Fringing Analysis and Simulation for the Vera C. Rubin Observatory’s Legacy Survey of Space and Time

Zhiyuan Guo (郭致远)1 ©, C. W. Walter1 ©, Craig Lage2 ©, and Robert H. Lupton3 ©
The LSST Dark Energy Science Collaboration

1 Department of Physics, Duke University, Durham, NC 27708, USA; zhiyuan.guo@duke.edu
2 Department of Physics, University of California, Davis, One Shields Avenue, Davis, CA 95616, USA
3 Princeton University, Princeton, NJ, USA

Received 2022 November 21; accepted 2023 February 22; published 2023 March 27

Abstract

The presence of fringing in astronomical Charge-Coupled Device (CCD) images will have an impact on photometric quality and measurements. Yet its impact on the Vera C. Rubin Observatory’s Legacy Survey of Space and Time (LSST) has not been fully studied. We present a detailed study on fringing for CCDs already implemented on the Rubin Observatory LSST Camera’s focal plane. After making physical measurements and knowing the compositions, we developed a model for the e2v CCDs. We present a method to fit for the internal height variation of the epoxy layer within the sensors based on fringing measurements in a laboratory setting. This method is generic enough that it can be easily modified to work for other CCDs. Using the derived fringing model, we successfully reproduce comparable fringing amplitudes that match the observed levels in images taken by existing telescopes with different optical designs. This model is then used to forecast the expected level of fringing in a single LSST y-band sky background exposure with Rubin telescope optics in the presence of a realistic time-varying sky spectrum. The predicted fringing amplitude in LSST images ranges from 0.04% to 0.2% depending on the location of a CCD on the focal plane. We find that the predicted variation in surface brightness caused by fringing in LSST y-band sky background images is about 0.6 μJy arcsec^-2, which is 40 times larger than the current measurement error. We conclude that it is necessary to include fringe correction in the Rubin’s LSST image processing pipeline.

Unified Astronomy Thesaurus concepts: Astronomical instrumentation (799); CCD observation (207); Observational cosmology (1146); Astronomy data modeling (1859)

1. Introduction

Fringing in astronomical Charge-Coupled Device (CCD) detectors results from the interference of incident and reflected light between multiple layers within the CCD chip. The non-uniformity of the layers’ thickness creates varying interference conditions that lead to the observed fringing pattern, which becomes prominent in the Near-Infrared (NIR) as the declining absorption coefficient of the photosensitive silicon layer makes the silicon more transmissive to photons. Night-sky emission lines produced by excited molecules and radicals in the upper atmosphere are the primary source of the light that causes fringing in direct imaging. The Vera C. Rubin Observatory’s Legacy Survey of Space and Time (hereafter LSST) aims to explore six optical bands (u, g, r, i, z, y) ranging from 320 to 1000 nm (LSST Science Collaboration et al. 2009). The fact that non-trivial fringing patterns are observed in night images taken by a LSST prototype sensor with LSST y filter at a Naval Observatory (Brooks et al. 2017) suggests that fringing may appear in images taken by Rubin Observatory’s camera. It is important to account for the fringing effects properly to obtain accurate measurements, especially for LSST NIR images since observations in the NIR are crucial for transient studies such as supernovae. Thus, a deep understanding of LSST fringing patterns is needed and a fringing model needs to be added to the LSST image simulation tool (LSST Dark Energy Science Collaboration et al. 2021) in order to verify that the fringe removal algorithm in the LSST image processing pipeline works properly.

The LSST focal plane array consists of 189 fully depleted 4k × 4k pixel CCD sensors made by two manufacturers, Arizona Image Technology Laboratory (ITL) and Teledyne-e2v (e2v). The photosensitive silicon regions of these sensors are made 100 μm thick to improve the IR sensitivity (O’Connor 2019). These back-illuminated CCDs are assembled into 21 Raft Tower Modules (RTM) in groups of 9 on the science focal plane (O’Connor et al. 2016).

Knowing the composition of a CCD’s structure, namely the material of each layer, the fringing amplitude can be deduced from geometrical optics calculation. Many efforts have been
made to account for fringing in images taken by previous telescopes. Malumuth et al. (2003) devised a multi-layer fringing model and used the observed fringing amplitude across Hubble Space Telescope (HST)’s STIS CCD to derive the spatial variation of the photosensitive silicon layer. Walsh et al. (2003) used a similar method to fit the thickness of HST ACS WFC and HRC CCD layers. In this study, we develop a fringing model for LSST e2v sensors and fit the model to flat field data obtained from electro-optics (EO) test stands at the SLAC National Accelerator Laboratory for nine e2v CCDs in one LSST science raft using a pixel-by-pixel fitting method adopted from Malumuth et al. (2003). By fitting the fringing amplitude derived from flat fields, a thickness map of the epoxy layer that causes the fringing observed in the lab is obtained for each of the nine sensors. The fringing model is then used to forward model realistic fringing in LSST sensors based on telescope optics and sky emission lines.

This paper is structured as follows. In Section 2, we present the fringing model and the e2v CCD structure used in this study. This model is then used to verify the epoxy layer that gives rise to the fringing pattern observed in lab data. Fringing in ITL CCD is also discussed in this section. In Section 3, we describe the data reduction and fitting algorithm and present the sensor fitting results. In Section 4, we discuss sky emission lines, LSST telescope optics, and their relevance to simulating realistic fringing in a sky background image. In Section 5, we verify the fringing model by applying it to simulate the fringing amplitudes of CCD sensors implemented on MonoCam and Hyper-Suprime Camera and compare those with observed values. Finally, the forward modeling of fringing in LSST sky background images is presented.

## 2. LSST Fringing in General and Simulation

### 2.1. Fringing Simulation in General

Based on previous studies (Malumuth et al. 2003; Walsh et al. 2003), a multilayer optical model can be constructed for simulating fringing if the material composition for every layer of a CCD is known. The problem is equivalent to determining the electromagnetic solution for a light wave traveling through stratified thin films, which can be solved by using the Transfer-Matrix Method. One key feature of this method is that the finite series of interfering waves between layers are implicitly taken into account in the calculation. In this study, we use TMM, a python package developed by Byrnes (2016), to solve such problems. We only illustrate key equations here and refer readers to the reference above for more details.

