

1 **Kaczmarz Holography: Holographic Projection in the** 2 **Fraunhofer Region**

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7 **Abstract.** The Kaczmarz algorithm is an iterative method for solving linear equations in the form of
8 $Ax = b$. It is widely used in computed tomography (CT) and digital signal processing (DSP) but has yet
9 to be adopted in computer-generated holography (CGH). Phase retrieval algorithms such as Gerchberg-
10 Saxton or Fienup are significantly more popular in this field, however, in this paper we propose a unique
11 and alternative approach to projecting a replay field through Discrete Fourier Transform (DFT) matrices and
12 have shown that there are legitimate benefits to implementing this approach. The gradient descent iteration
13 mechanism adopted by Kaczmarz, for instance, provides finer granularity control over the individual pixels
14 in the replay field. We consequently demonstrate the quality of the image is significantly improved when
15 compared to Gerchberg-Saxton.

16 **Keywords:** Computer-Generated Holography, Kaczmarz, Gerchberg-Saxton.

17 **1 Introduction**

18 Photography by wavefront reconstruction, later known as holography has seen significant
19 developments since its first publication in 1948 by Dennis Gabor.¹ It works by record-
20 ing onto a photographic plate, the interference pattern between a coherent reference light
21 source and this same light source scattered off an object. A replay field is then projected
22 by shining a coherent light source through this interference pattern.

23 With the advancement in computer technology, it is now possible to generate these
24 interference patterns computationally without the need for photographic plate or physi-
25 cal object. The rapid availability of computer-controlled spatial light modulators (SLMs)
26 facilitated the uptake of computer-generated holography for a wide range of applications
27 from optical tweezing² to telecommunications.³ SLMs modulate the spatial profile of an
28 incoming coherent beam replacing the use of plates of conventional holography. The con-
29 figuration of a hologram in modern computer-generated holography is dynamic, allowing
30 for the projection of moving images as opposed to still images in conventional holography.

31 The development of these technologies has contributed towards the use of SLMs in
32 augmented and virtual reality (AR and VR) systems.⁴⁻⁷ SLMs are capable of projecting
33 three-dimensional images with appropriate depth cues that do not exhibit what is known
34 as vergence-accommodation conflict (VAC).⁸⁻¹⁰

35 This paper focuses on projecting two-dimensional holographic images in the Fraun-
 36 hofer region, also known as the far-field or replay field. In these systems, an SLM is
 37 configured to display a hologram that will modulate the coherent light beam displaying
 38 the desired image in the replay field.¹¹⁻¹³ Goodman¹⁴ demonstrates that taking the discrete
 39 Fourier transform (DFT) of the aperture function of the SLM will reproduce the pattern of
 40 the image in the far-field. The SLM is a pixelated device in the diffraction field (H) and
 41 as such is represented by a grid of discrete sampling points. The resultant replay field (R)
 42 is therefore also represented by discrete pixels and the relationship between both is shown
 43 in Fig 1. The Fourier transform is represented here as a DFT sum, however, it is usually
 44 represented as an integral.¹⁵

