Fine cophasing of segmented aperture telescopes with ZELDA, a Zernike wavefront sensor in the diffraction-limited regime

P. Janin-Potiron¹, M. N’Diaye¹, P. Martinez¹, A. Vigan², K. Dohlen², and M. Carbillet¹

¹ Université Côte d’Azur, Observatoire de la Côte d’Azur, CNRS, Laboratoire Lagrange, UMR 7293, CS 34229, 06304 Nice Cedex 4, France
e-mail: Pierre.Janin-Potiron@oca.eu
² Aix-Marseille Université, CNRS, LAM (Laboratoire d’Astrophysique de Marseille) UMR 7326, 13388 Marseille, France

Received 23 February 2017 / Accepted 4 April 2017

ABSTRACT

Context. Segmented aperture telescopes require an alignment procedure with successive steps from coarse alignment to monitoring process in order to provide very high optical quality images for stringent science operations such as exoplanet imaging. The final step, referred to as fine phasing, calls for a high sensitivity wavefront sensing and control system in a diffraction-limited regime to achieve segment alignment with nanometric accuracy. In this context, Zernike wavefront sensors represent promising options for such a calibration. A concept called the Zernike unit for segment phasing (ZEUS) was previously developed for ground-based applications to operate under seeing-limited images. Such a concept is, however, not suitable for fine cophasing with diffraction-limited images.

Aims. We revisit ZELDA, a Zernike sensor that was developed for the measurement of residual aberrations in exoplanet direct imagers, to measure segment piston, tip, and tilt in the diffraction-limited regime.

Methods. We introduce a novel analysis scheme of the sensor signal that relies on piston, tip, and tilt estimators for each segment, and provide probabilistic insights to predict the success of a closed-loop correction as a function of the initial wavefront error.

Results. The sensor unambiguously and simultaneously retrieves segment piston and tip-tilt misalignment. Our scheme allows for correction of these errors in closed-loop operation down to nearly zero residuals in a few iterations. This sensor also shows low sensitivity to misalignment of its parts and high ability for operation with a relatively bright natural guide star.

Conclusions. Our cophasing sensor relies on existing mask technologies that make the concept already available for segmented apertures in future space missions.

Key words. techniques: high angular resolution – instrumentation: high angular resolution – instrumentation: adaptive optics

1. Introduction

The James Webb Space Telescope (JWST), NASA’s forthcoming orbiting observatory, will follow the Hubble Space Telescope and provide new insights into the birth and evolution of galaxies, stars, and planets (Clampin 2014). This observatory includes a 6.5 m diameter telescope with a segmented primary mirror made of 18 hexagonal segments. To perform as a single monolithic mirror telescope and provide diffraction-limited images, a wavefront sensing and control system is required to sense and correct for any errors in the optics. The segment alignment process (e.g., Acton et al. 2012; Knight et al. 2012; Lightsey et al. 2014; Glassman et al. 2016) is a long operation including multiple steps needed to pass through commissioning (coarse alignment, coarse phasing, and fine guiding) to maintenance procedure (fine phasing and wavefront monitoring). Segments exhibit misalignment with excursions that are initially much larger than the observing wavelength and which must be aligned to a few nanometer residuals. All of these alignment processes that assume different sensing requirements are mandatory to align the telescope into a high-performance observer. The last procedure is referred to as the fine phasing process, which produces a sharp and coherent point spread function (PSF) near the diffraction limit.

One of the potential successors to JWST, the large UV/optical IR surveyor (LUVOIR) is currently studied by NASA and should provide a larger segmented aperture from eight to 16 m with a primary mirror made up of at least 36 segments (France 2016). One of the primary science goals of LUVOIR is to directly image and characterize Earth-like planets around nearby stars (Crooke et al. 2016). Such observations require contrast levels up to $10^{10}$ and wavefront stability down to a few picometer levels, calling for, amongst other features, a precise control of the cophasing errors (e.g., Yaitskova et al. 2003; Lyon & Clampin 2012; Stahl et al. 2013; Redding et al. 2014). In anticipation of the increase in the system complexity expected with these future telescopes, it is worth exploring new concepts for segment cophasing, and in particular within the diffraction-limited domain.

Most of the current cophasing sensors are based on existing wavefront sensors that are usually employed in adaptive optics (AO), but thoroughly re-adapted considering that traditional AO sensors assume continuity of the wavefront. The recovery of the segment misalignment can be directly obtained from the information in the image plane (Lofsdahl et al. 1998; Delavaquerie et al. 2010; Martinache et al. 2013; Pope et al. 2014; Janin-Potiron et al. 2016), in a pupil plane (e.g., Chanan 1989; Montoya-Martinez 2004; Esposito et al. 2005; Dohlen et al. 2006; Mazzoleni et al. 2008; Pinna et al. 2008), or in intermediate planes (e.g., Chanan et al. 1999; Cuevas et al. 2000; Chueca et al. 2008). Image plane techniques are advantageous especially for space applications since they require a limited
amount of hardware. However, pupil plane methods bring together the following advantages: (1) the relationship between the sensor and the telescope pupil locations is direct and straightforward; (2) in addition to segment alignment (piston and tip/tilt), pupil plane methods are more favorable to access segment figuring errors (defocus, astigmatism, trefoil) and segment higher order wavefront error.