If \( r_{p,p+1} \) and \( t_{p,p+1} \) denote the reflection and transmission going from \( p \)th to \( p+1 \)th layer, \( d_p \) as the thickness of layer \( p \) and \( k_p \) as the wavevector normal to the layer interfaces of the \( p \)th layer, then the relations between \( v_{p,p} \) and \( v_{p,p+1} \), the amplitude of the forward traveling wave at \( p \)th, \( p+1 \)th side and, \( w_{p,p+1} \), the amplitude of backward traveling wave at \( p \)th, \( p+1 \)th side, can be derived to be as:

\[
\begin{pmatrix}
v_p \\
w_p
\end{pmatrix} = M_p \begin{pmatrix}
v_{p+1} \\
w_{p+1}
\end{pmatrix}
\]

\[
M_p \equiv \begin{pmatrix}
e^{-i\delta_p} & 0 \\0 & e^{i\delta_p}\end{pmatrix} \begin{pmatrix}1 & 1 \\r_{p,p+1} & 1\end{pmatrix} \begin{pmatrix}1 \\1_{p,p+1}\end{pmatrix}
\]

with \( \delta_p = d_p k_p \). The matrix relates the wave entering and exiting the stack via Equation (14) in Byrnes (2016):

\[
\begin{pmatrix}1 \\
r\end{pmatrix} = \begin{pmatrix}1 \\
M_{00} & M_{01} \\M_{10} & M_{11}\end{pmatrix} \begin{pmatrix}t \\
0\end{pmatrix}
\]

where

\[
M_{0,1,0,1} = \frac{1}{r_{0,1}} \begin{pmatrix}1 & r_{0,1} \\1 & 1\end{pmatrix} M_1 M_2 \cdots M_{N-2}.
\]

Thus, the transmitted, \( T \), and reflected power, \( R \), can be calculated as follows:

\[
R = |r|^2
\]

\[
s\text{-pol:} \quad T = |t|^2 \frac{\text{Re}[n \cos \theta]}{\text{Re}[n_0 \cos \theta_0]}
\]

\[
p\text{-pol:} \quad T = |t|^2 \frac{\text{Re}[n \cos \theta^*]}{\text{Re}[n_0 \cos \theta_0^*]}
\]

where \( \theta_0 \) and \( \theta \) are the light propagation angles in the previous and present layer calculated based on Snell’s law, \( n \) is the refractive index, \( s\text{-pol} \) stands for \( s\text{-polarized} \) light that has an electric field polarized perpendicular (parallel) to the plane of incidence, the total absorption power of the stack is given by:

\[
A = 1 - T - R.
\]

Since electron–photon pairs are generated in the detection layer of CCD, we are interested in the absorption in the silicon detection layer of the stack model. This is achieved in TMM by calculating the energy flow (Poynting vector) at the beginning of each layer.

\[
s\text{-pol:} \quad S_\phi = \frac{\text{Re}[\langle n \rangle (\cos \theta)(E_f^s + E_b^s)(E_f - E_b)]}{\text{Re}[n_0 \cos \theta_0]}
\]

\[
p\text{-pol:} \quad S_\phi = \frac{\text{Re}[\langle n \rangle (\cos \theta^*)(E_f + E_b)(E_f^* - E_b^*)]}{\text{Re}[n_0 \cos \theta_0^*]}
\]

where \( E_f \) and \( E_b \) are the E-field for the forward and backward traveling wave at that point (\( z = 0 \)) in the layer of interest. Based on the assumption that all reflected light remains in the initial layer and all transmitted light gets absorbed in the final layer, the power absorption in each layer can be obtained by taking the difference of energy flow calculated above between each consecutive pair of layers. The fringing model...
implemented in this study is performed on a pixel-by-pixel level. All the calculations presented in this paper are polarization-averaged. Further extensions to this model are possible, and have been employed by other authors. For example, Malumuth et al. (2003) modified the above treatment by including roughness parameters. However, our goal in this work is solely to accurately predict fringing patterns due to sky emission lines, and we have determined this model to be accurate enough for this purpose with no other extensions.

Finally, it should be noted that for point sources and other compact objects, the scale of interest is the Point-Spread Function (PSF). However, studying the impact of fringing on photometry over a scale as small as that of the PSF, which may by necessity include the influence of the CCD readout structures is beyond the scope of this paper. We leave this work to a future study.

2.2. e2v CCD 250

The multi-layer optical model used to characterize the structure of e2v CCD used in this study is derived from the physical measurements from Lage (2019) as such information is not provided by the vendor. Detailed descriptions of the e2v CCD stack model and corresponding parameters are listed in Table 1. The top layer (layer 0) in the model is a vacuum, in which the photons travel before encountering the CCD. Layers 1–6 constitute the CCD chip structure. For the anti-reflection (AR) coating (layers 1 and 2), we use the AR coating material compositions and values derived through fitting the Quantum Efficiency (QE) measurement for a LSST CCD (A. Rasmussen 2023, private communication). Based on an inspection of the cross-section of the sensor, the e2v CCD is glued to a supporting silicon (layer 8) via a epoxy layer (layer 7) (Lage 2019). The final layer (layer 9) is the substrate beneath the whole stack. The multi-layer stack model implemented in TMM requires the ambient medium and substrate to have infinite thickness.

The calculations introduced in Section 2.1 are sensitive to the optical properties of the materials in the stack model. References to measured refractive indices, as a function of wavelength, of the materials implemented in the fringing model are listed in Table 1. The temperature dependence of silicon’s refractive index and extinction coefficient is included in the model. The normalized temperature coefficients from Green (2008) are used to calculate those two values at different temperatures. It is noteworthy that the measured doping level of the p-type doped silicon in LSST CCD is about $2 \times 10^{12} \text{cm}^{-3}$ (Lage 2019). This doping level is too small to significantly impact the optical properties of silicon (Jellison et al. 1981).

Previous studies (Malumuth et al. 2003; Walsh et al. 2003) attributed the fringing patterns observed in other CCD sensors to the spatial variations of the silicon detection layer of CCDs. However, this is not the case for the fringing pattern observed in the e2v sensors. For light with normal incidence, which is a good assumption for the lab data that will be discussed later, the minimum resolution of the monochromator required to resolve fringing related to a material with a certain thickness at a particular wavelength is given by:

$$\delta \lambda = \frac{\lambda^2}{2nd + \lambda}$$

where $n$ is the refractive index, $d$ is the thickness and $\lambda$ is the wavelength at which the fringing spacing is evaluated. For a 100 $\mu$m thick silicon layer with $n_{si} = 3.6$ at $\lambda = 960$ nm, the spacing between two adjacent fringes is about 1.2 nm. This implies that an illumination bandwidth of less than 1.2 nm is needed to resolve the fringes related to spatial variations of the silicon layer. With any bandwidth greater than 1.2 nm, these fringes will be smeared out by the large bandpass of the illumination setup.

