Daylight imaging of SSA targets through distributed volume non-Kolmogorov and anisotropic deep turbulence at low elevation angles.

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Abstract
Monte-Carlo simulations featuring Kolmogorov phase screens have long been used in understanding both the effects of atmospheric turbulence on imaging and the performance of turbulence mitigation methods. We have developed a new, high fidelity, non-Kolmogorov phase screen model that allows the orientation of anisotropy to vary arbitrarily. Our model is used to generate simulated video sequences featuring a common satellite-imaging target in low-elevation, daylight imaging scenarios, where these effects are thought to be common. These sequences demonstrate the effects of anisotropic, non-Kolmogorov turbulence, which can be unintuitive. We also briefly examine how these non-traditional features affect the performance of certain post-processing techniques, such as speckle imaging.

1. INTRODUCTION
In Daylight Imaging (DI) of Space Situational Awareness (SSA) targets many objects of interest are often visible from ground-based imaging assets only at low elevation angles. Imaging at low elevation angles is challenging for a number of reasons and even more so during daytime. By far the most challenging aspect is the length of the ground-to-space observation path through the atmosphere. Considering the long path length turbulence is certain to be severe confounding image recovery algorithms. Also challenging is the uncertainty regarding the nature of turbulence along these paths. Recent evidence [1-3] suggests that turbulence along these paths does not obey traditional Kolmogorov, 5/3 power-law fluctuation statistics. This work also suggests that turbulence may be highly anisotropic [3]. The deviation of turbulence from these two norms has been explored in depth analytically by a number of authors and accurate, high fidelity modeling on these effects via simulation is active area of research.

In this work we focus, primarily, on visualizing the effects of non-Kolmogorov and anisotropic turbulence on imaging during DI engagement scenarios using sequences of simulated images. The key development enabling this comparison is a high-fidelity non-Kolmogorov phase screen model. Monte-Carlo phase screen simulations are indispensable in both understanding the effect of turbulence, of any kind, on imaging system and especially in understanding the limits of turbulence mitigation methods such as Adaptive Optics, and image reconstruction methods such as Speckle Imaging (SI) [4] and Multi-Frame Blind Deconvolution (MFBD) [5]. To that point, we briefly examine the effect of anisotropy and changing power law exponent on the ability of SI methods to recover images from multiple turbulence corrupted images. In general, we find that smaller power-law exponent increase image blur though images are subject to relative less tip-tilt distortion. On the other hand, power-law exponents larger than the standard 5/3-value result in increased tip-tilt with less overall blurring. As one might expect, the effect of anisotropy is to preferentially blur images along the direction of the anisotropy. Images recovered using SI-techniques from simulated, turbulence-degraded imagery with these features are surprisingly similar. This similarity implies that turbulence strength and image frame SNR is more critical than the specific nature of the turbulence distortion, i.e. power-law exponent or anisotropy.

2. BACKGROUND
Monte-Carlo simulation involving Kolmogorov phase screens have been use for many years and are a standard tool in atmospheric optics. Though other methods exist [6,7] most commercially available atmospheric optics products, such as WaveProp, make use of phase screens generated using FFT-based techniques. FFT-based methods begin with a matrix, \( W_n \), of complex circular Gaussian random numbers with the same number of samples as the desired phase screen. The inverse Fourier transform of the product between an appropriately sampled covariance matrix,

\[
\phi_n = \mathcal{F}^{-1}\{W_n \ast (\Phi(\kappa))^{1/2}\}^{-1}
\]

results in two independent phase screens with the correct spatial statistics. In Eq. (1), \( \Phi(\kappa) \), is the Power Spectral Density (PSD) model of the turbulence fluctuation statistics.
Monte-Carlo simulations featuring Kolmogorov phase screens have long been used in understanding both the effects of atmospheric turbulence on imaging and the performance of turbulence mitigation methods. We have developed a new, high fidelity, non-Kolmogorov phase screen model that allows the orientation of anisotropy to vary arbitrarily. Our model is used to generate simulated video sequences featuring a common satellite-imaging target in low-elevation daylight imaging scenarios, where these effects are thought to be common. These sequences demonstrate the effects of anisotropic, non-Kolmogorov turbulence, which can be unintuitive. We also briefly examine how these non-traditional features affect the performance of certain post-processing techniques, such as speckle imaging.
A recognized weakness of this method is an under sampling at low spatial frequencies. This weakness is especially worrisome considering that most of the turbulence energy in a volume of atmosphere is found in the outer scales and therefore also the lower spatial frequencies. Lane [8] originally suggested what has become known as the “sub-harmonic” compensation method. Lane’s method samples the lowest spatial frequency region of the PSD model enhancing the accuracy of the model. Frelitch [9], and later Schmidt [10], both describe an implementation of the sub-harmonic technique including 3 sub-harmonics, though generalizing the technique to additional samples is a trivial task.