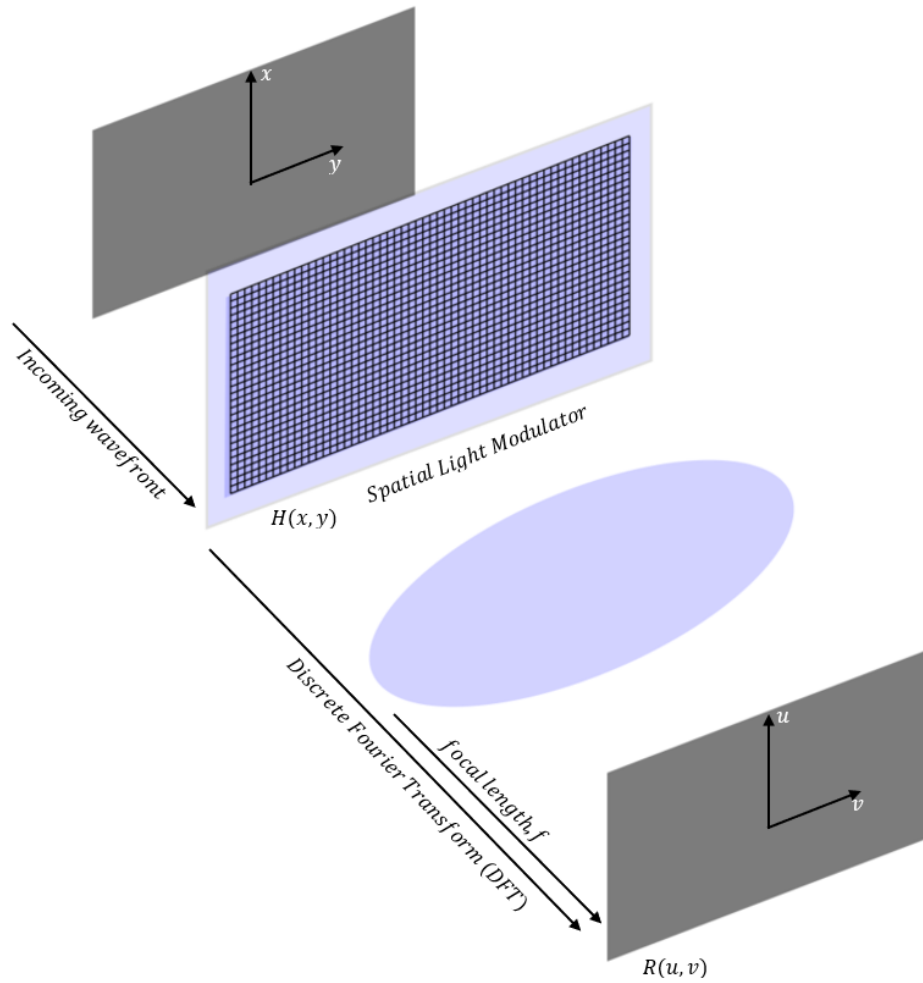


Fig 1: Diffraction/Hologram (H) and replay (R) fields coordinate system

45 The diffraction field coordinates are indicated as x and y while replay field coordinates
 46 are portrayed as u and v . It is worth noting, N_x and N_y represent the sizes of the diffraction
 47 and replay fields and are represented as follows.

$$R_{u,v} = \mathcal{F}\{H_{x,y}\} = \frac{1}{\sqrt{N_x N_y}} \sum_{x=0}^{N_x-1} \sum_{y=0}^{N_y-1} H_{x,y} e^{-2\pi i \left(\frac{ux}{N_x} + \frac{vy}{N_y} \right)} \quad (1)$$

$$H_{x,y} = \mathcal{F}^{-1}\{R_{u,v}\} = \frac{1}{\sqrt{N_x N_y}} \sum_{u=0}^{N_x-1} \sum_{v=0}^{N_y-1} R_{u,v} e^{2\pi i \left(\frac{ux}{N_x} + \frac{vy}{N_y} \right)} \quad (2)$$

48 The SLM is restricted to modulating either the phase or the amplitude of an incident
 49 coherent light beam, not both. As a result, the challenge lies in constructing a hologram
 50 that can be displayed on this SLM under these restrictions and accurately projecting the
 51 target image in the replay field.

52 This paper analyzes the performance of the Kaczmarz algorithm in two-dimensional
 53 holographic image projection. Alternative methods such as Gerchberg-Saxton,¹⁶ Fienup¹⁷
 54 and Wirtinger¹⁸ are very popular in this field and the aim of this paper is to introduce an
 55 alternative method for projecting these images. This method iterates pixel by pixel in the
 56 replay field and therefore offers finer granularity over these individual pixels, a property
 57 that distinguishes it from other algorithms. To facilitate comparison between the different
 58 algorithms, error metrics are applied to quantitatively assess the difference between the re-
 59 play field, R , and original target image, T . Mean squared error (MSE) and peak signal to
 60 noise ratio (PSNR) will be used throughout this investigation.

$$MSE(T, R) = \frac{1}{N_x N_y} \sum_{u=0}^{N_x-1} \sum_{v=0}^{N_y-1} |T_{u,v} - R_{u,v}|^2 \quad (3)$$

$$PSNR(T, R) = 20 \log_{10} \left(\frac{N_x N_y}{\sqrt{MSE}} \right) \quad (4)$$

61 **2 Kaczmarz Algorithm**

62 First proposed by Polish mathematician Stefan Kaczmarz,¹⁹ this iterative algorithm is used
 63 in computational tomography (CT), digital signal processing (DSP), microscopy and meta-
 64 surface antennas²⁰⁻²⁴ to solve linear equation systems in the form of $Ax = b$. Typically,
 65 holography has been proven to work using Gerchberg-Saxton and Fienup algorithms since
 66 these are very effective in retrieving the phase of an image. However, the work carried
 67 out by Wei²⁵ and subsequently Tan²⁶ on the Kaczmarz algorithm suggests it is also very
 68 effective in solving this phase retrieval problem.