This argument is also supported by simulation. Based on the stack model shown in Table 1 with a constant epoxy thickness of 14 $\mu$m, two sets of simulations with different assumptions for illumination bandwidths, 1 and 2 nm, are generated. A Gaussian distribution is assumed for the monochromatic light profile in which the illumination bandwidth is the Full Width at Half Maximum of the Gaussian. The two panels in Figure 1 show absorption power for the stack as a function of wavelength for the two bandwidths respectively. The upper panel shows the calculation results for 2 nm bandwidth and the calculation for 1 nm bandwidth is presented in the lower panel. With a 2 nm illumination bandwidth, only one fringing pattern with 19 nm spacing between fringes is observed in the simulation result. Using Equation (1), it can be easily verified that this spacing corresponds to fringing caused by the 14 $\mu$m epoxy layer with $n_{epoxy} = 1.6$ at around 960 nm. With 1 nm bandwidth, a second fringing pattern with a much smaller amplitude appears, as shown in the inset figure in the lower

| Layer | Purpose | Material | Thickness (μm) |
|-------|---------|----------|---------------|
| 0     | Ambient medium | Vacuum | Inf.         |
| 1     | AR coating | MgF$_2$ (Li 1980) | 0.1221 |
| 2     | AR coating | Ta$_2$O$_5$ (Rodríguez-de Marcos et al. 2016) | 0.0441 |
| 3     | Detection | Si (Green 2008) | 100 |
| 4     | Gate Oxide | SiO$_2$ (Malitson 1965) | 0.1 |
| 5     | Gates | Si | 0.3 |
| 6     | Insulation | SiO$_2$ | 1.0 |
| 7     | Adhesive | Epoxy | CCD dependent |
| 8     | Support | Si | 165 |
| 9     | Substrate | Si$_3$N$_4$ (Philipp 1973) | Inf. |

The calculations introduced in Section 2.1 are sensitive to the optical properties of the materials in the stack model. References to measured refractive indices, as a function of wavelength, of the materials implemented in the fringing model are listed in Table 1. The temperature dependence of silicon’s refractive index and extinction coefficient is included in the model. The normalized temperature coefficients from Green (2008) are used to calculate those two values at different temperatures. It is noteworthy that the measured doping level of the p-type doped silicon in LSST CCD is about $2 \times 10^{12} \text{cm}^{-3}$ (Lage 2019). This doping level is too small to significantly impact the optical properties of silicon (Jellison et al. 1981).

Previous studies (Malumuth et al. 2003; Walsh et al. 2003) attributed the fringing patterns observed in other CCD sensors to the spatial variations of the silicon detection layer of CCDs. However, this is not the case for the fringing pattern observed in the e2v sensors. For light with normal incidence, which is a good assumption for the lab data that will be discussed later, the minimum resolution of the monochromator required to resolve fringing related to a material with a certain thickness at a particular wavelength is given by:

$$\delta \lambda = \frac{\lambda^2}{2nd + \lambda}$$

where $n$ is the refractive index, $d$ is the thickness and $\lambda$ is the wavelength at which the fringing spacing is evaluated. For a 100 $\mu$m thick silicon layer with $n_{si} = 3.6$ at $\lambda = 960$ nm, the spacing between two adjacent fringes is about 1.2 nm. This implies that an illumination bandwidth of less than 1.2 nm is needed to resolve the fringes related to spatial variations of the silicon layer. With any bandwidth greater than 1.2 nm, these fringes will be smeared out by the large bandpass of the illumination setup.

This argument is also supported by simulation. Based on the stack model shown in Table 1 with a constant epoxy thickness of 14 $\mu$m, two sets of simulations with different assumptions for illumination bandwidths, 1 and 2 nm, are generated. A Gaussian distribution is assumed for the monochromatic light profile in which the illumination bandwidth is the Full Width at Half Maximum of the Gaussian. The two panels in Figure 1 show absorption power for the stack as a function of wavelength for the two bandwidths respectively. The upper panel shows the calculation results for 2 nm bandwidth and the calculation for 1 nm bandwidth is presented in the lower panel. With a 2 nm illumination bandwidth, only one fringing pattern with 19 nm spacing between fringes is observed in the simulation result. Using Equation (1), it can be easily verified that this spacing corresponds to fringing caused by the 14 $\mu$m epoxy layer with $n_{epoxy} = 1.6$ at around 960 nm. With 1 nm bandwidth, a second fringing pattern with a much smaller amplitude appears, as shown in the inset figure in the lower
Figure 1. Absorption power vs. wavelength of e2v CCD based on different illumination bandwidths. Upper panel: Calculation based on 2 nm bandwidth ($d_{\text{epoxy}} = 14 \mu m$). Spacing between green dashed lines: 19 nm fringe spacing for the first fringing pattern related to the epoxy layer. Lower panel: Calculation results based on 1 nm bandwidth. Spacing between red dashed lines: 1.2 nm fringe spacing for the second fringing pattern related to the Si detection layer.

Panel. At around 960 nm, the second set of fringes has a spacing of about 1.2 nm, coming from the 100 \mu m silicon detection layer with monochromatic light of normal incidence.

Before being assembled into the focal plane, LSST CCDs were sent to Brookhaven Nation Laboratory (BNL) for comprehensive tests and integration using the eotest software.\(^5\) The sensors were first tested individually on a single-sensor test stand (BNL-TS3). Those that passed the qualification were assembled onto rafts and each raft was tested in BNL test stand (BNL-TS8). The rafts were then sent to SLAC National Accelerator Laboratory with an identical test stand setup for re-testing. The data analyzed in this paper come from SLAC Test Stand 8 (henceforth SLAC-TS8). The monochromatic light used to obtain flat field data is generated from a 4\(^4\) integrating sphere which is set 1 m away from the CCD. Thus, a collimated beam of light falls at normal incidence is a good approximation, which is the assumption that all the calculations in this section are based on. Figure 2 shows an example of the fringing pattern observed in one particular e2v CCD taken under monochromatic light of approximately 2 nm bandpass in SLAC-TS8. It is noticeable that only one fringing pattern is observed in this e2v CCD. This finding implies that the observed fringing should correspond to the thickness variation in the epoxy layer since any fringing related to the silicon detection layer will be smeared out by the large bandpass of the illumination setup.

As mentioned earlier, BNL-TS8 and SLAC-TS8 have an identical test stand setup. However, BNL-TS8 has a narrower monochromator bandpass compared to SLAC-TS8. Data taken for e2v sensors in BNL-TS8 indeed shows a second set of fringing patterns, which should be related to the 100 \mu m silicon based on the conclusions drawn from simulations. However, for this study, we only focus on fringing in the epoxy layer. This is because simulation results imply that the amplitude of the fringes from the epoxy layer is much greater than that from the silicon layer. Additionally, when simulating realistic fringing in telescope image, the simple assumption of normal incident light for lab illumination setup will be replaced with telescope optics. The large aperture of the telescope tends to average out the fringing patterns and further decreases the observed amplitude (Groom et al. 2017), making the impact of this second fringing pattern in the silicon layer negligible compared to the one in the epoxy layer. See Sections 2.3 and 4.2 for more discussion.