Among these concepts, we here consider the Zernike wavefront sensor (ZWFS), which is based on the phase-contrast method (Zernike 1934). These concepts aim to modulate phase aberrations on an unresolved star image with a phase-shifting mask into intensity variations in a pupil plane. Over the past few years, different ZWFS kinds have been proposed in astronomy to address various applications, such as wavefront sensing in AO systems (Bloemhof & Wallace 2003, 2004; Dohlen 2004), calibration of the non-common path aberrations in ground-based facilities (Wallace et al. 2011; N’Diaye et al. 2013, 2016), and measurement of the pointing errors or focus drifts in exoplanet imagers (Zhao 2014; Shi et al. 2015). Such sensors have also been envisioned for the cophasing procedures of segmented aperture telescopes which include coarse and fine alignment regimes (Dohlen 2004; Dohlen et al. 2006; Surdej et al. 2010; Vigan et al. 2011).

For many years, ZWFS studies have been focused on the coarse phasing regime to address the discontinuous wavefront with segmented aperture telescopes on the ground under atmospheric turbulence. In this context, cophasing errors of a few tens of nanometer need to be detected in the presence of a few microns of wavefront errors that evolve at a rate of a few hundreds of Hz. During the preliminary European-extremely large telescope (E-ELT) studies, the Zernike unit for segment phasing (ZEUS) was proposed in the framework of the European southern observatory (ESO) active phasing experiment (APE, Gontes et al. 2008) to align telescope segments under seeing conditions (Dohlen et al. 2006). On-sky demonstrations showed the ability of this concept to reconstruct a discontinuous wavefront with an accuracy better than 15 nm rms (Surdej et al. 2010).

To achieve such a performance, the ZEUS design is optimized with a mask of seeing disk size. This allows the sensor to filter out the low spatial frequency content of the wavefront errors that is dominated by the atmosphere and hence, to enhance the high spatial frequency information that is related to the segment misalignment. The relative piston, tip, and tilt between the segments are then retrieved thanks to a careful analysis of the signal at the border between adjacent segments. While these alignment measurements are limited to half a wave by the ZEUS capture range in monochromatic light, Vigan et al. (2011) have shown that multilayer-absorbed-based strategies enable the iterative calibration with this sensor of piston and tip tilt errors that are larger than one wave.

Despite these encouraging results in the coarse phasing regime, the use of a ZWFS has been little unexplored in the field of fine cophasing. Neat and actively controlled alignment is nevertheless mandatory when instruments yield diffraction-limited images to reach a high, stable wavefront quality in the high-Strehl regime. In particular, this aspect is crucial for high-contrast observations of circumstellar environments with future large observatories. The current implementation for segment phasing ZEUS is well-suited for coarse alignment under seeing conditions. Still, this Zernike sensor cannot address small wavefront errors for two reasons: its seeing-disk-size mask is not suitable to produce an interpretable signal in the diffraction-limited regime and the segment edge-to-edge analysis proves limited in the presence of a faint signal. Sensing a few nanometer cophasing errors in the high-Strehl regime with a ZWFS requires a mask with a size adjusted to the resolution element diameter of the star image but also a revisiting of the sensor signal analysis.

Recently a ZWFS called ZELDA has been proposed to calibrate the residual aberrations in exoplanet direct imaging instruments. This concept was validated with success on a real exoplanet direct imager using the Spectro-Polarimetric High-contrast Exoplanet Research (SPHERE) on the very large telescope (VLT, Beuzit et al. 2008), providing measurements of small aberrations with nanometric accuracy (N’Diaye et al. 2013, 2016). ZELDA uses a diffraction-limited size mask and an algorithm based on the analysis of the whole sensor signal to retrieve residual wavefront errors. Such features make this concept attractive for the calibration of segment phasing errors.

In this paper, we investigate the use of ZELDA for the measurement of piston, tip, and tilt errors in segmented apertures within the diffraction-limited domain. In Sect. 2, we recall the principle of the concept and propose an algorithm to extract the segment phasing errors from the sensor signal. In Sect. 3, we provide a calibration scheme for segment cophasing in closed-loop operations using our sensor. In Sect. 4, we finally discuss the overall performance of the sensor.

2. Analytical approach and system response

2.1. Zernike phase filtering sensor

The ZELDA sensor uses a phase-shifting mask located in the focal plane downstream of the telescope pupil to sense aberrations on an unresolved star image (see layout in N’Diaye et al. 2013, 2016). With a relatively good centering of the focal plane mask (FPM) on the star image, the starlight contributions going through and outside the mask interfere in the relayed pupil plane, yielding a light distribution that is directly related to the phase wavefront errors in the entrance pupil $\phi$, according to the mask characteristics, that is, the diameter $d$ and the introduced phase delay $\theta$. In the following, $A$ and $D$ denote the wavelength of observation and the telescope aperture diameter.