69 This paper analyses the effectiveness of this algorithm in holography as well as its
70 necessary adaptation to make it applicable to this use-case.

71 2.1 Simple Kaczmarz Algorithm

72 The Kaczmarz algorithm was originally devised to solve overdetermined linear systems of
73 the form $Ax = b$.²⁷ For A_r the r th row of A , and A_r^* its conjugate transpose, the following
74 update rule is used.

$$x_{l+1} = x_l + \frac{b_r - A_r x_l}{\|A_r\|_2^2} A_r^* \quad (5)$$

75 The row index r is chosen sequentially with $r = l + 1 \pmod{m}$.

76 Wei adapts the algorithm for the case where equality of magnitudes $|Ax| = b$ is desired
77 and phase is a free parameter (the generalised phase retrieval problem,²⁵). The l th phase
78 term is determined as $\theta_l = \angle A_r x_l$, after which the original update rule has a natural
79 extension.

$$x_{l+1} = x_l + \frac{b_r e^{j\theta_l} - A_r x_l}{\|A_r\|_2^2} A_r^* \quad (6)$$

80 Each iteration can be thought of as projecting the previous solution x_l onto the hyper-
81 plane H_r defined as follows.

$$H_r^\theta = \left\{ x : A_r x = b_r e^{j\theta} \right\} \text{ where } \theta = \angle A_r x_l \quad (7)$$

82 The above choice of θ is optimal in the sense that the distance from x_l to H_r^θ is minimised
83 for $\theta = \angle A_r x_l$.

84 2.2 Randomized Kaczmarz Algorithm

85 The randomized Kaczmarz algorithm was developed more recently by Strohmer and Ver-
86 shynin,²⁰ as they demonstrate an exponential convergence for a randomized version of the
87 simple Kaczmarz method described in the previous section. This algorithm is shown to
88 outperform the simple Kaczmarz method in terms of convergence rates, converging to a
89 minimum faster than the simple Kaczmarz method.²⁸ Wei derives further analytic conver-
90 gence results for the analogous phase retrieval use case.²⁵

91 This proposed version of the algorithm iterates through rows of A randomly rather
92 than sequentially. The probability of choosing each row in matrix A is proportional to the
93 square of its Euclidean norm.

$$P\{r = k\} \propto \|A_k\|_2^2 \quad k = 1, 2, \dots, m \quad (8)$$

94 **3 Holography using Kaczmarz algorithm**

95 The Kaczmarz algorithm has been shown to be effective in multiple applications such
96 as computer tomography, digital signal processing, microscopy as well as metasurface
97 antennas. In this paper, however, we demonstrate it can also be effectively used in two-
98 dimensional holographic projections in the Fraunhofer region. Wei²⁵ demonstrated this
99 algorithm possessed phase retrieving properties as well as the possibility of converging
100 better than other more established algorithms such as Gerchberg-Saxton and Wirtinger
101 flow. Nevertheless, this algorithm has to be adapted to suit our problem of projecting a
102 holographic image in the far-field region.

103 Rather than using Fast Fourier Transforms (FFTs), as in Gerchberg-Saxton, Kacz-
104 marz phase retrieval algorithm uses DFT matrices to iteratively recover the original target
105 image. Multiplying a two-dimensional DFT matrix with the hologram will reproduce a
106 two-dimensional replay field in the far-field.

107 For this linear system algorithm to work, A is the DFT matrix, x is the image holo-
108 gram and b is the target image. Both x and b are vectors. Much like Gerchberg-Saxton,
109 the algorithm starts with a random hologram and the product of the DFT matrix and the
110 hologram is consistently compared with the target image b . DFT matrices usually refer
111 to the 1D DFT matrix, however, we had to create a 2D DFT matrix for this algorithm to
112 work accordingly. The expression below demonstrates how the linear system was adapted
113 to implement holography.

$$\begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} x_{11} \\ \vdots \\ x_{m1} \end{pmatrix} = \begin{pmatrix} b_{11} \\ \vdots \\ b_{m1} \end{pmatrix}$$

114 2D DFT Holo Target

115 One of the challenges faced in this project was constraining the amplitude of the holo-
116 gram. While the replay field is naturally constrained, it was found that this was not the case
117 in the diffraction plane. As previously mentioned, the SLM has certain restrictions and in
118 order to successfully project an image in the replay field, one must construct a hologram
119 under these constraints. One of these constraints is the amplitude of the hologram must be
120 equal to one.

