) in CODE V. Since we do not intend to compare different global optimization algorithms in this paper, we choose for comparison with SMS only GS that is known to perform very well in comparison with other global optimization and search methods in lens design [19,20]. For comparison, the systems designed all use Qcon polynomials as the representation for the aspheric surface. The variables are the curvatures, conic constants and higher-order coefficients of the Qcon polynomials (up to the 12th order for system 1, and 16th order for system 2). The vertices are fixed in all cases. For non-GS optimization, we use the default CODE V transverse ray aberration merit function (MF). Two types of constraints are used. One is to constrain the effective focal length (EFL), and the other is to limit the normalization radii of the surfaces to a value exceeding the size of the surface aperture by 5%. The local optimization is done within CODE V based on a damped least square method. Vignetting was readjusted every 20 cycles, since the surface shape may vary significantly during the optimization process. Iterations are stopped when the MF value drops to less than 10^{-6} (convergence is then assumed). For GS optimization, the same MF and constraints are applied. However, vignetting can only be adjusted after the GS is finished.

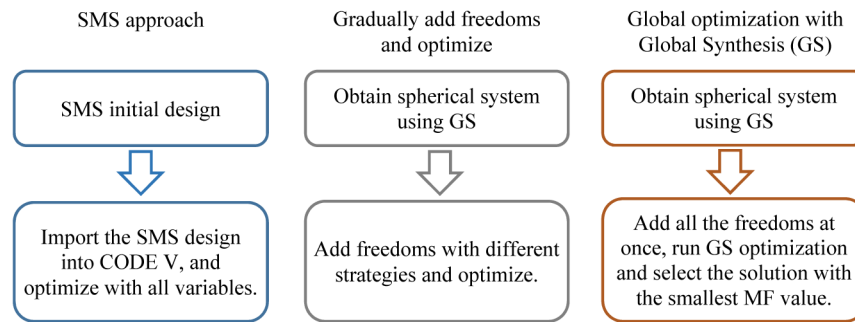


Fig. 3. Flowchart for our different design approaches.

The SMS 2D procedure enforces stigmatic imaging for meridian rays at a few discrete design fields. After the system is designed with SMS, it is imported into CODE V to be optimized for better performance throughout the whole field of view and the full pupil. The surface parameter of the two SMS constructed systems are given in the Appendix. The left flowchart in Fig. 3 shows the design flow with SMS.

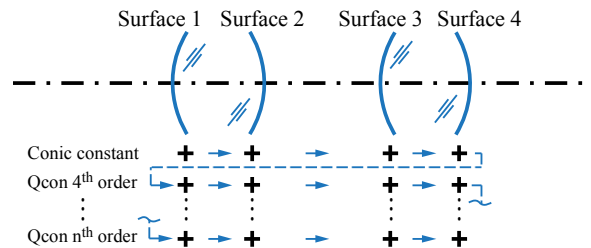


Fig. 4. The strategy for stepwise optimization. The blue shapes represent the optical surfaces. Each cross represents a freedom on the surface described by the Q_{con} polynomial. The blue arrows indicate how the freedoms are added step by step up to the highest order coefficients used for Q_{con} polynomial.

The second approach is conventional and intuitive. We start with a good spherical system. This system is the only solution when spherical surfaces are used, and it is confirmed by GS of CODE V. Next, the aspheric coefficients are added as new variables. There are different strategies for adding variables to the system. The direct way is to free all the variables at once and optimize. For an optimization landscape with multiple minima, the optimized solutions will depend on the chosen starting point. Designers will not have control of the optimization route, and it is easy to get trapped in a poor local minimum. A common practice of experienced designers is adding variables step by step. In this paper, we apply three different strategies for adding new variables: a single-step optimization, a two-steps optimization, and an extensive stepwise approach. A two-steps approach is initiated by adding all the conic coefficients at once and then optimizing to a local minimum with conic surfaces. Subsequently, the higher-order coefficients of Q_{con} polynomial are added all together and optimized. The stepwise approach requires more steps by adding one freedom and then optimize each time: We start by freeing the conic constant on surface 1. After optimization, new freedom is added on the surface following the previous one, as shown in Fig. 4. Surface 1 with higher-order coefficient follows after surface 4 with lower-order coefficient. In practice, one can add the freedom to the system in several ways, keeping in mind that on each surface, the coefficients should be added from low order to high order. Since it is not known which stepwise strategy will lead to a better solution, in this paper, we choose an easy stepwise strategy for demonstration.