As previously stated in N’Diaye et al. (2013, Eq. (8) of that paper), the ZELDA relayed pupil plane intensity $I$ can be expressed as

$$I(\phi) = P^2 + 2b^2 (1 - \cos \theta) + 2Pb [\sin \phi \sin \theta - \cos \phi (1 - \cos \theta)],$$

where $P$ is the amplitude pupil function and $b$ the electric field amplitude diffractioned by the FPM in the re-imaged pupil plane.

Assuming a mask angular diameter $d = 1.06 \lambda/D$ with a circular pupil gives a value $b$ close to 0.5 over the pupil. In the classical case where $\theta = \pi/2$, the previous equation becomes

$$I(\phi) = P^2 + 2b^2 + 2 \sqrt{2}Pb \sin (\phi - \phi_0),$$

where $\phi_0 = \pi/4$ represents the reference phase in the absence of aberrations, an inherent property of ZELDA. In the small aberration regime, a first-order Taylor expansion of $\phi$ allows one to retrieve the phase term from the measured intensity (N’Diaye et al. 2013). For the estimation of cophasing errors, we introduce no approximation in this equation and we directly work with the re-imaged pupil intensity, keeping the exact expression of $I$ as a function of $\phi$, as shown in Eq. (2).

2.2. Cophasing estimators

We now define a set of estimators to retrieve the cophasing errors in a segmented pupil made of $N$ segments (see Fig. 1 in which
Fig. 1. Illustration of the ZELDA cophasing sensor principle. From the intensity in the relayed pupil plane $I$, three estimators are evaluated on each segment and the system evolves in closed loop.

the axes are defined). Each segment is subject to aberrations of piston (translation along the optical axis), tip, and tilt (rotation around the $x$-axis and $y$-axis). Vectors describing these aberrations for all the segments are respectively defined as

\[ p = [p_1 \cdots p_n \cdots p_N], \]
\[ t = [t_1 \cdots t_n \cdots t_N], \]
\[ T = [T_1 \cdots T_n \cdots T_N]. \]

(3)

The segmented telescope is assumed to be a reflective system and the aberrations are expressed as the mirror displacement at the wavefront. The elements $t_n$ and $T_n$ correspond to the gradients according to the $x$- and $y$-axes. The wavefront error $h_n$ from the $n$th segment at the position $(x, y)$ is thus given by

\[ h_n(x, y) = p_n + t_n x + T_n y \quad \forall (x, y) \in S, \]

(4)

where $S$ is the surface defined by a single segment. On the $n$th segment, $\phi$ is related to $h_n$ by

\[ \phi = \frac{2\pi}{\lambda} h_n. \]

(5)

The mechanical displacement on the segment $n$, considering a reflective system, is given by $h_n/2$.

To retrieve the piston and tip-tilt aberrations with the ZELDA signal, we define three estimators, $\varphi_0$, $\varphi_1$, and $\varphi_2$, following our work with the self-coherent camera-phasing sensor (SCC-PS) in Janin-Potiron et al. (2016, Eqs. (17)–(19) of that paper). Our calculation relies on measurements over an arbitrary square zone $H$ that is centered on the segment (see Fig. 1). For the piston estimation $\varphi_0$, the signal $I$ is integrated over $H$, while for the tip-tilt estimations $\varphi_1$ and $\varphi_2$ the gradient of the signal $I$ is integrated over $H$ and therefore these estimators are expressed as

\[ \varphi_0 = \int_{H} I(\phi) dx dy, \]

(6)

\[ \varphi_1 = \int_{H} \nabla_x I(\phi) dx dy, \]

(7)

\[ \varphi_2 = \int_{H} \nabla_y I(\phi) dx dy, \]

(8)

where $\nabla_x$ and $\nabla_y$ stand for the gradient along the $x$ and $y$ axes. Using Eqs. (2) and (4), we derive their theoretical expression:

\[ \varphi_0(p_n, t_n, T_n) = 2\sqrt{2} p_n b H^2 \sin \left( \frac{2\pi p_n}{\lambda} - \phi_0 \right) \sin \left( \frac{\pi H t_n}{\lambda} \right) \]

\[ \times \sin \left( \frac{\pi H T_n}{\lambda} \right) + H^2 \left( p^2 + 2b^2 \right), \]

(9)

\[ \varphi_1(p_n, t_n, T_n) = 4\sqrt{2} p_n b H^2 \cos \left( \frac{2\pi p_n}{\lambda} - \phi_0 \right) \]

\[ \times \left( \frac{\pi H t_n}{\lambda} \sin \left( \frac{\pi H T_n}{\lambda} \right) \right), \]

(10)

\[ \varphi_2(p_n, t_n, T_n) = 4\sqrt{2} p_n b H^2 \cos \left( \frac{2\pi p_n}{\lambda} - \phi_0 \right) \]

\[ \times \left( \frac{\pi H T_n}{\lambda} \sin \left( \frac{\pi H t_n}{\lambda} \right) \right). \]

(11)

2.3. System response

As shown in Eqs. (9)–(11), each estimator depends on piston, tip, and tilt. To investigate this dependence, we assess the behavior of these estimators first with a single aberration and then, with a combination of piston, tip, or tilt on the central segment. Hereafter, we define $r$ as the radius of the circumscribed circle for a given segment. In our simulations, we arbitrarily set the $H$ size to $0.4 r$ that appears as a reasonable trade-off between signal-to-noise ratio and sensitivity to pupil shear (see Sect. 4.3.2).

