## CHEOPS PIPE manual

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#### 1 Introduction

PIPE (PSF Imagette Photometric Extraction) is developed to extract photometry from CHEOPS (Benz et al. 2021) data using PSF (point-spread function) photometry. CHEOPS data typically consists of a cube of 200 pixel diameter subarray frames that come from a region of the full  $1024 \times 1024$  pixel detector. For bright sources  $(G < 8 \,\mathrm{mag})$ , each subarray frame may be co-added from several exposures to save data rate on the downlink. In those cases, also smaller imagettes frames of 60 pixel diameter are downloaded that are not co-added, or at least less co-added for the brightest stars. PIPE was originally developed to extract photometry from the imagettes, where PSF extraction is particularly useful since the background varying in time, can be fitted simultaneously with the PSF. The small size of the imagettes otherwise makes it difficult to estimate the background with aperture photometry. Advantages in using the imagettes for photometry is the higher time resolution from shorter cadence, but also the reduced sensitivity to bad pixels, cosmic rays, and satellite transits.

#### 1.1 When to use PIPE

The first question to ask is Do I really need PIPE? All CHEOPS data sets are processed by the CHEOPS Data Reduction Pipeline (DRP, Hoyer et al. 2020), which uses aperture photometry on the subarrays to provide extracted photometry. The DRP products are robust and optimal for intermediately bright stars (G = 8 - 11 mag) without complicated background and no imagettes available; in those cases there is usually no benefit to using PIPE, so the simplest approach is to use the extracted photometry already provided by the DRP. There are however some circumstances when PIPE can improve the quality of the extracted photometry:

- For bright stars ( $G < 8 \,\mathrm{mag}$ ) where imagettes exist and when higher time resolution is desired, e.g. to resolve stellar activity or improve the time constraint on transits. There is usually no noise penalty in doing photometry on imagettes, binned to the same time resolution as the subarray photometry, imagette photometry is as good or better (due to more efficient filtering out of satellite transits and cosmic rays).
- Whenever there are nearby background stars that are suspected to contaminate the photometric aperture; PSF photometry disentangles the contribution from the background stars, significantly reducing the roll modulation.
- For faint stars (G > 11 mag) where hot and telegraphic pixels (pixels with a strong variable dark current) contribute significantly to the noise. Since PSF photometry can effectively mask any contributions from bad pixels, this reduces their impact on the photometry and improves the quality of extraction. As CHEOPS ages, the number of bad pixels increases with time, moving the limit where PSF photometry becomes beneficial to increasingly brighter targets.
- Whenever the DRP-extracted light curve shows unexpected behaviour, the independent PIPE extraction
  can be used for verification. In particular, PIPE has a different sensitivity to background stars contaminating the DRP aperture.

Usually there is no penalty for using PIPE, except for the extra time it may take to repeat the reduction and extraction.

## 2 Installation

The easiest way to install PIPE is using pip:

```
pip install pipe-cheops
```

This should install PIPE and all its dependencies and make them available from your python installation. Alternatively, the latest development version can be downloaded from GitHub:

```
git clone https://github.com/alphapsa/PIPE.git
cd PIPE
python setup.py install
```

**Note:** To use PIPE after installation, you also need to set the paths to data and calibration files, as well as download the proper calibration files. See subsections below.

#### 2.1 Dependencies

PIPE is developed for python 3 and has been tested for python 3.7 and later, but earlier versions might also work. PIPE makes use of, and is thus dependent on, the following four external modules: NumPy, SciPy, Astropy, and scikit-image (an image processing package). You probably already have the first three installed. If anything is missing, pip should automatically install them for you as part of the installation process of PIPE. If you install directly from the GitHub repository on the other hand, you will need to ensure that the required modules are installed and available on your system. In general we recommend Anaconda as a good way of installing and maintaining most of your scientific python needs.

Bug note: Astropy up to 5.1 has a known incompatibility with NumPy 1.24 and later. Best update to Astropy 5.2 or later, in which case there is no problem. Should you be unable to upgrade, a workaround is to set the PIPE parameter mjd2bjd = False (see §A.1); this bypasses the MJD to BJD computation where the problem appears but produces tables where no barycentric correction has been made.