121 **4 Kaczmarz Convergence of Holographic Images**

122 The following test images of 128x128 pixels, seen in Fig 2, were used throughout this
123 investigation. The performance of Kaczmarz in projecting a replay field in the far-field
124 was compared to Gerchberg-Saxton. The Wirtinger algorithm, expressed by Wei²⁵ is op-
125 timized for images in near-field applications^{18,29} and this is not compared here as this

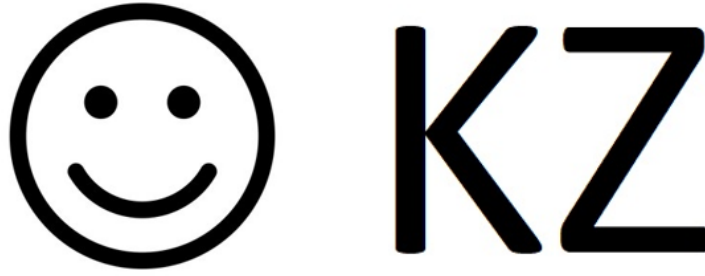


Fig 2: Target Images used throughout this investigation

126 represents a different optical configuration where the image is formed before the far-field.
127 Each algorithm was run for ten iterations and their respective image reconstructions were
128 quantitatively analysed and compared.

129 *4.1 Simple Kaczmarz*

130 The first variant of the algorithm presented is the simple Kaczmarz algorithm. Fig 3
131 displays the performance of Kaczmarz and Gerchberg-Saxton after 10 image iterations.
132 One iteration of Kaczmarz is counted after all image rows have been processed whereas
133 Gerchberg-Saxton performs the fast Fourier transform of one image all at once. Whilst
134 Gerchberg-Saxton utilises a more computationally-efficient mechanism, Kaczmarz offers
135 an interesting approach to projecting in the far-field and converges to a minimum in fewer
136 iterations when compared to GS, as seen in Fig 4.

137 *4.2 Randomized Kaczmarz*

138 The randomized version of this algorithm is also compared to Gerchberg-Saxton and while
139 it is similarly converging faster after 10 iterations, the image quality looks poorer than
140 the simple Kaczmarz version presented earlier. This is illustrated in Fig 5. Much like
141 simple Kaczmarz, the error metrics in Fig 6 show randomized Kaczmarz converges in
142 fewer iterations when compared to Gerchberg-Saxton.

143 **5 Experimental Results**

144 This section compares the simple Kaczmarz algorithm to Gerchberg-Saxton on a holo-
145 graphic projector. Much like the previous section, the image quality of 10 iterations of
146 Kaczmarz will be compared to 10 iterations of Gerchberg-Saxton. It is worth noting the
147 comparison between both will be made using a binary SLM as opposed to the multi-phase
148 approach adopted in the previous section. Therefore, rather than comparing these results to
149 simulations we compared them with respect to each other under the exact same conditions.