In addition to the two approaches mentioned above, a global optimization method based on CODE V's GS is used for comparison as well. The proprietary algorithm of GS can help

the designer to automatically explore the solution space and synthesize new configuration from (nearly) arbitrary starting point [21]. To apply GS, we use the spherical system mentioned above as a starting point. All the freedoms are added at once to the system. GS is executed with constraints on EFL and normalization radii. A time limit of two hours is set for the GS (i. e. GS stops after two hours if it is not finished). The resulting system with a reasonable system shape and the smallest MF value is chosen for the comparison.

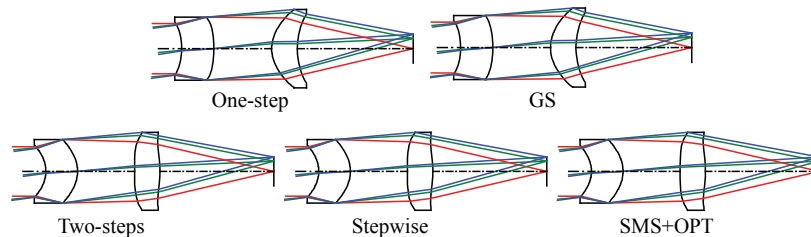


Fig. 5. System 1: the system shapes obtained using different design approaches.

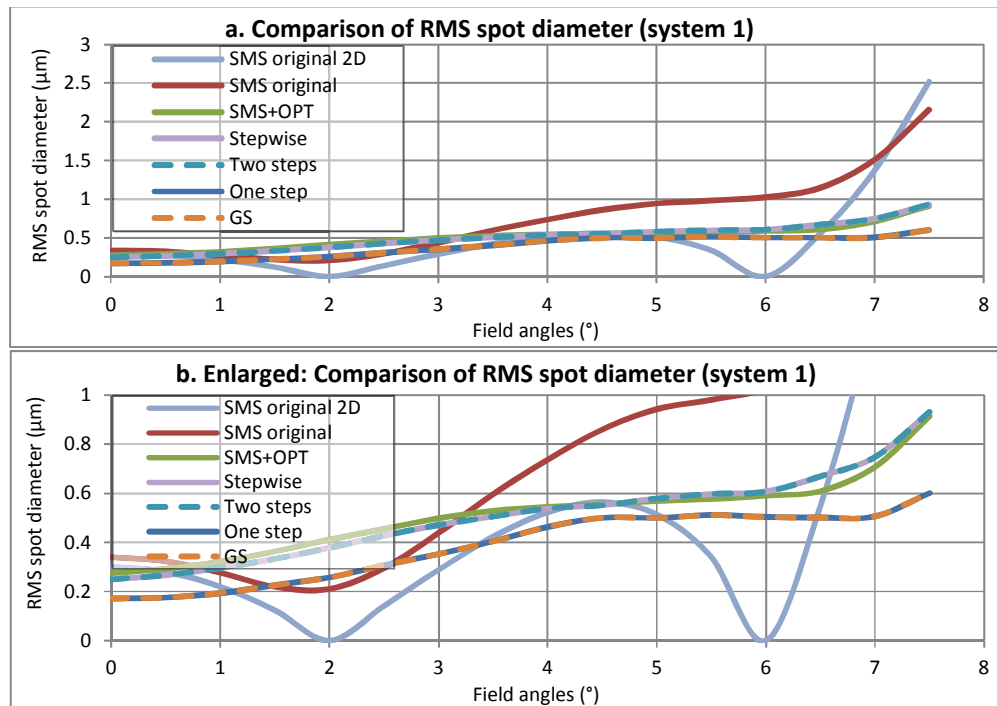


Fig. 6. RMS spot diameter distribution curves for different design approaches considered: complete curves (a); enlarged section (b). The RMS spot diameter values of the starting spherical system vary from 60 to 96 μm and are not presented in the graph.