Apart from some early, generalized, work by Martin [11] and Coles [12] there has been little attention paid to the non-Kolmogorov power-law phase screen in the area of optical wave propagation in the atmosphere. Most work in the area, up to now, has focused on the development of analytical models. The first non-Kolmogorov phase screen model was reported by Gudimetla [13]. Recently, Roggemann [14], and separately Bos [15] have presented work on simulations involving non-Kolmogorov phase screens. In both cases, the user may select the power law exponent as well as inner and outer scale. The latter are modeled using a generalized form of the analytical Tatarkii spectrum in terms of radial wavenumber $\kappa$ and power-law exponent $\alpha$ is

$$
\Phi(\kappa) = A(\alpha)^{\beta} \exp\left(-\frac{\kappa^2}{\kappa_m^2}\right) \left(\frac{\kappa^2 + \kappa_0^2}{\alpha^2}\right)^{\alpha/2}
$$

In Eq.(2), $A(\alpha) = \frac{1}{4\pi^2} \Gamma(\alpha - 1) \cos\left(\frac{\alpha \pi}{2}\right)$ is a parameter that maintains consistency between structure function and power spectrum models of index of refraction fluctuations. The parameter $\beta$ is a strength parameter equivalent to $C_n^2$ when $\alpha = 11/3$. The variables $\kappa_0 = 2\pi / L_0$ and $\kappa_m = c(\alpha) / l_0$ parameters for the outer-scale, $L_0$, and inner-scale, $l_0$, respectively with $c(\alpha)$ being the inertial-dissipative consistency parameter [18]. Both Bos [15] and Roggemann [14] use include Lane’s sub-harmonic compensation method. The model used here adds the ability to vary both the degree and direction of anisotropy. Anisotropy is introduced following the model described by Wheelon [16] and later Gudimetla [17] via affine transform of the spatial coordinates.

$$
\tilde{\mathbf{f}} = (f_x^2 + f_y^2)^{1/2} = (ab)^{1/2} \left(\frac{1}{a} f_x^2 + \frac{1}{b} f_y^2\right)^{1/2}
$$

Varying the direction of the anisotropy is enabled via a second affine transform of the covariance-sampling matrix.

$$
f_x' = f_x \cos \theta + f_y \sin \theta 
$$

$$
f_y' = -f_x \sin \theta + f_y \cos \theta
$$

3. GENERATED SIMULATED IMAGERY USING PHASE SCREEN MODEL

![Fig. 1: Pristine SEASAT Image](image-url)
Phase screens generated using the technique described above were used to generate simulated imagery of a common “SEASAT” (Fig.1) SSA target corrupted by atmospheric turbulence. Simulation was carried out by placing a single phase screen across $D=3.5\text{m}$ imaging aperture using the technique described by Roggemann [4]. When not specified, turbulence strength was fixed at $D/r_0 = 10$. Taking a new draw of the circular complex Gaussian random matrix, $W_n$, creates new, independent phase screens and subsequently, new, statistically independent images. By placing these images in sequence we can create video sequences that either demonstrate the range of fluctuations under static conditions or visualize the effect of a changing variable. Keystones for several of these video sequences are shown in the next section.

4. VIDEO SEQUENCES: EFFECT OF VARIED $D/r_0$ FOR VARIOUS POWER LAW EXPONENTS

Fig. 2: $D/r_0 = 10$ to 40, $\alpha = 11/3$ (Kolmogorov turbulence)

Fig. 3: $D/r_0 = 10$ to 40, $\alpha = 3.1$ (non-Kolmogorov turbulence, near lower bound)

Fig. 4: $D/r_0 = 10$ to 40, $\alpha = 3.8$ (non-Kolmogorov, near upper bound)