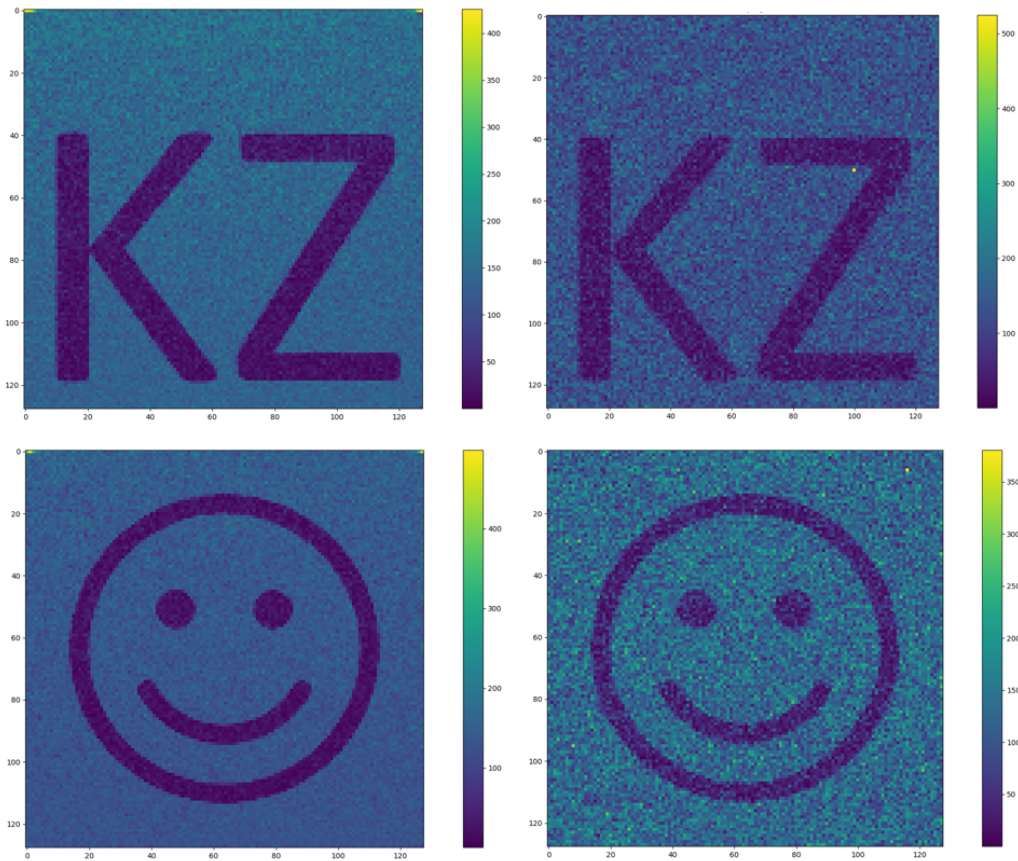


Fig 3: Images retrieved through Simple Kaczmarz (Left) and Gerchberg-Saxton (Right)