Figure 2 presents the results in the case of a single aberration on the segment. On the left plot, we consider the case of piston only and present $\varphi_0$ response to the introduced piston value. As expected, the estimator $\varphi_0$ is $\lambda$-periodic. This effect represents the well-known $\pi$-ambiguity problem (see, e.g., Vigan et al. 2011) that limits the reachable capture range in which the measured piston is achieved unambiguously. The $\varphi_0$ capture range is asymmetric and ranges from $-\lambda/8$ to $3\lambda/8$ as highlighted by the green zone in Fig. 2. This $\lambda/8$-asymmetry corresponds to the reference phase $\phi_0 = \pi/4$ in the ZELDA signal as shown in Eq. (2). Such a property will have an impact on the cophasing process for pistons larger than $\lambda/8$ as discussed...
Figure 2. Normalized estimators as a function of the introduced aberrations for piston (left) and tip-tilt (right). The green zones represent the capture range for each configuration.

Figure 3. Normalized estimators \( \varphi_0 \) (left) and \( \varphi_1 \) (right) as a function of the introduced combination of piston and tip.

Hereafter in Sect. 4.2. On the right plot, we present the \( \varphi_1 \) and \( \varphi_2 \) responses to the introduced tip and tilt values. Both estimators are periodic and symmetric with a capture range that is related to \( \mathcal{H} \) and extends from \(-5\lambda/4 \) to \(5\lambda/4\). In the absence of piston, large segment tip/tilt can be estimated with our sensor signal.

Figure 3 presents the same rationale but with combined piston, tip, and tilt. The left plot represents \( \varphi_0 \) for a combined segment piston and tip. The correlation between \( \varphi_0 \) and the two introduced aberrations is clearly visible. As the tip increases in absolute value, the \( \varphi_0 \) estimation decreases following a cardinal sine shape, leading to an underestimation of the piston. Even though not critical, this dependence has an influence on the cophasing process and is studied in Sect. 4.2. A correlation is also observed on Fig. 3 (right) when measuring \( \varphi_1 \) in the presence of piston and tip. The same results are obtained if we measure \( \varphi_2 \) for a combination of piston and tilt, as suggested by Eqs. (10) and (11). As tip varies, the \( \varphi_1 \) estimation oscillates in a sinusoidal shape. The tip estimation is therefore strongly affected by the presence of piston. A sign change of the estimator is even observed for large values of piston, leading to a misinterpretation of the segment orientation. We also note that \( \varphi_1 \) can be larger than 1. This particular behavior comes from the normalization of \( \varphi_1 \) by its peak value at null values of piston and tilt. When the piston is equal to \( \lambda/8 \), the normalized \( \varphi_1 \) value reaches \( \sqrt{2} \). Both effects can prove critical in aberration measurements and impact the segment cophasing process. We discuss the operating mode in Sect. 4.2.

3. System calibration

3.1. Numerical assumptions

Numerical simulations are performed with a pupil made of 91 hexagonal segments that are distributed over five concentric rings as shown in Fig. 1. Each segment is sampled with 100 pixels from corner to corner (i.e., \( r = 50 \) pixels). In this study, no central obstruction nor spiders are taken into account as they have no incidence on the ZELDA measurement. To ensure a fine sampling within the focal plane mask and a fast computation, the ZELDA signal is calculated using the semi-analytical method based on the Matrix Fourier Transform (MFT)
described by Soummer et al. (2007). For the phase mask, we set \( d = 1.06 \lambda / D \) and sample it with 500 pixels. We consider a source in monochromatic light at wavelength \( \lambda = 600 \) nm.

### 3.2. Calibration and properties

The calibration process of the system is based on the traditional adaptive optics scheme. A calibration matrix is built by actioning successively and independently each segment in piston, tip, and tilt, and by measuring \( \varphi_0, \varphi_1, \) and \( \varphi_2 \). The amplitude of the displacements is chosen within the linear regime of the sensor. In practice, the poke values are set to 1/10 of the capture range for each estimators, that is, \( \lambda / 20 \) for piston and \( \lambda / 4r \) for tip and tilt. The calibration process is done using a perfectly aligned and flatened system. At this stage, no camera noise or photon noise are taken into account. The calibration matrix is square and has dominant diagonal terms. However, some patterns are visible outside the diagonal and are due to the correlation between the three estimators. Finally, piston, tip, and tilt errors for each segment are retrieved by solving the linear system of equations by means of the inverted calibration matrix and are applied to the segments. In any on-sky application, the calibration matrix is conventionally measured using either an internal calibration source or a stellar target. A pseudo-synthetic calibration matrix can also be built by experimentally measuring the estimators on a single segment and replicating them in all the other segments in the matrix structure deduced from simulations (e.g., Yaitskova et al. 2006). While straightforward when looking at a reasonable number of segments, the calibration matrix is avoidable by using an alternative method based on the ZELDA signal normalization.