The designed results for system 1 are presented in Fig. 5. By examining the shapes of the surfaces, we observe two groups of systems: GS and one-step form one group; two-steps, stepwise and SMS + OPT form the other group. This is consistent with the two groups of RMS spot curves in Fig. 6. It indicates that solutions from five different strategies end in two different minima in the design space.

In Fig. 6, we see that all five systems generated by different approaches show good performance with the maximum RMS spot diameter smaller than 1 μm at the field of 7.5°.

The starting SMS system has relatively large RMS spot diameter when the field is increased. Since SMS 2D construction involves only meridional rays, RMS spot diameter is

zero for the design fields (2° and 6°) only for those meridional rays (SMS original 2D curve in Fig. 6). In Fig. 6, a minimum is observed at 2° for the SMS original curve, however, at 6° there is no obvious minimum. Since the RMS spot diameter of SMS original in Fig. 6 results both from meridional and skew rays, it is reasonable to expect that when the whole pupil is considered, skew rays start to have a larger influence on the spot diameter. The optimization of the SMS constructed system balances the spot diameter over the field with the consideration of the rays from the whole pupil. Except the SMS original and SMS original 2D curve, the curves in Fig. 6 form two groups, with GS and one-step forming one group, and SMS + OPT, two-steps and stepwise forming the other group, as expected from the system drawings shown in Fig. 5. The results from GS and one-step are slightly better than the other group. It can be observed that the curves of GS and one-step almost overlap. The same happens for two-steps and stepwise curves. The overlap of the curves indicates that the optimization strategies lead to the same local minimum in the optimization landscape. The curve of SMS + OPT only has a very small difference shown in field dependence of the RMS compared to the stepwise and two steps curves, which is practically considered as the same local minimum. Because of the smaller values of aperture and field considered here, we have many degrees of freedom, and it is easy to achieve diffraction limited performance. In this situation, several good minima exist in the solution space as illustrated in Fig. 7. Even though the designer still has no control during a one-step optimization, the chance for the optimization reaching a good minimum is higher than in a more difficult design problem.

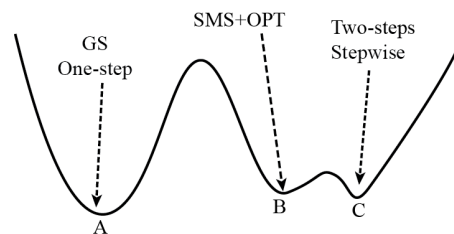


Fig. 7. An illustration of the design landscape for system 1. We have found three equally good local minima which form two groups. Solution A (MF 0.0301) forms the first group. Solution B (MF 0.0461) and Solution C (MF 0.0450) form the second group.

Figure 8 shows the comparison of the efficiency of the strategies considered in this paper. The GS result is not included because of its very different approach compared to the others: instead of optimizing from one starting point, it searches for different starting points, applies local optimizations to them, and produces multiple solutions [21]. We usually choose the system with the smallest merit function value and perform an extra local optimization to make sure it converges. In this case, we cannot easily use the number of optimization cycles to represent the efficiency of GS. Three solutions were generated with GS after two hours (Intel i5-3470 dual-core @3.20 GHz system). In comparison, it took around 30 seconds for a one-step optimization to run 1000 cycles on the same computer. In Fig. 8, the starting points are shown at cycle 0. The two groups have their average MF values as 0.0454 and 0.0301 (the MF value of the GS result is 0.0301). An MTF analysis shows that all five systems are diffraction limited at the wavelength used here. This is consistent with our statement that the design problem is easy and multiple equally good minima are present. Therefore, this example shows that in simple design problems like system 1 simple methods can easily lead to a good solution.

System 1 has a relatively small pupil and field, which reduces the difficulty of the design. With the same approach, we used SMS to construct system 2, which has both a larger pupil and field of view. Figure 9 shows the five plots of the systems produced by five different optimization approaches. In contrast with the solutions of system 1 in Fig. 5, four different system shapes are obtained, where stepwise and SMS + OPT resulted in the same system shape. The operation with GS for system 2 was not as straightforward as that for system 1.