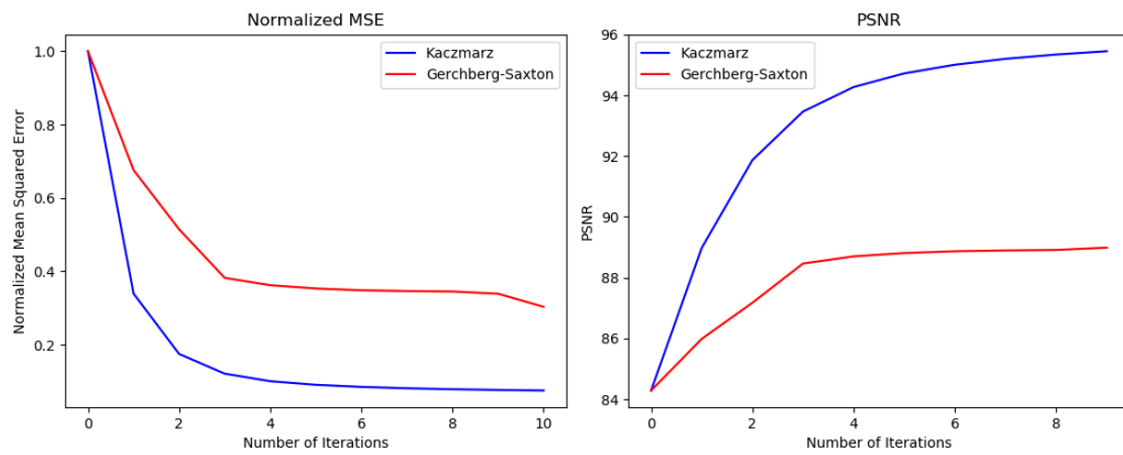


Fig 4: Error Metrics of Simple Kaczmarz and Gerchberg-Saxton

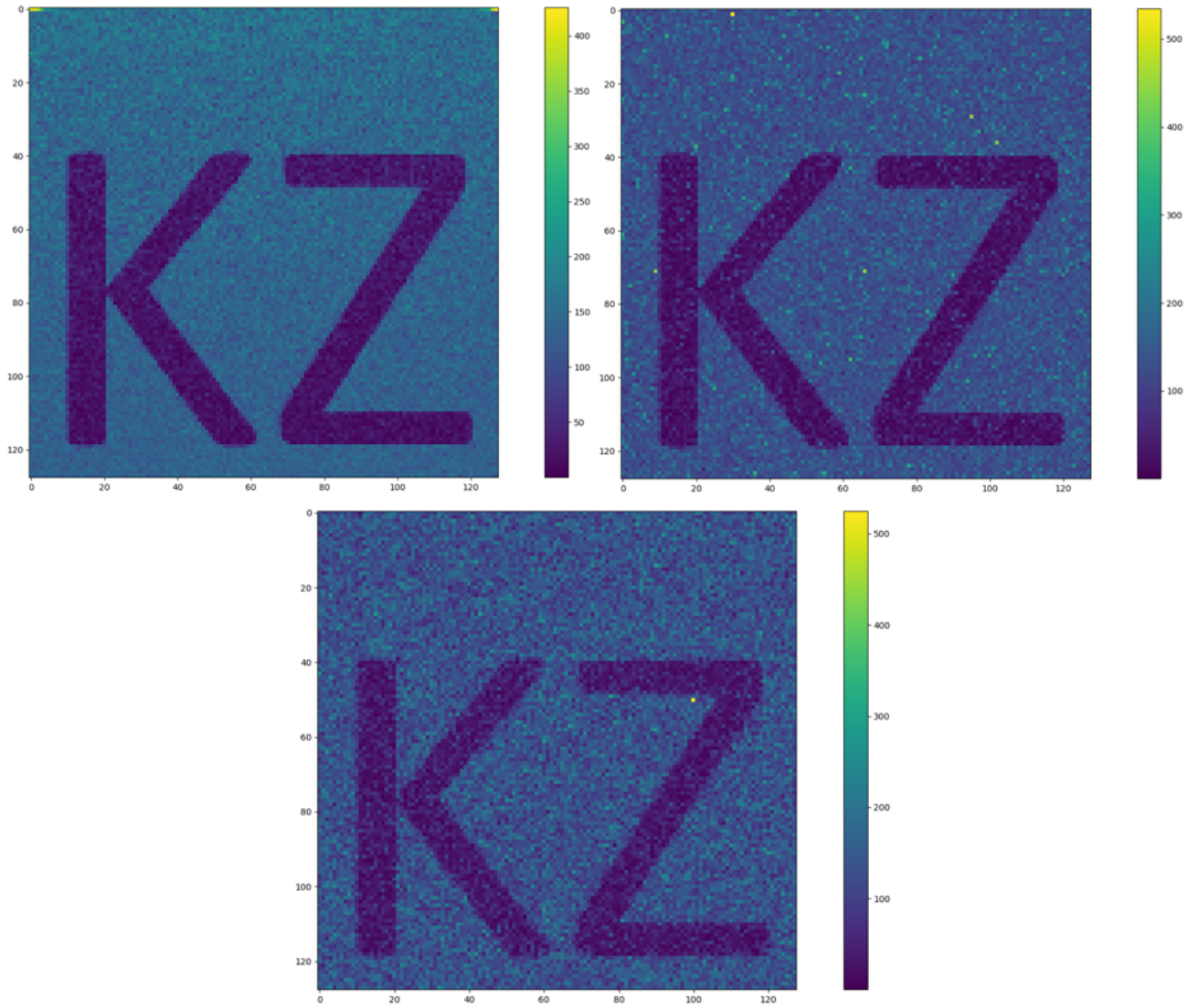


Fig 5: Images retrieved through Simple Kaczmarz (Top Left), Randomized Kaczmarz (Top Right) and Gerchberg-Saxton (Bottom)