2.3. ITL STA3800C CCD

Fringes are not observed in ITL sensors at SLAC-TS8. The reason is two-fold: first, as mentioned in Section 2.2, the silicon layer fringing is expected to be washed out by the large bandpass of SLAC-TS8 illumination. Second, because the ITL sensor has an additional Litho-Black coating applied to it. This highly absorbent black coating will absorb light passing through the epoxy layer and greatly reduce the amount of reflected light from the epoxy. With illumination light coming from a monochromator with a smaller bandpass compared to SLAC-TS8, fringing is observed in ITL sensors at BNL-TS8. However, studies have shown that when using a f/1.2 beam, which closely resembles the overall focal ratio of the LSST telescope (f/1.23) (Ivezić et al. 2019), fringing is not observed in ITL sensors even with 1 nm monochromator bandpass (L. Craig 2023, private communication). This fact supports the argument we made at the end of Section 2.2. We conclude that fringing caused by the thickness variation of 100 \mu m silicon layer will be trivial for LSST therefore this study focuses on the fringing pattern related to the epoxy layer in e2v sensors.

3. Fitting Thickness of the Epoxy Layer

3.1. SLAC Test Stand Flat Fields Data

To simulate the observed fringing pattern from the fringing model, the thickness of the epoxy layer ($d_{\text{epoxy}}$) must be derived at each pixel across the CCD. This is achieved by fitting for $d_{\text{epoxy}}$ based on the observed fringing amplitude as a function of wavelength for every pixel of a CCD. Thus, the number and sampling of available data points are crucial for

---

\(^5\) https://github.com/lsst-camera-dh/eotest
constraining the thickness of the epoxy layer. A series of flat fields at different wavelengths were measured for e2v CCD sensors with monochromatic light illumination in SLAC-TS8 under the temperature of $-90^\circ$C. The smallest wavelength spacing between each successive flat field data available in SLAC-TS8 is 10 nm for nine e2v sensors implemented on RTM-020. As it will be shown in Section 3.4, we are able to obtain reasonable fitting results with this sampling in wavelength. In the following sections, we describe the method used to derive the epoxy thickness map for all of the nine CCDs.

3.2. Data Reduction

The flat field data are preprocessed using LSST Data Management software (Axelrod et al. 2010; Jurić et al. 2017). This includes an overscan correction that removes the average signal introduced by reading a CCD and bias subtraction which helps to de-bias a CCD image by subtracting the pixel-to-pixel structure in the read noise from the raw image. To further reduce the noise level of the preprocessed flat field images, a Gaussian filter with a kernel size of 16 x 16 pixels is applied to the overscan corrected image.

Figure 3 shows the depth in silicon at which 99% of the incident light is expected to be absorbed, as a function of wavelength under two different temperatures. For LSST CCDs with silicon thickness of 100 $\mu$m operating under $-90^\circ$C, the silicon will start becoming transparent to incident light at 800 $\mu$m. At this point, light will reach the epoxy layer below the 100 $\mu$m silicon and some will be reflected back to interfere with incident light in previous layers. Fringing will become more apparent at longer wavelengths as more light reaches the back of the sensor. Based on this conclusion and combined with a visual inspection of the test data, fringe flat field data ranging from 880 to 990 nm in steps of 10 nm are chosen to fit the fringing amplitude and derive the value of epoxy layer thickness. The fringing amplitude at an individual pixel is defined as the number of counts per pixel over the overall mean number of counts in the image with unity subtracted:

\[
\text{Fringe Amp.} = \frac{\text{Counts}}{\text{Overall Mean}} - 1. \quad (2)
\]

3.3. Pixel by Pixel Fitting Algorithm

To infer the thickness variation map of the epoxy layer across an entire sensor, we adopt the fitting method described in Malumuth et al. (2003). Since only the variation of $d_{\text{epoxy}}$ determines the frequency of fringes, the thickness of all the other layers is assumed to be known and constant across the sensor. The boundary between each layer is assumed to be planar for simplicity (Table 1). All the calculations are based on the assumption of collimated beam and 2 nm illumination bandwidth. The algorithm contains the following steps:

Step 1: An arbitrary pixel (we pick pixel $X = 2100$, $Y = 2100$ in the case of e2v-CCD250-321) is chosen as the starting pixel. Then simulated fringing amplitude from 880 to 990 nm is calculated for a range of epoxy thickness $d_{\text{epoxy}}$ ranging from 5 to 30 $\mu$m. The value of $d_{\text{epoxy}}$ that minimizes the reduced $\chi^2$ of the fit to the observed fringing amplitude is chosen as the best fit, $d_0$, for this starting pixel. The reduced $\chi^2$ is defined as $\chi^2/(n_d - n_p)$, with $n_d$ and $n_p$ being the number of data points and a number of fitting parameters respectively. Figure 4 presents the fitting results for this pixel.

Step 2: We then move to the next pixel in the same column ($X = 2100$, $Y = 2101$). Using the derived thickness value of the initial pixel, $d_0$, as a reference point, we calculate the fringing amplitudes for a set of $d_{\text{epoxy}}$ values within the range of one order of fringe, 0.3 $\mu$m, centered on that value ($d_0 \pm 0.15 \mu$m). The order of fringes can be related to varying thickness of the epoxy layer ($\Delta d$) via (Janesick et al. 1987):

\[
\Delta d = \frac{\delta \lambda \cos \theta}{4\pi n_{\text{epoxy}}} \quad (3)
\]

In the case of normal incidence, one order of fringe ($\delta = 2\pi$) corresponds to approximately $\Delta d_{\text{epoxy}} \approx 0.3 \mu$m. Within this given range, the value of $d_{\text{epoxy}}$ minimizing the reduced $\chi^2$ of the fit is assumed to be the thickness of this pixel. This value will be updated as the reference point for the next pixel that the algorithm will be working on. Looking for potential best-fit
values of \(d_{\text{epoxy}}\) in a limited range in this way helps to ensure the derivation of a smooth thickness variation map of the epoxy layer.

Step 3: Step 2 is repeated until reaching the end of the column \((X = 2100, Y = 4000)\). Then we move down to the next column in that same row \((X = 2101, Y = 4000)\), and work up the row in the same manner as described in the previous steps.

Step 4: The above steps are repeated until reaching the end \((X = 4000, Y = 4000)\). Upon this point, we move back to the initial pixel \((X = 2100, Y = 2100)\) and repeat the same process in reverse order until reaching pixel \(X = 1, Y = 1\).

3.4. Fitting Results

Figure 5 shows the derived epoxy layer thickness map of e2v-CCD250-321 in the left panel. Using the fringing model, we successfully reproduce the observed fringing pattern, which is shown in the right panel in Figure 5, based on this derived height variation map. Figure 6 shows the comparison between Gaussian-smoothed real data and simulation results for a specific row of the sensor. It is clear that the phases and amplitudes are in agreement between simulation and real data.

Using this method, we generate the fringing patterns observed in the other eight sensors in the same RTM from the simulation. Figure 7 shows the simulated fringing patterns for all nine CCDs. Most of the fringing patterns have been successfully recovered.

4. Recipes for Realistic Fringing Simulation in Sky Background Image

With knowledge of the compositions of sensor structure and height variation in the epoxy, we can further use the fringing model to predict the expected level of fringing in LSST images. In this section, we discuss all the ingredients needed to simulate fringing in real sky images captured by a telescope in general.