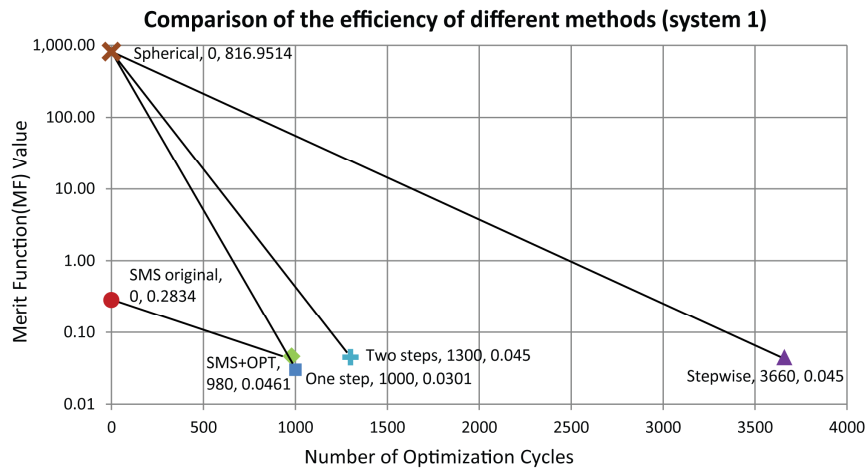


Fig. 8. Comparison of the efficiency of different design methods analyzed, evaluated as the merit function value versus the number of cycles performed. The merit function value of the resulting system from GS is 0.0301.

Optimization with conic constants as variables ended with systems having an extremely large conic constant. Therefore, we first fixed the conic constant on all the surfaces to -1 , and ran the GS which resulted in twelve solutions. After that, the conic constants on all the surfaces of the twelve systems were freed at once as variables. Subsequent local optimization, which is the same as the one used in the other four strategies, was applied to each of the twelve systems. The system with the smallest MF value was chosen as the final result. In Fig. 10, the curves of RMS spot diameter can be again divided into two groups. One-step and two-steps optimization generate poorer solutions than the rest of the approaches. SMS with optimization, stepwise optimization and GS obtain good solutions with close RMS dependence along the field (Fig. 10(b)). Consistent with the same system shape in Fig. 9, the RMS curves of SMS + OPT and stepwise in Fig. 10 almost overlap each other. On the other hand, the result of GS with a different system shape produces a different RMS curve, where the RMS spot diameter is bigger under 7.8° and smaller above 7.8° .

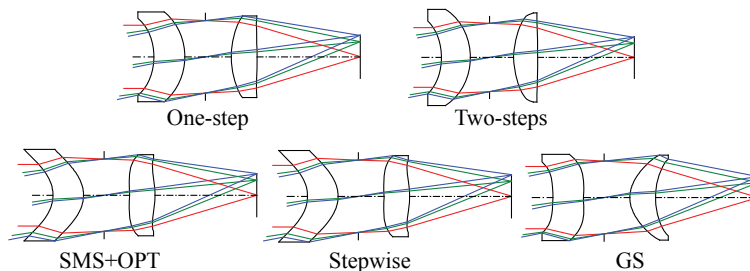


Fig. 9. System 2: the system shapes obtained using different design approaches.

Figure 11 shows the comparison of the efficiency of different methods. Stepwise needs a large number of optimization cycles (8100 cycles) to converge. With a similar merit function value, fewer optimization cycles are needed for SMS optimization. This shows the SMS constructed starting point is closer to the minimum in the optimization landscape. Both one-step and two-steps optimization are trapped in poor solutions with MF values at least nine times of the best solution.

From the two examples we studied, to design a simple system with small field and aperture, all approaches including a one-step optimization were able to obtain good solutions. However, when the field and aperture increased, simple strategies like one-step and two-step

started to fail by getting trapped in poor local minima as shown in the example of system 2. A stepwise approach leads to a good local minimum with a large number of optimization cycles. An SMS constructed starting point located closer to a good solution shows its advantage in both cases where fewer optimization cycles are needed to reach a good local minimum.