#### 2.2 Setting paths

After calling PIPE for the first time after installation, e.g. by executing

```
import pipe
```

from the python prompt, it will ask for root paths to data files (data\_root) and reference/calibration files directories (ref\_lib\_data). These are the locations where CHEOPS data are expected to be found in subdirectories, and where all calibration files are expected to be found, respectively. If the paths are not explicitly set, the default paths cheops-pipe directory inside the home directory will be created and used. It is possible to change the paths afterwards by executing

```
from pipe.config import get_conf_paths
get_conf_paths(overwrite=True)
```

#### 2.3 Calibration files

In the ref\_lib\_data directory, there are a number of calibration files expected in order for PIPE to be able to reduce data from CHEOPS and extract photometry. Most files are optional in that they will not be required if their use is switched off by setting the corresponding PIPE parameter to False; the exception is the gain parameter that has to be set if not calculated, and the the PSF library that always needs to be present.

Table 1 lists the various calibration files used by PIPE. The non-linearity definition file and the PSF library are for now separately provided from a dropbox; the rest can be downloaded from the CHEOPS archive. Use the *Reference Data Query* tab and select the desired files to query for from the drop list. The bad pixel maps and dark frames are updated on a monthly bases, so you will want to add calibration files that are close in time to the observations to be reduced.

# 3 Using PIPE

## 3.1 Data files

Best practice is to download all files relevant for a visit and put it in a subdirectory. PIPE assumes all data is under the previously configured data path data\_root organised in subdirectories according to target and visit as data\_root/target/visit, where target and visit are arbitrary strings and the visit directory contains all data from the visit as downloaded from DACE or the CHEOPS archive. From DACE, choose to download "All data products", where each visit is unpacked to the target directory and optionally rename the visit directory. Instead of using the name from the full file key, e.g. PR100006\_TG00313\_V0300, it can be renamed to 313 so that the full data path becomes data\_root/55Cnc/313/ (naming the target 55Cnc, but could be named anything). Data can also be downloaded from the CHEOPS archive, though care has to be taken that all required files listed in Table 2 are downloaded.

Table 1: Calibration files used by PIPE.

| Calibration file       | Function  |
|------------------------|---|
| REF_APP_GainCorrection | File with gain correction parameters, expected directly in the  |
|                        | ref_lib_data path. Can be downloaded from the CHEOPS  |
|                        | archive. Optional, overridden if gain PIPE parameter is set (to   |
|                        | e.g. $1.95 e^{-}/ADU$ ); see §A.2.  |
| nonlin.npy             | Non-linearity definition file, expected directly in the ref_lib_data  |
|                        | path. Optional, can be ignored (with no non-linear correction) if   |
|                        | the PIPE parameter non_lin = False. See §A.2.   |
| REF_APP_FlatFieldTeff  | Contains the pre-launch determined flat field as a function of effec-   |
|                        | tive temperature of the spectral energy distribution of the target.   |
|                        | Huge file (~1 GB) that can be downloaded from the CHEOPS  |
|                        | archive. Optional, disabled if PIPE parameter flatfield =   |
|                        | False.  |
| psf_lib/               | The PSF library containing sub-directories with PSF model pa-   |
|                        | rameters. The library must be re-populated if the target is re-   |
|                        | located to a new position on the CHEOPS detector, see §5.   |
| BadPixels/             | Directory containing bad pixel maps. These can be downloaded  |
|                        | from the CHEOPS archive as REF_APP_BadPixelMap and put into   |
|                        | this directory. Choose files that are from near the date of the tar-  |
|                        | get observation. When PIPE runs it selects the bad pixel map file   |
|                        | nearest the date of the observation. Taking bad pixels into account can be switched off by setting the PIPE parameter mask_badpix |
|                        | = False. See §A.4.  |
| DarkFrames/            | Directory containing dark current frames. These can be down-  |
| Darki Lames/           | loaded from the CHEOPS archive as REF_APP_DarkFrame and put   |
|                        | into this directory. Choose files that are from near the date of  |
|                        | the target observation. When PIPE runs it interpolates the dark   |
|                        | frames bracketing the date of the observation (or picks the nearest   |
|                        | if not bracketed). Dark current subtraction can be switched off   |
|                        | by setting the PIPE parameter darksub = False. See §A.2.  |
|                        | , , , , , , , , , , , , , , , , , , ,   |