The system evolves in a closed-loop architecture and the convergence quality is assessed by using the residual root mean squared (RMS) wavefront error \( \sigma_{\text{pupil}} \) over the entire pupil. Its analytical expression is given by

\[
\sigma_{\text{pupil}} = \sqrt{s_p^2 + \frac{5}{24} r^2 \left( s_i^2 + s_\varphi^2 + \mathbb{E} [T]^2 + \mathbb{E}[T]^2 \right)},
\]

where \( s_p \), \( s_i \), and \( s_\varphi \) stand for the piston, tip, and tilt samples standard deviation, and \( \mathbb{E} […] \) denotes the mathematical expectation. A detailed development of Eq. (12) is given in Appendix A.

### 4. Performance and discussions

We now present the results for the complete computation of a full segmented pupil in contrast to Sect. 2.3 in which the response of an individual segment has been characterized. To assess the performance of the system, we proceed to the following tests: (1) a closed-loop convergence evaluation; (2) a statistical analysis of the converging process; (3) a sensitivity analysis of hardware misalignment combined with probabilistic insights; and finally (4) a sky coverage evaluation.

#### 4.1. Closed-loop accuracy

The results of the closed-loop operation on the phasing process for different initial wavefront errors (expressed in nanometer RMS) are presented in Fig. 4. We differentiate between the results for the piston error only (left) and for the combined piston, tip, and tilt errors (right), where each case uses its specific calibration matrix that includes the considered aberrations only. A set of reference aberrations \([p_0, t_0, T_0]\) is created by drawing each \( p_n, t_n, \) and \( T_n \) from a Gaussian distribution with zero mean and \( \sigma_p, \sigma_t, \) and \( \sigma_T \) standard deviations. In order to insure an equal contribution of each aberration, the standard deviations are set according to their respective capture range as

\[
\begin{align*}
\sigma_p &= \lambda / 2, \\
\sigma_t &= 5 \lambda / 2r, \\
\sigma_T &= 5 \lambda / 2r.
\end{align*}
\]

A practical set of aberrations are then scaled using a gain factor \( \alpha \) as \( \alpha \times [p_0, t_0, T_0] \) to produce different levels of initial RMS wavefront errors. The typical cophasing process occurs as follows:

1. Set \( \alpha \) such that \( \sigma_{\text{pupil}}(\alpha \times [p_0, t_0, T_0]) \) is equal to the initial wavefront error we want to start with.
2. Compute a ZELDA image using the MFT.
3. Calculate \( \varphi_0, \varphi_1, \varphi_2 \) and convert to piston, tip, and tilt using the calibration matrix.
4. Calculate the residual piston, tip, and tilt expected after moving the segments.
5. Repeat steps 2–4 until convergence is reached or for a given number of iterations.
In this section, we investigate the probabilistic behavior of the phasing process with ZELDA. The objective is to propose from a statistical point of view a predictive model of convergence as a function of the initial aberration error value. In the following, the starting wavefront errors correspond to a fraction of the capture range as already explained in Sect. 4.1. Piston error only (1) and piston combined with tip-tilt (2) are addressed separately for the sake of clarity.

In case (1), both empirical and semi-analytical approaches are carried out. The empirical method is performed by considering the outcome of 100 samples of phasing attempt processes. The approximated probability of convergence for a given variance is then

$$\mathbb{P}_E(\sigma_{\text{pupil}}^2) = n_{\text{conv}} / 100,$$

where \(n_{\text{conv}}\) is the number of positive outcomes, referring to the situation for which the phasing process achieves a phased and stable mirror state, that is, the phasing process converges. Alternatively, negative outcomes correspond to the case of the diverging phasing process. Figure 5 (left) presents the occurrences for converging cases (blue bars) and diverging cases (red bars). We note that \(\mathbb{P}_E = 0.5\) for \(\sigma_{\text{pupil}} \approx 65\) nm RMS.

The semi-analytical approach is based on a 10 000-sample Monte Carlo integration used to infer the probability \(\mathbb{P}_{\text{SA}}\) for each segment to be within the given range \([a, b]\) around its mean position. This probability is expressed as

$$\mathbb{P}_{\text{SA}}(\sigma_{\text{pupil}}^2, a, b) = \int_{\mathbb{R}^N} \prod_{n=1}^{N} \mathbb{I}\left(a < p_n - \mathbb{E}[p] < b\right) \times \prod_{n=1}^{N} f_{p}(p_n \mid 0, \sigma_{p}^2)(dp_1 \cdots dp_n),$$

where \(\mathbb{I}\) is the indicator function. Since the introduced pistons are randomly issued from a zero-mean normal distribution, \(f_{p}(p \mid 0, \sigma_{p}^2)\) is the probability density of the Gaussian distribution \(N(0, \sigma_{p}^2)\). Based on Fig. 2, we set \(a\) and \(b\) to \(-\lambda/4\) and \(3\lambda/4\) respectively, that strictly delimits the domain outside which the system corrects for in the wrong direction. The result of the Monte Carlo integration corresponds to the yellow curve presented on Fig. 5 (left), and it fits well with the data from the empirical method. This provides confidence in the initial assumption that a system converges when all the individual pistons over the pupil are within the range \([\varphi_0, \varphi_1]\) around its mean.