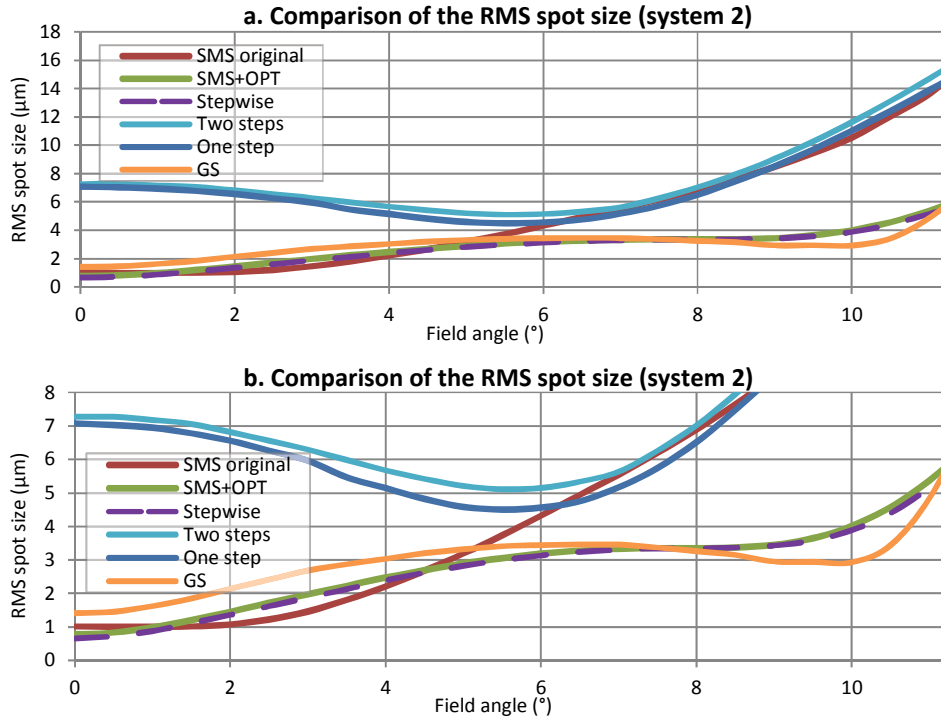


Fig. 10. RMS spot diameter curves for second system using different design approaches: complete curves (a); enlarged section (b). The RMS spot diameter values of the starting spherical system vary from 124 to 142 μm from the center to the full field and are not shown in the graph.

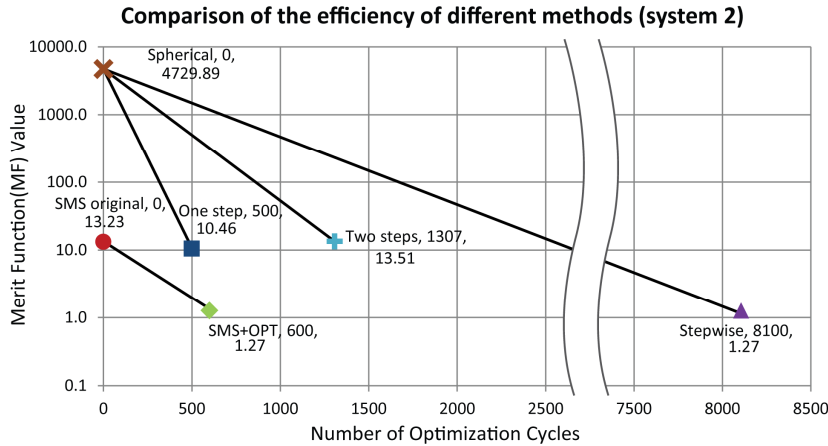


Fig. 11. Comparison of the efficiency of different design methods. The starting points are shown at zero cycles and are connected with the straight lines with the results obtained using different design methods. The merit function value of the GS result is 1.03.

4. Design landscape for aspheric systems

In our examples, we have noticed that when the aspheric variables are not introduced, there is only one solution in the optimization landscape given the boundary condition, where rays can be traced. However, when aspheric coefficients are added, new minima start to appear.