Table 2: Data files used by PIPE.

| Data file             | Function   |
|-----------------------|--|
| SCI_RAW_Attitude      | Stores attitude information, such as time, pointing, and roll angle. |
| SCI_RAW_HkExtended    | Contains housekeeping parameters used for gain correction. Op-       |
|                       | tional, overridden if gain PIPE parameter is set.                    |
| EXT_PRE_StarCatalogue | A catalogue of stars in the field, retrieved from the Gaia DR2.      |
|                       | Used to model the star background; optional, set PIPE parame-        |
|                       | ter bgstars = False if not available.                                |
| SCI_RAW_SubArray      | A datacube with the raw subarray data from the visit.                |
| SCI_RAW_Imagette      | A datacube with the raw imagette data from the visit; optional.      |
|                       | PIPE checks for availability and extracts only subarray photometry   |
|                       | if imagettes are not available.                                      |

Not all downloaded data files are used by PIPE, and some are optional (e.g. the star catalogue). In particular, none of the DRP generated files are used by PIPE, but can be good for reference, in particular the DRP report. Table 2 lists the data files used by PIPE.

## 3.2 Running PIPE

The PIPE workflow is controlled by a class called PipeControl that accepts a parameter object generated by PipeParam. When creating the PipeParam object, all parameters relevant to PIPE are set to default values that can then be modified before supplying it to PipeControl.

```
from pipe import PipeParam, PipeControl

# Name of target and visit. The data is assumed to be organised as
# data_root/target/visit where the visit directory contains all data
target = 'TOI-2085'
```

```
visit = '3801'
# Generate PIPE parameter file
pps = PipeParam(target, visit)
# Default parameters in pps can be modified, here are some examples:
pps.nthreads = 8  # Number of threads to use; default is
                    # number of virtual CPU cores - 1.
pps.klip = 5
                 # Number of principal components for PSF fit
                            # This enables PSF photometry of
pps.fit_bgstars = True
                            # bright stars in the field that
                            # are not the target
# Generate a PIPE control object using the parameter file
pc = PipeControl(pps)
# This is the time-consuming step where PIPE is executed
pc.process_eigen()
```

Execution time depends on many factors but a very rough estimate can be found from the following scaling law:

 $T_{\rm sa} = 10 \,\mathrm{min} \times \left(\frac{10}{n_{\rm threads}}\right) \left(\frac{n_{\rm sa}}{400}\right) \left(\frac{\kappa}{5}\right) \left(\frac{R_{\rm fit}}{40}\right)^2$  (1)

where  $n_{\rm threads}$  is the number of threads assigned,  $n_{\rm sa}$  is the number of subarray frames in the visit,  $\kappa$  is the number of principal components used for the PSF fit, and  $R_{\rm fit}$  is the fitting radius for the PSF. This scaling relation merely gives the order of magnitude as it depends on many details, including the CPU. A similar scaling relation holds for the imagettes, that typically take more time if they are significantly more numerous. If background stars are additionally to be fitted, the time scales almost linearly with the number of background stars fitted  $n_{\rm BG}$  so that  $T_{\rm sa} \propto (1 + n_{\rm BG})$ .

PIPE is not optimised for memory, as it loads all subarrays and imagettes and also maintains intermediate processing copies in memory. The memory footprint is about 5 GB for 1000 subarray frames, and about twice that if there are imagettes available. For faint stars (defined as having 60 s per exposure), this means  $\sim 5$  GB for a 10-orbit visit. For one of the absolute longest CHEOPS visits, a 4-day visit of the bright HD 172555 ( $G=4.7\,\mathrm{mag}$ ) with 4552 subarray frames and 54624 imagettes, the peak RAM usage of PIPE was observed to be 30 GB, slightly less than the given rule-of-thumb estimate.