In case (2), two complementary approaches are also used. The same empirical method as previously defined is performed by producing 100 samples of system iterations and evaluating...
the outcomes to derive $\mathcal{P}_E$. Figure 5 right presents the results, where the blue bars indicate the converging cases and the red one the diverging cases.

The second method is different from the semi-analytical approach that has been previously used for the case of piston only. With the correlation between piston, tip, and tilt estimators, elementary rules of thumb on the initial conditions cannot be settled to frame the convergence outcomes. To overcome this issue, we use the theoretical expressions of the estimators shown in Eqs. (9)–(11) to emulate the converging process where only the diagonal terms of the calibration matrix are kept. The results are represented by the yellow curve in Fig. 5 right and match the previous data. A slight offset, however, indicates that the method tends to underestimate the number of converging systems. This effect is also distinguishable on the piston-only configuration. This underestimation reflects the shortcoming of a binary condition to predict the outcome of the phasing process. In an actual system, while few segments might initially be out of the capture range, and since the estimators are correlated, it might occur against the odds that the amplitude of misalignment of these segments meets the boundaries of the capture range, and finally succeeds in converging to zero error residuals. Finally, by analogy with case (1), a phasing system with a symmetric capture range will converge for larger initial aberrations (Fig. 5, right, purple curve).

The models defined in this section allow one to infer the probability of convergence for a given initial wavefront error. Although the presented results have been obtained with a pupil made of 91 segments, the models can account for different configurations. The limit between the cophasing regimes can be clearly set based on a given probability threshold. Fixing this probability to $P = 95\%$, one can conclude that for an asymmetric sensor the limit between fine and coarse phasing is 120 nm, while for a symmetric sensor this limit is 250 nm for piston and tip-tilt cases. These algorithms can be used for preliminary studies of a system to conjecture the needs in terms of cophasing accuracy and capture range.

4.3. Sensitivity to misalignment

The sensitivity of the cophasing system regarding hardware misalignment is studied by investigating both the FPM and the pupil misalignment. For the sake of similarity with the JWST, we consider a pupil made of 19 segments. A specific vector of initial aberrations $[p_1, t_1, T_1]$ is created (see Sect. 4.1). Because the phasing process is subject to statistically driven behaviors, these aspects are addressed in Sect. 4.3.3.

4.3.1. Focal plane mask misalignment

To determine the effect of the FPM misalignment, we evaluate the convergence of the system and record the residual wavefront error at each step for several FPM displacements. This process is repeated for various sets of aberration over the pupil. Figure 6 left presents the results for a pupil exhibiting an initial aberration of 100 nm RMS. This value is chosen for the system to have a probability to converge $P \approx 1$. We note that the residual global tip-tilt over the pupil, originating from the system compensation for the FPM misalignment, is numerically removed from the data (the telescope active optics correct for the global tip-tilt and defocus). Figure 6 left shows that the system can accommodate a FPM misalignment up to 0.7 $\lambda/D$. Since levels of mask positioning can be guaranteed to a level of $10^{-2}$ to $10^{-3}$ $\lambda/D$ using a low-order wavefront sensor (e.g., Mas et al. 2012), the sensitivity of our closed-loop system to FPM misalignment is negligible.

4.3.2. Pupil misalignment

In real observing conditions, the position of the pupil may be shifted by a significant fraction of its diameter ($3\%$–$4\%$ for the JWST, Bos et al. 2011). This can possibly affect the phasing process as the calibration matrix is built on a perfectly centered system. We study this effect by numerically shifting the analysis zone $\mathcal{H}$ and processing the closed-loop operation. Figure 6 right presents the results of the convergence for several displacements. As for the FPM misalignment analysis, the initial wavefront error is about 100 nm RMS (the global tip-tilt is also removed from the data). For displacements as large as $4\%$ of the pupil diameter, corresponding in our simulations to $17.5\%$ of the segment diameter, the system converges. This indicates a low sensitivity of the sensor to the pupil shear. We propose a detailed statistical study of these aspects in the following.
4.3.3. Statistical behavior of the misalignment

We estimate the limit in FPM and pupil displacements which the system can accommodate for different initial RMS wavefront error. Figure 7 presents the results for the FPM (blue points) and the pupil (red points) displacements. The mean value has been computed for 100 realisations. The error bars represent the 1σ standard deviation. As expected for both FPM and pupil analysis, the larger the initial wavefront error the lower the tolerance. Nevertheless, a difference in the standard deviation of the FPM and pupil results is observable. While the standard deviation is roughly constant with the FPM displacement, it increases significantly with the pupil misalignment. Pupil and FPM displacements therefore impact on the system differently.