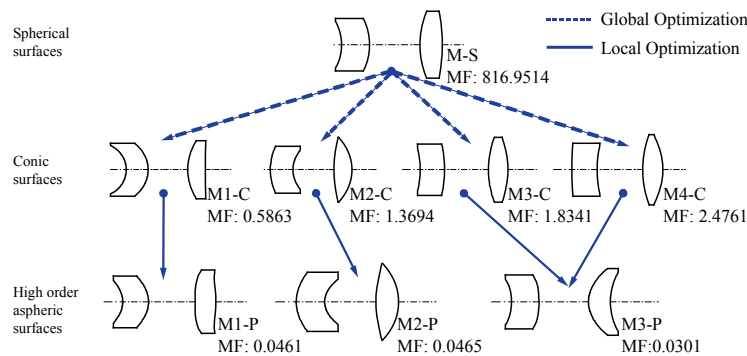


Fig. 12. Evolution of the minima with the change of aspheric coefficients (system 1). Four minima are found with conic surfaces. They become three minima when optimized with aspheric coefficient up to 12th order. All three are found with the different approaches we used. M1-P and M3-P are the two solutions previously found with the different approaches (Fig. 5).

After adding the conic constant as a variable, multiple minima appear in the design landscape. We used GS of CODE V, in this case, to search for different solutions. For system 1, we have found four minima (Fig. 12), and for system 2 three (Fig. 13).

The following step included adding all other higher-order aspheric coefficients at once to locally optimize the systems from the conic minima. For system 1, four different minima in conic space converge to three minima. In Fig. 12, two of the three minima (M1-P, M3-P) are identical to the ones shown in Fig. 5, and M2-P was also found by the GS, which is not shown in Fig. 5. For system 2, three solutions result from the three conic minima. In Fig. 13, M1-P and M2-P are same as the results of the two-steps approach and the SMS with optimization respectively. M3-P is a poor local minimum which is not found by the approaches we used.

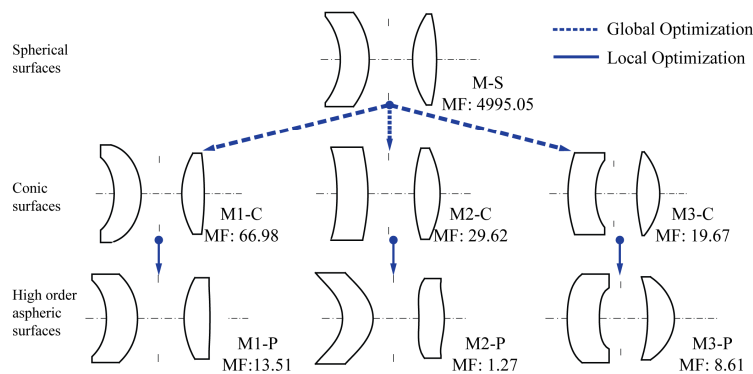


Fig. 13. Evolution of the minima with the change of aspheric coefficients (system 2). Three minima are found with conic surfaces. Adding higher-order aspheric coefficients (up to 16th order) to these minima results in three different solutions. M1-P is the solutions found by the two-steps approach, and M2-P is the solution found by the SMS with optimization in Fig. 9.

The results show the dynamic behavior of the optimization landscape. Namely, the number of minima may change depending on the number and type of variables used. In the two systems we have studied in this paper, the introduction of the conic surface creates new minima in the design landscape, and adding higher-order aspheric coefficient deepens the

original minima. For a simple system, adding higher-order aspheric coefficients does not complicate the landscape. With local optimization, the conic solutions lead to the good solutions with higher-order aspheric coefficients. In the case of system 2, where aperture and field are large, adding higher-order aspheric coefficients complicates the landscape by creating new local minima. That is the reason why simple strategies are trapped in poor local minima. In both cases, the SMS constructed starting points are already close to one of the good local minima in the landscape. Therefore, local optimizations with few optimization cycles lead to good solutions.

5. Conclusions

Two different optical systems comprising of two lenses were analyzed. In the analysis of the design landscape of the systems, we have observed that with aspherising the surfaces, new minima emerge into the landscape. In both examples, introducing lower order aspheric coefficients (conic constant) to all the surfaces brings new local minima into the landscape, and adding higher-order aspheric coefficients deepens the minima. In a simple design with both small aperture and field (system 1), adding higher-order coefficients does not complicate the landscape by creating new local minima. In contrast, for a different design with large aperture and field (system 2), higher-order aspheric coefficients introduce new local minima, where a design process can be trapped.