4.1. Hydroxyl Radical (OH) Emission Lines

The night sky spectrum is dominated by the emission lines produced by the rotational and vibrational transitions of hydroxyl (OH) radicals. Each vibrational transition produces a band in the observed spectrum and the transition between rotational levels associated with the two vibrational levels gives rise to the fine structure of the band. The vibration-rotation spectrum of the hydroxyl radical was first observed by Meinel (1950a, 1950b). These narrow emission lines are the main sources that give rise to fringing in the observed images. The intensity of vibrational bands and the population of rotational levels within each band can be well described by Boltzmann distributions specified by vibrational temperatures, \(T_{\text{vib}}\), and rotational temperatures, \(T_{\text{rot}}\). The typical values of \(T_{\text{vib}}\) and \(T_{\text{rot}}\) are about 10,000 K and 200 K respectively (Rousselot et al. 2000). These lines are subject to both temporal and spatial variations. Nevertheless, the relative intensities of rotational transition lines are expected to remain roughly the same since \(T_{\text{rot}}\) varies much less than \(T_{\text{vib}}\) (Noll et al. 2015; Hart 2019).

As discussed in Section 3.2, fringing will start becoming prominent in e2v sensors as the wavelength goes beyond 880 nm. Real images will be taken with telescope filters, and the wavelength range most relevant to fringing falls within the bandpass of the LSST y filter. The top panel in Figure 8 shows the throughput curve of LSST y-band filter and the OH
emission spectra taken by the Prime Focus Spectrograph (PFS) (R. H. Lupton 2023, private communication) of the Subaru telescope (Tamura et al. 2016). As confirmed by observations and theoretical calculations in previous studies, there are six transitions between vibrational bands \( 7 \rightarrow 3, 8 \rightarrow 4, 3 \rightarrow 0, 9 \rightarrow 5, 4 \rightarrow 1, 5 \rightarrow 2 \) (Osterbrock et al. 1996, 1997; Rousselot et al. 2000; Noll et al. 2015) that fall within the LSST \( y \)-band. The color coding of the lines in Figure 8 represents the vibrational group a line belongs to.

When simulating fringing in the presence of OH lines, we assigned a normalized weight, \( w_{\text{line}} \), to each individual line. This weight is calculated based on the combination of each line’s relative intensity and the value of the LSST \( y \) filter throughput curve at the corresponding wavelength. The bottom panel in Figure 8 shows the normalized weight assigned to each line. Due to the low throughput of the LSST \( y \) filter at wavelength bluer than 900 nm and redder than 1050 nm, the contributions to fringing from lines in vibrational groups 7–3 and 5–2 is small compared to those from other groups. The final simulation result, \( S_{\text{total}} \), is obtained by coadding the simulation image for each individual line, \( S_{\text{line}} \), together. This coaddition is performed as a weighted average:

\[
S_{\text{total}} = \sum_{\text{line}} w_{\text{line}} S_{\text{line}}.
\]  

4.2. Telescope Optics

A telescope has a finite aperture that makes the incident light come from a range of angles rather than solely at normal incidence. Thus the previous assumption of collimated beams used in simulating lab results does not hold. Light arriving at different angles tends to average out the observed fringing amplitude (Groom et al. 2017). The Vera C. Rubin Observatory telescope is a three-mirror design. The three aspheric mirrors, an 8.4 m primary mirror, a 3.4 m convex secondary mirror, and a 5.0 m tertiary mirror, give an overall focal ratio of \( f/1.234 \) (Olivier et al. 2008; Bond et al. 2018; Ivezić et al. 2019). To accurately count the range of light incident angles from the \( f/1.234 \) beam on different positions of the LSST focal plane in
fringing simulation, we employ Batoid. Batoid is a C++ based python optical raytracer package that characterizes the optical performance of survey telescopes based on geometric optics developed by Meyers et al. (2019). Figure 9 demonstrates several examples of light paths inside the LSST telescope generated by Batoid. In Batoid, the directions of the incident beams are described by incoming slopes in two directions, $\frac{dx}{dz}$ and $\frac{dy}{dz}$. The incident angle $\theta$ on the incidence plane, one of the inputs of the fringing model, can be derived from the two slopes:

$$\theta(\text{rad}) = \arctan \sqrt{\left(\frac{dx}{dz}\right)^2 + \left(\frac{dy}{dz}\right)^2}.$$ 

Figure 10 shows the incident slope distributions of light landing on CCDs in two different locations on the LSST focal plane.
plane, one with the sensor being located in the center and the other being close to the edge of the LSST focal plane.

The incident inclination of the LSST beam makes it possible for absorbed photons in the photosensitive region to travel into neighboring pixels instead of staying in the same pixel where it initially landed as in the case of normal incidence. To check if this may affect the fringing simulation results in a significant way, we did a detailed simulation in which each absorbed photon is tracked to its final location where the electron–photon pair is generated in the photosensitive region of CCD. Figure 11 presents the distribution of the distances photons have traveled in x and y directions as specified by the two
slopes before getting absorbed in Silicon for 1000 photons assumed to be landed at the center of a pixel. The two slopes will also change upon refraction into silicon. The two cases presented in Figure 11 correspond to the angle distributions shown in Figure 10. The color bar indicates the z direction (depth) that a photon has traveled in Silicon. Each pixel of LSST CCD sensor is 10 μm in width and length, and 100 μm in depth (Ivezić et al. 2019). Simulation results in Figure 11 indicate that inter-pixel migration of absorbed photons does exist. However, even in extreme cases where light lands on the focal plane edge, 98% of the photons will travel less than 10 μm, which corresponds to the size of one pixel. Since the typical size of the observed fringe is about 20–30 pixels, widening by one pixel will merely affect the simulation result.

Figure 8. Top panel: Light blue line: OH emission line from the Prime Focus Spectrograph (PFS) of the Subaru telescope (Tamura et al. 2016). Red dashed line: LSST y filter throughput curve. Bottom panel: Blue line: Normalized output of OH line intensity as weighted by LSST y filter throughput.

Figure 9. The light path inside the LSST telescope calculated using Batoid (Meyers et al. 2019).

Figure 10. Left panel: Incident slope distribution of LSST beam landed on the center of the focal plane. Right panel: Incident slope distribution close to the edge of the focal plane.
Thus we consider this effect to be negligible for the purpose of simulating the fringing pattern of 4000 by 4000 pixels images. Thus, to save computational time, an absorbed photon is always assumed to end up in the same pixel as the one it initially landed on. Additionally, the distribution of incident slopes of light is assumed to be constant for all pixels of a CCD since this distribution is expected to vary very slowly across the focal plane. To account for the range of incident angles, the fringing simulation at a given wavelength is obtained via a weighted average of simulations over all the angles. We can further write Equation (4)

$$S_{\text{total}} = \sum_{\text{line}} \sum_{\text{line}} w_{\text{line}} w_{\theta} S_{\theta}$$

(5)

where $S_{\theta}$ is an individual fringing simulation for a given incident angle $\theta$ with normalized weight $w_{\theta}$ derived based on the angle distributions. Equation (5) implies that simulating fringing in sky background images with real telescope optics properly requires coadding simulations over all the OH lines and over all the light incident angles.