Analytical development of the estimators shows that the pupil displacement affects the measurement on each segment by introducing a systematic error proportional to the displacement. All the segments are impacted alike and have the same statistical weight when considering their contribution to initiate the phasing process divergence. The effect is different when considering the contribution of a FPM displacement. Numerical simulations show that the system converges towards a configuration with a global tip-tilt over the pupil (compensating for the FPM displacement), meaning that the final position of each segment should converge to

\[ [p^r_n, t_n, T_n] = [p^r_n, r^*, T^*]. \]  

where \( r^* \) and \( T^* \) stand for the global tip and tilt respectively, and \( p^r_n \) the residual piston associated to the segment \( n \) and subject to \( r^* \) and \( T^* \). The expression for the piston on the \( n \)th segment is

\[ p^r_n = \frac{r}{2} \left[ 3k_n r + \sqrt{3}l_n T \right] \quad \forall (k_n, l_n) \in \mathbb{Z}, \]  

where \( r \) denotes the segment radius; \( k_n \) and \( l_n \) are specific to each segment and their values depend on the individual segment position in the pupil. The FPM displacement contribution affects the piston distribution non-uniformly in the pupil, and thus alters the estimators differently on each segment. Outer ring segments are more impacted and are thus predisposed to initiate the phasing process divergence well before the inner ring ones.

To summarize, while the contribution of the pupil displacement is local, the FPM displacement contribution is global. Thus the number of segments capable of initiating the process of divergence is different. While for a pupil displacement borderline segments are equiprobable, for a FPM displacement outer ring segments are overweighted. This can explain the widening of the standard deviation observed in Fig. 7.

4.4. Stellar magnitude sensitivity

As for hardware misalignment, the sensitivity to stellar magnitude is critical to evaluate the capacity of the sensor to operate under realistic conditions. The analysis is performed assuming an 8 m telescope observing in \( V \)-band (spectral bandwidth \( \Delta \lambda = 90 \) nm) for 30 s integration time. The pupil is composed of 91 segments. In these simulations, we assume photon noise, readout noise (1 e\(^-\) RMS), and dark current (1 e\(^-\)/s). We use the residual RMS wavefront error after 20 iterations as a metric to determine the quality of the convergence. To smooth out the random behavior of the process, the median of 50 different sets of aberrations is considered. Figure 8 presents the results and as expected, the sensitivity to stellar magnitude depends on the initial RMS wavefront error. Simulations show that ZELDA has a sensitivity to stellar magnitude that is consistent with other state-of-the-art phasing sensors (e.g., Pinna et al. 2008; Gonté et al. 2009; Surdej et al. 2010). Assuming a \( V \) magnitude of 15, ZELDA provides a residual wavefront error smaller than 5 nm RMS for an initial wavefront error as large as 140 nm RMS.

5. Conclusion

In this paper, we demonstrate the suitability of ZELDA, hitherto proposed for non-common path aberrations, for the precise measurement of piston, tip, and tilt errors in segmented
apertures within the diffraction-limited domain. Measurements of the segment alignment errors do not require changing the typical ZELDA set-up, but the signal estimators. ZELDA arises as a multitask sensor because it can simultaneously measure cophasing aberrations, segment figuring errors, and continuous wavefront aberrations due to optics misalignment (N'Diaye et al. 2016).

Our closed-loop simulations show that within its capture range the sensor drives the segment piston, tip, and tilt to sub-nanometric residuals in a few iterations. However, its asymmetric capture range limits the convergence dynamic. Nonetheless, it can be used as a stand-alone sensor for wavefront errors up to 120 nm RMS with a 95% chance of phasing the system. Improvements on the capture range are foreseen using either multiwavelength or coherence-based methods (see Martinez & Janin-Potiron 2016, for a review of these techniques), or by implementing innovative FPM designs (Jackson et al. 2016).

We study the phasing process from a statistical point of view. To our knowledge, the probabilistic behavior of the phasing process has not previously been explored in a systematic way. The proposed probabilistic models are independent of the phasing sensor architecture and only rely on the boundaries of the sensor capture range. These models provide the opportunity to predict the success of the phasing process as a function of the initial capture range limits the convergence dynamic. Nonetheless, higher order modes will introduce additional cross-terms into the system, possibly reducing the capture range to some extent. These effects are currently under study (e.g., Vigan et al. 2016).

Finally, the concept proposed in this paper will be tested in laboratory in the framework of the SPEED project (Martinez et al. 2016), and compared with other cophasing systems (Janin-Potiron et al. 2016; Martinache 2013).

Acknowledgements. P. Janin-Potiron is grateful to Airbus Defense and Space (Toulouse, France) and the Région PACA (Provence Alpes Côte d’Azur, France) for supporting his PhD fellowship. P.J.P would like to thank Y. Potiron for sharing his mathematical wisdom. M. N'Diaye acknowledges support from the European Research Council (ERC) through the KERNEL project grant #683029 (PI: F. Martinache) and would like to thank F. Martinache for his support.

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Appendix A: Analytical expression of the standard deviation over a segmented aperture

In this Appendix, we detail the calculation of the analytical expression for the standard deviation of wavefront errors in the context of hexagonal segmented apertures.