Multiprocessing note: Because multiprocessing for python is more of an afterthought than designed into the language, some peculiarities may surface. One such is related to how sub-processes are spawned that can result in instantiation errors on some systems. This is easily avoided by including this harmless if-statement in your python script before anything else is executed:

```
if __name__ == '__main__':
and then continue with the code.
```

# 4 PIPE output files

For a given visit, the extracted photometry and other optional diagnostic files are saved in the directory data\_root/target/visit/Output/version/, where Output is a generated sub-directory if not already existing, and version is a zero-leading number of five digits. By default, PIPE checks for the lowest available version number not already existing, starting with 00000 and continuing with 00001, 00002, etc. for subsequent runs. The version of the run can also be explicitly defined by giving an argument when constructing the PipeParam object, as in pps = PipeParam('TOI-2085', '3801', version=11). Output files will then be saved in 'data\_root/TOI-2085/3801/Output/00011/'. Below follows a description of the most common output files. See §A.6 for other optional diagnostic files.

#### 4.1 Logfile

A text file logfile.txt is constructed saving all parameters for the run and being updated with processing log information as the trun progresses.

#### 4.2 Extracted photometry

The extracted photometry for subarrays and imagettes (if existing) will be saved in fits table files called target\_visit\_sa.fits and target\_visit\_im.fits, respectively. The header of the fits file inherits keywords from the corresponding CHEOPS raw data files, and the columns for both files are of the following format:

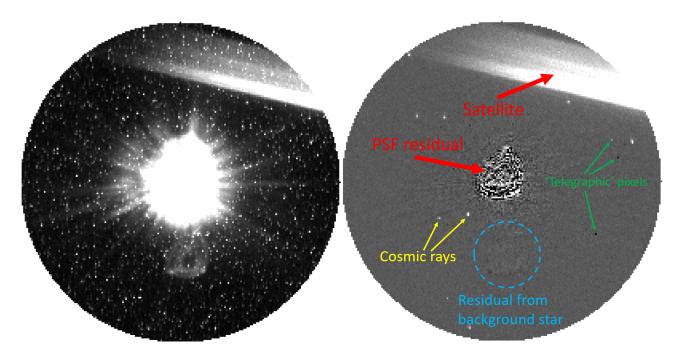


Figure 1: An observed raw subarray frame to the left, and the residuals after reduction and subtraction of target and background to the right. Artefacts in the residuals are annotated in the right image. Satellite tracks are visible in only a few % of the subarray frames. They are not modelled/removed but merely masked from the fit. Only in cases where the satellite passes very close to the target do they significantly affect the quality of the photometry. Telegraphic pixels are pixels with significant and variable dark current. They are normally masked from the PSF fit.

```
'MJD_TIME' - modified Julian Day number
'BJD_TIME' - barycentric corrected Julian Day number
'FLUX' - in electrons per (co-added) exposure
'FLUXERR' - in electrons per flux value
'BG' - electrons per (co-added) exposure and pixel
'ROLL' - roll angle of field in degrees
'XC' - X-coordinate of target relative to corner of image
'YC' Y-coordinate of target relative to corner of image
'FLAG' - integer sum of the following indicators:
  0 - no recognised issue with data point
  1 - deviating centroid (far off the median centroid)
  2 - abnormally high level of bad pixels in frame (could be satellite or CR)
       strongly deviating flux in frame (could be strong background or satellite)
    - source not found in subarray
^{\prime}UO^{\prime} to ^{\prime}U4^{\prime} are the relative weights of the first 5 PSF principal components
'thermFront_2' -
                the value of the thermFront_2 sensor, sometimes useful for de-trending
```

#### 4.3 Residuals

To see how well the PSF fit matched the data, it can be useful to look at the residuals of the fit. For each frame, the fitted PSF is subtracted from the data. If background stars are defined, then those are also subtracted. The resulting residual data cubes are saved in residuals\_sa.fits and residuals\_im.fits.