Fig. 2 shows several frames from a video sequence where the turbulence strength, in terms of $D/r_0$, is varied between 10 and 40 while the power-law exponent is fixed at $\alpha = 11/3$, or typical Kolmogorov turbulence. The full video sequence begins with a relatively unperturbed image that gradually becomes more distorted and also shifts randomly around the image frame, both characteristics of turbulence-induced imaging distortions. Fig. 3 is similar to Fig. 2 but the phase screen power-law exponent has been changed to $\alpha = 3.1$. We observe that the tip-tilt/random motion effects are nearly absent in this sequence while relative blurring is more severe. Finally, in Fig. 4 the power-law exponent has been changed to $\alpha = 3.8$. In Fig. 4 both blurring and tip-tilt distortions are more severe, the latter being more pronounced than in the other two cases. We note briefly here that the structure function definition of turbulence is limited to a range of $3 < \alpha < 5$. However, the wave structure function, and all parameters derived therefrom are
limited to the range $3 < \alpha < 4$. Our phase screen model uses the Fried parameter to set turbulence strength. Of course, the Fried parameter is defined as the point where the Kolmogorov wave structure function is equal to 6.88.

Fig. 5: Degree of anisotropy varied from 0 to 3 at $D/r_0 = 20$

We now focus on the topic of anisotropic turbulence, the presence of which is thought to be commonplace in the upper atmosphere. Accordingly, any low-elevation angle, earth-to-space imaging path will traverse long sections of the upper atmosphere and, therefore, likely subject to anisotropic turbulence. In Fig. 5 we demonstrate the effect of degree of anisotropy, $\varepsilon$, from 0 to 3. In our model, this variation results in anisotropy along the vertical direction. As expected, in Fig. 5 we observe a very clear increase in horizontal blurring as the anisotropy factor is increased.

5. SIMPLE SI IMAGE RESTORATION

In this section, we compare images reconstructed using a research SI processing technique from sequences of turbulence imagery. Our SI processing technique uses no SNR weighting or other optimizations and is identical to the bispectrum method described in [19]. As in that work, 5 sub-planes of the bispectrum are averaged to determine the image phase at each spatial frequency. Our objective here is to determine if changes in power-law or anisotropy may adversely affect reconstruction quality. In all cases, all parameters are fixed and 100 input frames are used for each reconstruction. To begin, additive and photon-counting noise effects are not included.

Fig. 6: 100-frame long-exposure PSF at $\alpha=3.66$ (11/3) (left). 100 frame SI-reconstruction at $D/r_0 = 10$ (right)

Fig. 7: 100-frame long-exposure PSF at $\alpha=3.1$ (left). 100 frame SI-reconstruction at $D/r_0 = 10$ (right)
We consider the isotropic, Kolmogorov results shown in Fig. 6 (right) as the baseline for comparison in this section. To that point, we observe that the PSF in both Fig. 7 and Fig. 8 are wider than the baseline. Likewise, in Fig. 8 the PSF is wider and, by examining the video sequence a considerable larger bulk tip-tilt component is observed. Comparing reconstruction results (right) in each figure we observe, in general, little difference in reconstruction quality between images. That is to say, there are no features in the baseline that cannot be observed in the other images. On a finer scale, both non-Kolmogorov cases result in a degradation of image quality. In the “low” power-law case ($\alpha = 3.1$) the image is blurred compared to the baseline. It is possible that additional tuning of the noise parameter in the pseudo-Weiner inverse filter may further improve this image. In contrast, the “high” power law case ($\alpha = 3.8$) contains imaging artifacts common in SI-reconstructions where the object drifts beyond the image frame. It is often possible to correct these artifacts so, while undesirable, they don’t rule out the use of speckle imaging here.

Fig. 9: 100-frame, long-exposure PSF (left) and SI-reconstruction (right) for non-Kolmogorov turbulence with anisotropy factor $\varepsilon=2$.

Fig. 10: 100-frame, long-exposure PSF (left) and SI-reconstruction (right) for non-Kolmogorov turbulence with anisotropy factor $\varepsilon=2$. The direction of anisotropy varies continuously from 0 to 180 degrees.
In Fig. 9 we show an example turbulence-degraded image, PSF, and the 100-frame reconstruction for turbulence with an anisotropy factor of $\varepsilon = 2$. In contrast to Fig. 5, atmospheric distortions are more severe in the horizontal, rather than vertical direction in Fig. 9. Relative turbulence strength is kept fixed at $D/r_0 = 10$. The reconstruction in Fig. 9 is subjectively equivalent to, if not slightly better than, the baseline reconstruction in Fig. 5. This result is expected considering the use of the affine transform in our phase screen model. Consequently, when anisotropy is increased along one axis it is decreased in the other effectively increasing coherence. Thus, we might expect improved resolution in the vertical direction relative to the isotropic case. In Fig. 9 it is difficult to conclude one way or the other if such improvement is present. Considering the cursory nature of this exercise we will leave such exploration for another time.