Let \( h_n(x, y) \) be the elevation on the segment \( n \) (with \( n \in [1, N] \)) at the coordinates \((x, y)\),

\[
h_n(x, y) = p_n + t_n x + T_n y, \quad \forall (x, y) \in S,
\]

where \( S \) is the surface defined by a single segment as shown in Fig. A.1. The expression of \( S \) is

\[
S = \{(x, y) \in \mathbb{R}^2 \mid x \in [-r, r], y \in [\zeta_1(x), \zeta_2(x)]\},
\]

with

\[
\zeta_1(x) = \left\{\begin{array}{ll}
-\sqrt{3}(x + r) & \text{for } -r < x < -\frac{r}{2} \\
-r_b & \text{for } -\frac{r}{2} \leq x \leq \frac{r}{2} \\
\sqrt{3}(x - r) & \text{for } \frac{r}{2} < x < r
\end{array}\right.
\]

and

\[
\zeta_2(x) = \left\{\begin{array}{ll}
\sqrt{3}(x + r) & \text{for } -r < x < -\frac{r}{2} \\
r_b & \text{for } -\frac{r}{2} \leq x \leq \frac{r}{2} \\
-\sqrt{3}(x - r) & \text{for } \frac{r}{2} < x < r.
\end{array}\right.
\]

Hereafter, the notation \( \bar{\ldots} \) is used to represent the averaged value over one single segment, while \( \mathbb{E}[\ldots] \) is used to represent the average over the pupil or the average of the piston, tip, and tilt vectors. The mean elevation on the \( n \)-th segment is

\[
\bar{h}_n(x, y) = \frac{1}{A} \int_S h_n(x, y) \, dx \, dy,
\]

\[
= \frac{1}{A} \left( p_n \int_S x \, dx \, dy + t_n \int_S y \, dx \, dy + T_n \int_S y \, dx \, dy \right),
\]

where \( A = 3 \sqrt{3}r^2/2 \) represents the segment area. It leads to

\[
\bar{h}_n(x, y) = p_n.
\]

The mean squared elevation on the \( n \)-th segment is given by

\[
\bar{h}_n^2(x, y) = \frac{1}{A} \int_S h_n^2(x, y) \, dx \, dy,
\]

\[
= \frac{1}{A} \left( \int_S (p_n^2 + t_n^2 x^2 + T_n^2 y^2 + 2p_n t_n x + 2p_n T_n y + 2t_n T_n x y) \, dx \, dy \right).
\]

Developing the previous equation, we obtain

\[
\bar{h}_n^2(x, y) = \frac{p_n^2}{2} + \frac{5}{24} r^2 t_n^2 + \frac{5}{24} r^2 T_n^2.
\]

Let also \( H \) be the union of all the segments composing the pupil:

\[
H = \bigcup_{n=1}^{N} h_n.
\]

The mean value of \( H \) is

\[
\mathbb{E}[H] = \frac{1}{N} \sum_{n=1}^{N} h_n = \mathbb{E}[p_n].
\]

The mean value of \( H^2 \) is

\[
\mathbb{E}[H^2] = \frac{1}{N} \sum_{n=1}^{N} h_n^2
= \mathbb{E}[p_n^2] + \frac{5}{24} r^2 \left( \mathbb{E}[t_n^2] + \mathbb{E}[T_n^2] \right).
\]

The variance over the full pupil is then expressed as

\[
\sigma_{pupl}^2 = \mathbb{E}\left[ (H - \mathbb{E}[H])^2 \right] = \mathbb{E}[H^2] - \mathbb{E}[H]^2,
\]

\[
= \mathbb{E}[p_n^2] - \mathbb{E}[p_n] + \frac{5}{24} r^2 \left( \mathbb{E}[t_n^2] + \mathbb{E}[T_n^2] \right).
\]

Finally, using the standard deviation of the piston, tip, and tilt samples, respectively \( s_p, s_t, s_T \), we obtain

\[
\sigma_{pupl} = \sqrt{s_p^2 + \frac{5}{24} r^2 \left( s_t^2 + s_T^2 + \mathbb{E}[t_n^2] + \mathbb{E}[T_n^2] \right)}.
\]

Appendix B: Impact of the ZELDA asymmetric capture range

In this appendix, we propose to assess the impact of the asymmetry in the ZELDA capture range on the closed-loop convergence. The SCC-PS (Janin-Potiron et al. 2016) has a symmetric piston capture range and is used as a reference for comparison purpose. For both sensors, we use the same configuration and the same initial conditions as stated in Sect. 4.1.

In this piston-only configuration, the limit of convergence is reached for different initial wavefront errors: below 70 nm RMS with ZELDA (Fig. B.1, top left), ~110 nm RMS with a symmetric capture range sensor (Fig. B.1, bottom left). This reduction with ZELDA is a direct effect of the piston capture range shifting as described in Sect. 2.2. For a larger amount of aberrations,
the system converges toward a stable state with one (or more) segment(s) phased at a multiple of the wavelength.

In the piston plus tip-tilt configuration, the limit of convergence is also reached for different initial wavefront errors: below $\sim 140$ nm RMS for ZELDA (Fig. B.1, top right), $\sim 200$ nm RMS with a symmetric capture range sensor (Fig. B.1, bottom right). For the symmetric capture range sensor (Fig. B.1, top right, green curve), when the initial wavefront errors are too large, the system converges toward a stable non-phased configuration, while it diverges for ZELDA. This effect is intrinsic to the SCC-PS used here as a reference cophasing sensor, and will therefore not be discussed further.