#### 4.4 List of PSF files used

PIPE normally automatically selects suitable PSFs from the PSF library that are closely matched to the target, given information about the position on the detector, the target star effective temperature, and the range of telescope tube temperatures (thermFront\_2 sensor readings) during the visit. The PSFs are then analysed using principal component analysis (PCA), forming a basis that is used to fit the target PSF in each frame. A list of the filenames of the PSFs used is saved in the text file psf\_filenames.txt. Apart from being a record, this list can be used as input to PIPE, to ensure that those specific PSFs are selected. This can be useful when e.g. comparing photometry between visits, as different PSFs might derive slightly different absolute levels of the photometry. To use this list as input for a subsequent PIPE run, define the parameter as in

```
pps.psf_filenames_file = 'fullpath/psf_filenames.txt'
```

where fullpath is indeed the full path to the location of the directory holding the file.

#### 4.5 Background star photometry

If any background star is selected for PSF photometric extraction, its photometry can be saved in the fits table bg\_star\_photometry\_sa.fits (and bg\_star\_photometry\_im.fits). The columns in the table are

```
'MJD_TIME' - modified Julian Day number

'BJD_TIME' - barycentric corrected Julian Day number

'f0' - in electrons per (co-added) exposure for first bg star

'f1' - in electrons per (co-added) exposure for second bg star

...

'fn' - in electrons per (co-added) exposure for n:th bg star
```

The flux columns are ordered in distance from the target so that the nearest extracted background star is the first column, and so on. The Gaia DR2 ID is listed as a text comment to the corresponding FITS-column. The tabulated flux for the background stars are relative to the visit-median of the target star flux.

## 5 Making custom PSF models

The existing PSF library is usually sufficient to match the target PSF well enough for good PSF photometry. How well the PSF matches can be checked in the residual output file, where systematic differences from white noise are easily seen (Fig. 2). A PSF mis-match can result in PSF noise being added to the extracted photometry, deteriorating the photometric quality. This is particularly relevant for bright sources ( $G \ll 8 \, \text{mag}$ ), where small deviations of the PSF can be the dominant source of noise. In those cases it can be beneficial to produce a PSF model from the data itself. Another instance when new PSF models are useful is if the observations have been centred on a location on the detector not previously archived in the library.

To produce high-quality PSF models, it is preferable to use visits with exposure times so short that imagettes are available. This is to reduce the motion smearing due to jitter that is inevitable during longer exposures. PIPE produces PDF models by fitting a single 2D spline to the normalised PSF of multiple frames coming from a full CHEOPS orbit. Together with the pre-processing that removes background stars this limits the PSF contamination coming from background stars, since the field makes a full rotation during one orbit. Combining data from multiple frames also reduces the effects of noise, outliers (like cosmic rays), and bad pixels (thanks to jitter). To avoid biasing the PSFs it is best to not remove the static image, i.e. set remove\_static = False (see §A.4). PSF library producing code can look something like:

```
from pipe import PipeParam, PipeControl

# Define run parameters
pps = PipeParam(name='TOI-2085', visit='3801')
pps.remove_static = False

# Process the data before extracting PSF models
pc = PipeControl(pps)
pc.process_eigen()
pc.make_psf_lib()
```

Running make\_psf\_lib is computational intensive. It uses one thread per orbit of the visit (up to the number of available virtual cores), and each thread takes 20-60 min to complete, depending on the number of subarrays/imagettes per orbit and on the CPU. A resulting PSF model for each orbit is put into the ref\_lib\_data/psf\_lib/XXXxYYY/ directory with the file format psf\_Teff\_TF2\_MJD\_Texp\_serial.npy, where XXX and YYY give the location on the detector of the PSF in integer pixel coordinates, Teff is the effective temperature of the star used, TF2 is the median (absolute) value of the thermFront\_2 sensor during the orbit, MJD is the modified Julian date of the observation, Texp is the exposure time, and serial is just a serial number counting up in case there is already a file of the same name. An example of a PSF model filename that illustrates the format in more detail is

```
'ref_lib_data/psf_lib/291x830/psf_05920K_12.20C_59821_11.7_0000.npy'
```

meaning that this is a PSF model determined for a visit at MJD = 59821 with a  $T_{\rm eff} = 5920\,\rm K$  star when thermFront\_2=  $-12.20^{\circ}\,\rm C$  and at exposure times of 11.7 s.