In Fig. 10, like Fig. 9, input images are blurred by anisotropic turbulence with a factor of $\varepsilon = 2$. However, rather than being fixed along one axis, the orientation of the anisotropy changes. Anisotropy is directed along the horizontal axis at the start of the sequence, as the sequence progresses the orientation changes; falling along the vertical axis at the midpoint and returning to the horizontal axis again by the last frame. By way of implementation the angle of anisotropy varies from 0 to 180 degrees. Assuming fixed anisotropy one can approximate this scenario as one where the orientation of the satellite varies relative to the anisotropy over the course an observation pass. Examining the reconstructed frame in Fig. 10 we observe both more contrast than Fig. 5 or Fig. 9 but also more prominent artifacts. We speculate that additional tuning of the noise parameter may ameliorate the differences producing an imaging with both improved contrast and/or detail and less prominent processing artifacts. Finally, we observe that the PSF in Fig. 10 (left) is no smaller than the isotropic, Kolmogorov case.

Fig. 11: Sample image frame (left) and SI-reconstruction (right) in including Poisson and Gaussian noise for isotropic, Kolmogorov turbulence.

Fig. 12: Sample image frame (left) and SI-reconstruction (right) in including Poisson and Gaussian noise for anisotropic turbulence with Kolmogorov power-law exponent.

Fig. 11 and Fig. 12 repeat the cases in Fig. 5 and Fig. 10 but include both additive Gaussian noise and Poisson noise associated with photo-detection events. Additive noise in this case models both the background noise and read-noise associated with the detector. Noise in this case was zero mean with a deviation of $\sigma_n = 100$. Poisson noise was modeled by using normalizing intensity to unit energy and scaling by $K = 5,000,000$ photo-events, $D/r_0 = 10$ in all cases. Images were zero thresholded to approximate the decrease in SNR associated with increased background illumination during DI scenarios. In Fig. 11 and Fig. 12 the image on the left is an example input image and the right
the SI output. Though a point result, the anisotropic results in Fig.12 are subjectively better than the isotropic, Kolmogorov case in Fig.11. Again, this result is not unexpected, as anisotropy results in increased coherence in one direction. Fig.12 also includes faint artifacts hinting that severe anisotropy may pose problems for speckle imaging. In Fig.13 and 14, example frames and point reconstructions are shown for power-laws of $\alpha=3.1$ and 3.8 respectively. The reconstruction in Fig.13 is degraded compared to Fig.11. The results here make sense if we compare the long-exposure PSF in Fig.6 and 7 and assume that the effect of the smaller power law is in effect the same as an increase in turbulence strength. In contrast, we judge Fig.11 and Fig.14 to be roughly equivalent. Though the PSF in Fig.8 is also wider than Kolmogorov case, as we discussed previously, the majority of the distortion here is from tip-tilt. The bispectrum is immune to in-frame tip-tilt distortion, and so, does not degrade image quality.

6. CONCLUSIONS

We have developed a new and novel method of producing turbulence phase screens with several features important to modeling DI engagements. We have used this model to produce video sequences featuring both non-Kolmogorov turbulence and turbulence with varied degree and direction of anisotropy. The improved modeling capability presented here will allow for the simulation of complex engagement scenarios where parameters such as turbulence strength, power-law, degree and direction of anisotropy all vary simultaneously. Preliminary results, shown here, indicate that basic SI techniques remain effective when turbulence is non-Kolmogorov and for reasonable levels of anisotropy even in the presence of Poisson and additive noise associated with increased background illumination. The one possible exception being scenarios where high power-law exponents increase turbulence-induced in-frame jitter. Optimistically, this jitter could be corrected using a Fast-Steering Mirror or by dropping frames where the object impingens on the edge of the frame.

Fig. 13: Sample image frame (left) and SI-reconstruction (right) in including Poisson noise for anisotropic turbulence with non-Kolmogorov power-law exponent of $\alpha=3.1$.

Fig. 14: Sample image frame (left) and SI-reconstruction (right) in including Poisson noise for anisotropic turbulence with non-Kolmogorov power-law exponent of $\alpha=3.8$. 
7. References