5. Fringing Simulation Results

To verify the robustness of the fringing model in simulating real sky images, we first apply it to simulating sky background fringing of MonoCam (Brooks et al. 2017) and Hyper-Suprime Camera (HSC) (Miyazaki et al. 2018) before making predictions for LSST. In this section, we first discuss the comparison between simulation results and observations in terms of the optics setup of MonoCam and HSC. Then we use the fringing model to predict the expected level of fringing in LSST sky background images. All simulations in this section follow the methods described in Section 4.

5.1. Fringing of MonoCam

MonoCam was a camera employing a single LSST prototype e2v-CCD250 sensor, with a reported fringing amplitude of around 2%. The data were taken with a 1.3 m reflector telescope with an overall focal ratio of $f/4.0$ and with LSST $y$ filter under a temperature of $-120^\circ$C (Brooks et al. 2017). Since the CCD used in MonoCam came from the same manufacturer as the sensor studied in this paper, the derived epoxy thickness map of e2v-321 is assumed for this prototype sensor for the purpose of fringing simulation.

The left panel of Figure 12 presents the fringing amplitude of the diagonal pixels of the simulated, noiseless image. The simulation result shows that the MonoCam fringing amplitude is about 1.5%, which agrees with the amplitude of the smoothed and noise-reduced midnight fringing pattern given in Figure 9 of Brooks et al. (2017). A full sensor image of the simulated sky background is shown in the right panel of Figure 12 based on the 60 s exposure time (Brooks et al. 2017) of MonoCam. Poisson photon noise and Gaussian read noise of CCD are added to the simulated full image by using the Galsim module (Rowe et al. 2015). It is clear that a nontrivial fringing pattern still appears in the image. These results suggest that our fringing model simulation is in good agreement with MonoCam observation.

5.2. Fringing in HSC CCD Sensor

HSC is an 870 megapixel prime focus optical imaging camera with an overall focal ratio of $f/2.0$ implemented on the
8.2 m Subaru telescope. 116 fully depleted 2048 × 4096 pixel CCDs with a thickness of 200 μm are employed in the focal plane (Miyazaki et al. 2018). The HSC optics offers an opportunity to test the fringing model’s response to a fast input beam. From the inspection of real HSC sky images, we find that fringing of HSC CCD has a sensor-dependent amplitude ranging from 0.2% to 0.6%. As an example, the left panel of Figure 13 shows the observed fringing pattern in a 1000 × 1000 pixel region of an HSC CCD. To reduce the noise and make fringes easier to see, the plotted data have been smoothed by a 16 × 16 pixel Gaussian kernel. The mini panel in the top right of the figure shows the fringing amplitude, which is about 0.3%–0.4%, along the diagonal pixel of the image. Fringes in HSC CCDs are likely to be caused by the non-uniformity in the 200 μm silicon layer. This is because they only exhibit one set of fringes with similar features as the ones observed in other back-illuminated sensors, such as the HRC CCD and WFC CCD, whose fringes are modeled based on height variation in silicon detection layer, studied by Walsh et al. (2003). Additionally, according to Miyazaki et al. (2018),
the scatter in the epoxy layer of HSC sensors is about 2 μm, which is not enough to produce the number of fringes observed in HSC images.

To simulate HSC sensor fringing, we made the following changes to the sensor model as depicted in Table 1. First, the thickness of epoxy is kept constant at 14 μm since HSC CCD fringing is caused by non-uniformity in 200 μm silicon instead of epoxy as discussed above. Second, a Gaussian Random Field (GRF) is used to characterize the variation of the detection layer. Since our goal is to verify that the fringing model can produce a comparable level of fringing amplitude as the one measured from a real HSC image, the actual fringing pattern can be arbitrary. Thus, GRF is a good approximation to the underlying thickness change of the silicon layer, whose variation can be inferred by counting the number of observed fringes across the image. Using Equation (3) and plugging in values for silicon ($n_s = 3.6$, $\lambda = 1000$ nm), we know that a 0.139 μm change in silicon thickness will give one order of fringe in HSC CCDs. A visual inspection of the HSC fringing image (left panel of Figure 13) suggests that there are 10–15 fringes across the image. This implies the overall variation of this particular patch is about 1.39–2.08 μm. These values set the scale of the GRF used to characterize the silicon layer of HSC CCD. We choose to use the average value, 1.74 μm, as the height variation of the GRF ([199.13 μm, 200.87 μm]) across 1000 by 1000 pixels. Batold is then used to generate the angle distribution of the incident beam in the center of the focal plane based on HSC telescope optics. The OH line intensities are weighted by the throughput curve of the HSC filter following the discussion of Section 4.1. In the right panel of Figure 13, we present the simulated, noiseless 2D fringing pattern and amplitude across the diagonal pixels of the simulated image. It is clear that the predicted amplitude (~0.3%) is comparable to the observed level in the left panel.

5.3. Prediction of Fringing in LSST Sky Background Images

After validating the fringing model with MonoCam and HSC optics, we simulated the expected fringing pattern of e2v-CCD250-321 based on LSST setups as described in Section 4.2. In Figure 14, we compared the simulated fringing patterns for two cases in terms of a CCD’s location on the LSST focal plane, which differ in their angle distribution of incident beam as shown in Figure 10. Compared with the values calculated in previous sections, the fringing amplitude for LSST decreased to 0.2%. This is caused by the fact that the wide range of incident angles of LSST optics decreases the overall fringing amplitude when incorporating the beams coming from all the angles in the simulation. As the sensor is placed further away from the center of the field of view, the fringing amplitude decreases to 0.04%. This is because the light arriving at the edge of the focal plane does not come from an exact f/1.23 beam due to the LSST optical design (Olivier et al. 2008), as manifested in Figure 10. This leads to a wider range of incident angles, which further decreases the overall fringing amplitude when incorporating the beams coming from all the angles in the simulation.

The left panel of Figure 15 shows a simulated LSST sky background image with Poisson photon noise and Gaussian read noise added for a single exposure of 30 s. The values have been converted from counts per pixel in the original simulation to surface brightness unit, mag arcsec $^{-2}$, by properly accounting for the sensor gain, $\gamma$-band zero-point, and pixel scale. The sky brightness level in our simulated image is close to both the expected value and in situ measurement of the sky level in LSST $\gamma$-band (High et al. 2010). Due to the low amplitude of fringing and relatively short single exposure time, fringing cannot be easily observed as the image is mostly dominated by noise. However, after applying a 2D Gaussian kernel, which serves as a low-pass filter to extract low amplitude structure on relatively large scales in image progressing, to the image, the fringing patterns become apparent as shown in the right panel of Figure 15. We discuss its implication in detail in the next section.