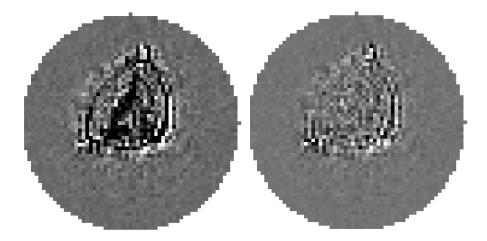


Figure 2: Residuals from the same imagette frame with different PSF models. To the left a poor match with a lot of systematic structure left in the residuals; to the right a good match with mostly white noise. A fits-cube with residual images is saved in the output directory by default.

# 6 Various tips

Accurate PSF photometry requires accurate PSF modelling. PIPE searches its PSF library for the best PSF matches according to criteria such as detector coordinate of the target, target radiation temperature, and CHEOPS telescope temperature (thermFront\_2 sensor values). The set of chosen PSFs are then analysed for their principal components (PCA) that serve as a basis for matching the PSFs of individual frames.

One trade-off is how many bases to use when matching PSFs. More bases can model more complex PSF behaviour but also become more sensitive to noise. Usually it is therefore better to have few bases for exposures of lower signal to noise, and more bases to model the PSFs of really bright stars. The number of bases are set by the klip value. Typically, klip = 1 is optimal for anything requiring 60 s exposure, while for the brightest and shortest exposures up to klip = 10 can give better results. Sometimes it is not easy to guess in advance what gives the best results, and in those cases it can make sense to empirically try out different approaches. This can be done using the "optimise" option in PIPE, see §A.5.

How well the model PSF matches the data can be best seen in the residual data cubes that are provided in the output directory by default. This residual data cube shows the data subtracted by the model for each frame, organised in a data cube (Fig. 1). A successful match leaves only white noise without systematic structures. The presence of systematic patterns are an indication of sub-optimal PSF fits (Fig. 2). The PSF match can sometimes be improved by increasing the number of bases used, or change the criteria for selecting model PSFs from the library (e.g. select more or fewer for the PCA).

The PSF tends to evolve during a visit, and this evolution is often well captured by the coefficients for the principal components. These coefficients are saved in the fits-table together with the extracted photometry (§4.2) and can be used to de-trend the light curve from e.g. a ramp.

Another important extraction parameter is the size of the region to be used for fitting the PSF. The PSF fit is noise-weighted, so in theory a larger region should always be better. In practice, larger regions exposes the fit to systematic deviations such as contaminating stars not perfectly taken into account, flat field effects and a less well-determined PSF in the far wings. Useful fitting regions typically range from radius  $R=25\,\mathrm{pix}$  for the smallest to  $R=60\,\mathrm{pix}$  for the largest.

If a contaminant star is significantly affecting the photometry because it is very bright or variable, an option is to attempt to fit it in the data (rather than just compute its assumed brightness and position from Gaia DR2). This can be sueful to exclude that the contaminant is an eclipsing binary. For background stars at very small separations (5–30 pix), fitting the star can introduce noise into the extracted photometry of the target since the fits of the target and the background star are not entirely independent but can interfere with each other.

When the background is strong, CHEOPS photometry shows a significant non-linearity that is difficult to get rid of during the data reduction. In general it is therefore recommended to mask any exposure with a background of more than  $300\,\mathrm{e^-\,pix^{-1}}$ , per exposure. If a subarray consists of e.g. 10 co-added exposures, this means that the recommended limit is to ignore any exposure with  $> 3000\,\mathrm{e^-\,pix^{-1}}$  background. One can also attempt to correct for the background non-linearity by calibrating for it, i.e. de-trend the flux against background. This requires a long visit (ideally > 5 orbits) with likely limited photometric quality for those data points, but can be warranted if the data points are important enough (e.g. right at the ingress/egress for a time-sensitive measurement),

## References

Benz, W., Broeg, C., Fortier, A., et al. 2021, Experimental Astronomy, 51, 109

Fortier, A., Simon, A. E., Broeg, C., et al. 2024, A&A, subm.