We have compared SMS design approach with other design techniques to design and optimize the two systems. In the case of the simpler system (system 1) considered, all design techniques including a one-step optimization found solutions close to diffraction limit. Nevertheless, surface shapes of the final designs were different which suggested that their optimization paths ended in different basins of attractions.

In the second design example (system 2) analyzed, SMS with posterior optimization, the stepwise approach and GS found optimal solutions. Carefully analyzing the three results obtained, we conjectured that they belong to two different minima. Since designing system 2 is more demanding than system 1, less robust methods such as one-step optimization and two-steps optimization did not lead to good minima. Despite that three approaches found good solutions, stepwise optimization usually takes the largest amount of optimization cycles. GS works well with the simple case. However, for system 2, effective constraints (in this case, fixed conic constant) and further optimizations are needed to be implemented into GS to get reasonable solutions. In practice, lens design problems are very different depending on the application and constraints. While we cannot guarantee that SMS will find good solutions in other design problems, the present results are encouraging. In both cases, SMS with posterior optimization can find good solutions, and we can see from Fig. 8 and Fig. 11 that in the two examples considered, SMS constructs good starting points which are close to the good solutions. It is our hope that with such comparative studies, we can provide the designers with insight into the characteristics of different methods.

Appendix: system parameters of the SMS constructed systems of system 1 and system 2

In this appendix, we list in Table 2 and Table 3 the surface parameters in CODE V of the two SMS constructed systems.

Table 2. Surface parameters in CODE V of system 1 constructed with SMS

Surface	1 (Object)	2	3 (Stop)	4	5	6 (Image)
Surface type	Sphere	Qcon Asphere	Qcon Asphere	Qcon Asphere	Qcon Asphere	Sphere
Material		PMMA		PMMA		
Curvature	0.0000	-0.3142	-0.3424	-0.1874	-0.2925	0.0000

Thickness	Infinity	2.50	4.50	2.00	9.00	0.00
Normalized Radius		2.20	2.40	2.75	2.90	
K		-1.2937	-1.3707	-7.0733	-0.9741	
QC4		5.9543E-02	2.7731E-02	1.1610E-01	1.2458E-01	
QC6		9.8831E-03	2.8067E-03	-7.3253E-03	-9.2028E-03	
QC8		-4.3303E-04	2.6278E-04	8.3891E-04	9.1018E-04	
QC10		-6.7898E-05	2.8571E-05	-2.4930E-05	-1.9440E-05	
QC12		5.5475E-06	-1.7237E-05	6.7655E-06	9.0658E-06	

Table 3. Surface parameters in CODE V of system 2 constructed with SMS

Surface	1 (Object)	2	3	4 (Stop)	5	6	7 (Image)
Surface type	Sphere	Qcon Asphere	Qcon Asphere	Sphere	Qcon Asphere	Qcon Asphere	Sphere
Material		PMMA			PMMA		
Curvature	0.0000	-0.0823	-0.1386	0.0000	-0.0876	-0.2183	0.0000
Thickness	Infinity	3.00	2.00	2.50	2.50	10.00	0.00
Normalized Radius		3.70	3.40		3.50	3.70	
K		-9.2639	-5.7686		6.0225	-0.3232	
QC4		1.4285E-01	1.8381E-01		2.6624E-01	2.1069E-01	
QC6		2.8156E-02	4.0717E-02		1.2446E-02	9.0396E-03	
QC8		-1.6114E-03	3.0800E-03		3.4723E-03	2.4930E-03	
QC10		-1.4590E-04	7.4292E-04		5.4780E-04	4.6335E-04	
QC12		2.0759E-05	1.2620E-04		9.8619E-05	9.9287E-05	
QC14		-1.4991E-06	2.9401E-05		1.7711E-05	2.0624E-05	
QC16		-2.7096E-06	3.8057E-06		3.1363E-06	4.9458E-06	

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