5.4. Impact of LSST Fringing on Measurements

During LSST survey operation, images taken by the Rubin observatory will be processed by Rubin’s LSST Data Management Science Pipelines software stack, developed by Rubin’s LSST Data Management (DM) team. We refer readers to Bosch et al. (2018, 2019) for more a detailed description of the LSST image processing pipeline. To better quantify the impact

---

6 https://smtn-002.lsst.io/
7 https://pipelines.lsst.io/
of fringing in the context of data measurement and image reduction, it is more convenient for us to convert units to \( \mu Jy \ arcsec^{-2} \). The overall variation caused by fringing across the image, as shown by the colorbar of the right panel of Figure 15, is about 0.6 \( \mu Jy \ arcsec^{-2} \). Based on the image post-processing results from HSC (Aihara et al. 2019, 2022), which uses a custom pipeline, known as hscPipe, that has been built upon the LSST Data Management codebase, the current measurement has a 10% error at the limiting surface brightness level of 26th AB magnitude, which approximates to 0.015 \( \mu Jy \ arcsec^{-2} \). The predicted level of variation from fringing is already 40 times larger than this value. In the future, with careful post-analysis, the surface brightness limit for LSST is expected to be 30th AB magnitude, which further reduces the 10% measurement error to 0.004 \( \mu Jy \ arcsec^{-2} \) (R. H. Lupton 2023, private communication). Thus, it is clear that fringing will impact measurements on single exposure images.

Principal Components Analysis (PCA) method has been proven to be an effective approach for fringing correction (Waters et al. 2020; Medford et al. 2021). In the PCA method, a set of orthogonal images/components are trained from a sample of fringing images. Linear combinations of these components can be used to construct bias fringe images, which will then be subtracted from the target image to remove fringing. However, since sky emission lines will vary temporally as mentioned in Section 4.1, fringing is also expected to vary over time. Since real LSST images are unavailable at this time, we infer the number of components needed for LSST fringing correction for a single CCD from simulation. In this study, we use the temporal variation of relative intensities of the OH lines within LSST y-band over one night from Noll et al. (2015) to generate 11 simulated sky background fringing images at each hour spanning from 7 pm to 5 am for training. For PCA calculation, we use code from the scikit-learn package (Pedregosa et al. 2011). We find that only two components are needed to fully characterize fringing in one CCD in this case, with the first component having 98% variance. The second component, which characterizes the temporal variation of fringing, has 2% variance, which is also comparable to the measured rms amplitude variations of the fringe patterns as observed by Megacam in the z-band within a given night (Prunet 2021). This implies that solely based on the one night’s worth of data of OH line variation, as given in Noll et al. (2015), the fringing pattern is expected to vary at the 2% level and we should be able to describe the fringes with two patterns for removal. However, the data from Noll et al. (2015) might not fully represent the OH line variation at the site of Rubin Observatory. Future studies should investigate a longer-term data sample to determine if the emission line data used in this study are representative enough and if night-to-night variations might be larger.

In terms of the impact of fringing in coadded images, it will be hard to properly characterize it until the effectiveness of the fringing removal algorithm of the Rubin’s LSST DM pipeline is known. Here, we study the effect of fringing in coadds for the case in which no fringing correction has been made in single exposure images from random sensors. This is done by looking at correlations between pixels at different scales for an increasing number of coadds. The coadds are simulated via stacking regions of randomly rotated single exposure sky background images, as picked randomly from the 9 sensors in Figure 7, together. Figure 16 shows the correlation function of the mean subtracted pixel fluxes as a function of pixel scale in terms of increasing the number of coadds. We use TreeCorr (Jarvis et al. 2004) to calculate the correlation...
functions. For comparison, we also did a null case in which we measured the same correlation function of a random single exposure background image without fringing. Based on Figure 16, it is clear that even at the level of 800 coadds, there is still a difference in the correlation function when compared to that of the null case. Based on the current observation strategy of LSST (Scolnic et al. 2018), the numbers of \(\gamma\)-band exposures of the Wide Fast Deep survey (WFD) and Deep Drilling Field mini-survey (DDF) after the 10 yr survey length of LSST are 180 and 2600 respectively. This implies that, without applying any fringing correction to single exposure images, fringing will still make a noticeable impact in coadded images for WFD at least. To truly characterize fringing in coadded images, we need to quantify the effectiveness of fringing removal algorithm by running simulated fringing images through the DM pipeline as a future study.

6. Summary

We have presented a fringing model for e2v CCD sensors implemented on the Rubin’s LSST camera focal plane. We have demonstrated that these observed fringes in e2v CCDs from SLAC-TS8 are caused by the thickness variations of the epoxy layer that glues the sensor and the supporting silicon together. We have shown that this model allows us to simulate the fringing patterns accurately observed in lab data on a pixel-by-pixel level.

We have shown that with sufficient flat field data taken with close wavelength spacing (<10 nm) and under certain assumptions for the illumination setups, such as illumination bandwidth and incident angle, an underlying thickness map as a function of pixel position can be derived for each sensor via the fitting algorithm adopted from previous studies. Based on the derived thickness map, we have successfully reproduced the fringing patterns observed in nine e2v sensors in one Raft Tower Module on LSST focal plane from SLAC-TS8.

We have demonstrated that, by properly incorporating all the relevant factors, such as telescope optics and OH emission lines that dominate the night sky spectra, into the fringing model, we are able to recover the observed fringing amplitude for MonoCam and HSC. We then use this model to predict the fringing pattern in LSST real sky images and find that the simulated LSST fringing amplitude ranges from 0.04% to 0.2% depending on the location of a CCD on the focal plane. Table 2 summarizes all the simulation results in terms of different optics setups and conditions.

Finally, we have shown that fringing will be nontrivial for Rubin’s LSST. In the case of single exposure, the variation in surface brightness caused by fringing is found to be 40 times larger than the current measurement error. By using a PCA method, we have shown that two components are needed to correct for fringing in single exposure images within the timescale of a single night. And more long-term fringing images are needed to account for the temporal variation in fringing caused by changes in the sky emission lines over a larger timescale. In the case of coadded images, by studying correlation functions of simulated coadd images in pixel space, we find that the fringing level is still greater than the noise level even in the case of 800 coadded images, assuming no fringing removal algorithm has been applied. It is thus necessary to include the fringing correction algorithm in Rubin’s LSST image processing pipeline and characterize the impact of fringing in coadded images.

This paper has undergone internal review by the LSST Dark Energy Science Collaboration. The internal reviewers were David Kirkby, Josh Meyers, and Andrew Bradshaw.
We are grateful to Pierre Antilogus, Jim Chiang, Mike Jarvis, Lee Kelvin, David Kirkby, Josh Meyers, Andrei Nomertoski, Paul O’Connor, Andy Rasmussen, Aaron Rodman, and Peter Yoachim for their helpful discussions.