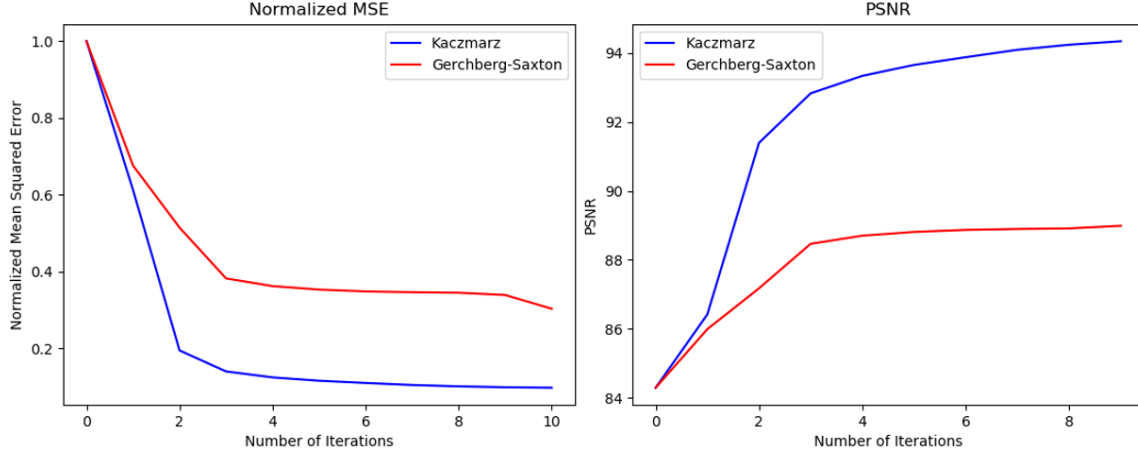


Fig 6: Error Metrics of Randomized Kaczmarz and Gerchberg-Saxton

150 A picture of the projector and its setup can be seen in Fig 7. A 532 nm laser is expanded
 151 by lens L1 (*focal length 5mm*) and then collimated by L2 (*focal length 150mm*) before
 152 reaching the SLM. The beam is then subsequently scaled down by lens L2 and polarized
 153 after passing the beam-splitter. It is then collimated by lens L3 (*focal length 3.4mm*) before
 154 being projected on a plane surface ahead of the projector. The SLM used is a 4DD SXGA
 155 ferroelectric LC device with 1280x1024 pixels.

156 Fig 8 demonstrates the results obtained. It is noticeable that, much like what was
 157 verified in the previous section, the simple Kaczmarz mechanism displays a higher quality
 158 image when compared to Gerchberg-Saxton after 10 iterations.

159 6 Future Work

160 This investigation presents a new mechanism to projecting a holographic image in the
 161 far-field. Whilst it is comparatively slower than algorithms like Gerchberg-Saxton, the
 162 improved image quality seen in the results above could be used if, for instance, a section
 163 of a larger image would require quality enhancement.

164 Arguably, whilst using matrices and vectors to project these replay field leads to im-
 165 proved image quality, it also leads to very large memory usage. The multiple matrix
 166 computations are both computationally and memory intensive, especially when compared
 167 to Gerchberg-Saxton. For images that require that extra quality, however, it becomes a
 168 legitimate option.

169 We have analysed its performance on image quality but there are other metrics that are
 170 an equally interesting avenue to pursue. These include, for instance, assessing the effects
 171 on speckle noise and camera shot noise. Iterating on a pixel by pixel basis in the replay

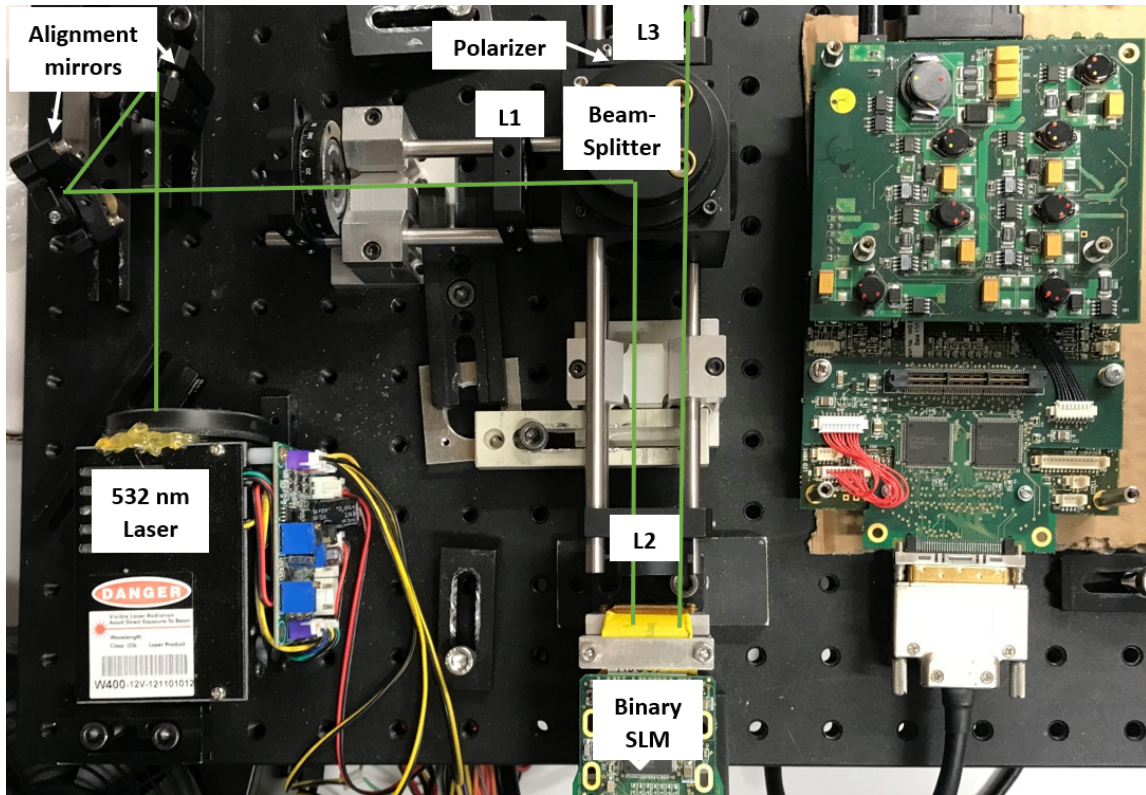


Fig 7: Holographic Projector Setup

172 field gives the user finer granularity control over each pixel. We have seen significant
 173 improvements in image quality but we need to analyse the effects on noise further.

174 7 Conclusions

175 This method provides an alternative to using FFTs by doing a series of deterministic matrix
 176 computations. FFTs determines all pixels of whole images at once while Kaczmarz iterates
 177 row by row in the DFT matrix and pixel by pixel in the replay field. However, while it is
 178 comparatively slower than Gerchberg-Saxton, it can be seen from Figures 3 and 5 that the
 179 image quality is significantly improved after the same number of iterations.

180 After simulating the effects of using Kaczmarz in holography we have successfully
 181 implemented the simple version of this algorithm on a holographic projector with binary
 182 SLM and we have shown the simple Kaczmarz demonstrates a higher image quality when
 183 compared to Gerchberg-Saxton after 10 iterations. This is due to the finer granularity
 184 control over individual pixels.

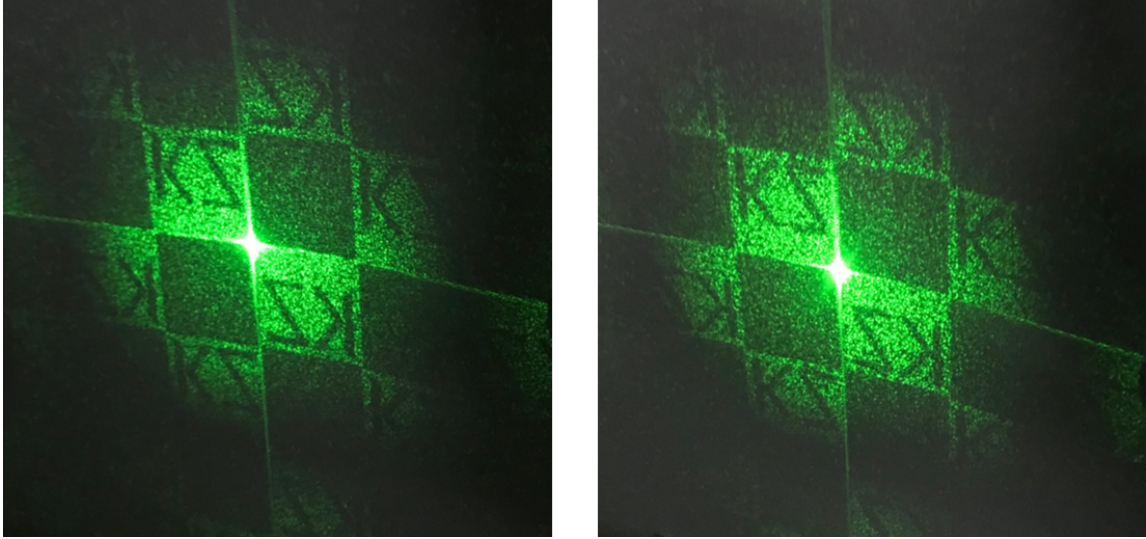


Fig 8: Kaczmarz (left) and Gerchberg-Saxton (right) replay fields projected from a binary SLM. MSE for images are 105.4 and 109.5 for Kaczmarz and Gerchberg-Saxton respectively. The PSNR is 55.9 and 55.7 for Kaczmarz and Gerchberg-Saxton respectively.

185 While there is still work to optimize this mechanism further, this investigation shows
186 the Kaczmarz algorithm can be successfully implemented in holography and results demon-
187 strate higher quality replay fields compared to Gerchberg-Saxton.

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192 **Disclosures**

193 The authors declare no conflicts of interest.

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