Z.G. and C.W. were supported by the Department of Energy, grant DE-SC0010007. C.S.L. gratefully acknowledges financial support from DOE grant DE-SC0009999 and Heising-Simons Foundation grant 2015-106.

The DESC acknowledges ongoing support from the Institut National de Physique Nucléaire et de Physique des Particules in France; the Science & Technology Facilities Council in the United Kingdom; and the Department of Energy, the National Science Foundation, and the LSST Corporation in the United States. DESC uses resources of the IN2P3 Computing Center (CC-IN2P3—Lyon/Villeurbanne—France) funded by the Centre National de la Recherche Scientifique; the National Energy Research Scientific Computing Center, a DOE Office of Science User Facility supported by the Office of Science of the U.S. Department of Energy under Contract No. DE-AC02-05CH11231; STFC DiRAC HPC Facilities, funded by UK BEIS National E-infrastructure capital grants; and the UK particle physics grid, supported by the GridPP Collaboration. This work was performed in part under DOE Contract DE-AC02-76SF00515.

The contributions from the primary authors are as follows. Z.G. analyzed data, built the fringing model, formulated the fitting algorithm, developed the simulation code, and wrote the paper. C.W. supervised the project and analysis. C.L. conducted physical measurements and provided layer composition information on the e2v sensors. R.L. provided the PFS OH emission line data and advised on the characterization of the impact of fringing on single exposure images and coadds.

ORCID iDs

Zhiyuan Guo (郭致远) https://orcid.org/0000-0001-9557-9171
C. W. Walter https://orcid.org/0000-0003-2035-2380
Craig Lage https://orcid.org/0000-0002-9601-345X
Robert H. Lupton https://orcid.org/0000-0003-1666-0962

References

Aihara, H., AlSayyad, Y., Ando, M., et al. 2019, PASJ, 71, 114
Aihara, H., AlSayyad, Y., Ando, M., et al. 2022, PASJ, 74, 247
Axelrod, T., Kantor, J., Lupton, R. H., & Pierfederici, F. 2010, Proc. SPIE, 7740, 774015
Bond, T., Roodman, A., Reil, K., et al. 2018, Proc. SPIE, 10705, 107050E
Bosch, J., AlSayyad, Y., Armstrong, R., et al. 2019, in ASP Conf. Ser. 523, Astronomical Data Analysis Software and Systems XXVII, ed. P. J. Teuben et al. (San Francisco, CA: ASP), 521
Bosch, J., Armstrong, R., Bickerton, S., et al. 2018, PASJ, 70, S5
Brooks, J., Fisher-Levine, M., & Nomertoski, A. 2017, JInst, 12, C05003
Byrnes, S. J. 2016, arXiv:1603.02720
Green, M. A. 2008, Solar Energy Materials and Solar Cells, 92, 1305
Groom, D. E., Haque, S., Holland, S. E., & Kolbe, W. F. 2017, JAP, 122, 055301
Hart, M. 2019, PASP, 131, 015003
High, F. W., Stubbs, C. W., Stalder, B., Gilmore, D. K., & Toney, J. L. 2010, PASP, 122, 722
Janesick, J. R., Elliott, T., Collins, S., Blouke, M. M., & Freeman, J. 1987, OptEng, 26, 692
Jarvis, M., Bernstein, G., & Jain, B. 2004, MNRAS, 352, 338
Jellison, G. E., Modine, F. A., White, C. W., Wood, R. F., & Young, R. T. 1981, PASP, 93, 1414
Jurić, M., Kantor, J., Lim, K. T., et al. 2017, in ASP Conf. Ser. 512, Astronomical Data Analysis Software and Systems XXV, ed. N. P. F. Lorente, Shortridge, & R. Wayth (San Francisco, CA: ASP), 279
Ivezić, Ž., Kahn, S. M., Tyson, J. A., et al. 2019, ApJ, 873, 111
Lage, C. 2019, arXiv:1911.09577
Li, H. H. 1980, JPCR, 9, 161
LSST Dark Energy Science Collaboration, Abolfathi, B., Armstrong, R., et al. 2021, arXiv:2101.04855
LSST Science Collaboration, Abell, P. A., Allison, J., et al. 2009, arXiv:0912.0201
Malitson, L. H. 1965, JOSA, 55, 1205
Malumuth, E. M., Hill, R. S., Gall, T., et al. 2003, PASP, 115, 218
Medford, M. S., Nugent, P., Goldstein, D., et al. 2021, PASP, 133, 064503
Meinel, A. B. I. 1950a, ApJ, 112, 120
Meinel, I. A. B. 1950b, ApJ, 111, 555
Meyers, J. E., Kirkby, D., Thomas, D. & Administration, U.N.N.S. 2019, batoid, Version 0.1, doi:10.11578/dc.20200708.1
Miyazaki, S., Komiyama, Y., Kawamoto, S., et al. 2018, PASI, 70, S1
Noll, S., Kausch, W., Kimeswenger, S., Unteruggenberger, S., & Jones, A. M. 2015, ACP, 15, 3647
O’Connor, P. 2019, JATIS, 5, 041508
O’Connor, P., Antilogus, P., Doherty, P., et al. 2016, Proc. SPIE, 9915, 327
Oliveir, S. S., Seppala, L., & Gilmore, K. 2008, Proc. SPIE, 7018, 70182G
Osterbrock, D. E., Fulbright, J. P. & Bida, T. A. 1996, PASP, 108, 277
Osterbrock, D. E., Fulbright, J. P., Martel, A. R., et al. 1996, PASP, 108, 277
Pedregosa, F., Varoquaux, G., Gramfort, A., et al. 2011, Journal of Machine Learning Research, 12, 2825
Phillip, H. R. 1973, JlEIS, 120, 295
Prunet, S. 2021, PASP, 133, 114502
Rodríguez-de Marcos, L. V., Larruquert, J. I., Méndez, J. A., & Aznárez, J. A. 2016, OptExp, 26, 3622
Rousselet, P., Lidman, C., Cuby, J. G., Moreels, G., & Monnet, G. 2000, A&A, 354, 1314
Rowe, B. T. P., Jarvis, M., Mandelbaum, R., et al. 2015, A&C, 10, 121
Scolnic, D. M., Gris, P., et al. 2018, PASP, 130, 055301
Scolnic, D. M., Lochner, M., Gris, P., et al. 2018, arXiv:1812.00516
Walsh, J. R., Freudling, W., Pirzkal, N., & Pasquali, A. 2003, Modelling the Fringing of the ACS WFC and HRC Chips, ST-ECF Instrument Science Report ACS 2003-012
Waters, C. Z., Magnier, E. A., Price, P. A., et al. 2020, ApJS, 251, 4