Hoyer, S., Guterman, P., Demangeon, O., et al. 2020, A&A, 635, A24

# **Appendices**

# A PIPE parameters

Here are listed the parameters most relevant for extraction; for a complete listing, see pipe\_param.py of the source code. In general, the prefix sa is for subarray and im for imagettes. The default value for each parameter is indicated after the '=' sign. When generating a PipeParam object, all parameters except name and visit are given default values that can subsequently be changed (as shown in the example of §3.2).

#### A.1 General

Table 3: General PIPE parameters.

| PIPE parameter     | Function  |
|--------------------|---|
| datapath           | The default location for the data is defined by the configured data_root        |
|                    | path, but this can be re-defined by this parameter. The directory with          |
|                    | data to be extracted is then expected to be in datapath/target/name.            |
| calibpath          | just like for the datapath, the default location for the calibration data       |
|                    | is ref_lib_data but can be changed with this calibpath parameter.               |
| nthreads           | PIPE is parallelised to make more efficient use of modern CPUs. This            |
|                    | parameter defines the number of threads to be used by PIPE. Not all             |
|                    | steps uses multiple threads, only the most computationally heavy work           |
|                    | loads do. The division of work into threads is usually trivial by giving        |
|                    | the threads one frame each to work on at the time. This also means that         |
|                    | the number of frames sets the limit to how many threads can efficiently         |
|                    | be used, but this is usually > 100 for even the shortest visits so no           |
|                    | practical limitation. The default value for nthreads is the number of           |
|                    | virtual CPU cores on the host machine, minus one. For CPUs with                 |
|                    | hyperthreading the number of virtual cores is usually twice the number          |
|                    | of physical cores, typically ranging from 8 to 64 virtual cores for modern      |
|                    | desktops. Workstations and servers can have many more cores than                |
|                    | that. The speedup with multiprocessing is significant, close to linear in       |
|                    | the number of threads.  |
| $sa\_range = None$ | Some visits are very long, and then it can be useful to only work on            |
|                    | a subset of the timeline to e.g. try out different extraction parameters.       |
|                    | sa_range is a tuple of two integers describing the subarray indices range       |
|                    | to work on. E.g., if there are 7092 subrray frames in the data set, setting     |
|                    | sa_range = (200,400) limits the extraction to frames between indices            |
|                    | 200 and 400, speeding up the extraction by $35\times$ . If there are imagettes, |
|                    | the imagettes corresponding to the selected subarrays are used. Default         |
|                    | is sa_range = None, in which case the full range is used.                       |
| Teff = None        | Normally the effective temperature from the star is retrieved from the          |
|                    | provided data files, but this parameter can be used to override it or           |
|                    | simply define it if not available in the data.                                  |
| mjd2bjd = True     | This switch encodes whether or not to make a barycentric correction             |
|                    | when computing BJD from MJD. A reason to not do it is to work around            |
|                    | an incompatibility between Astropy 5.1 or earlier and NumPy 1.24 or             |
|                    | later. Normally this parameter should be kept at its default True value.        |

# A.2 Data reduction

Table 4: Data reduction parameters.

| PIPE parameter       | Function   |
|----------------------|--|
| flatfield = True     | Switch set to True if the data are to be corrected for the flat field. Since             |
|                      | the flat field is wavelength-dependent, it is weighted by the assumed                    |
|                      | spectral energy distribution of the target. This flat field was determined               |
|                      | before launch, so it is not clear how valid it still is. If set to False, no             |
|                      | flat field file is required for processing. Also see §2.3.                               |
| darksub = True       | Switch set to True if the data are to be corrected for dark current.                     |
|                      | The dark frame is searched for in the calibration path, and the dark                     |
|                      | is interpolated over the nearest two dark frames bracketing the visit in                 |
|                      | time. If there is no dark frame available after the visit, then the latest               |
|                      | dark frame is used (with no extrapolation). Also see §2.3.                               |
| dark_min_snr = 15    | The dark current for various pixels is of variable quality. To ensure                    |
|                      | that dark subtraction does not add too much noise to image, only pixels                  |
|                      | with dark current determined to a signal-to-noise of this parameter is                   |
|                      | included.  |
| dark_min_level = 3.0 | Only pixels with dark current of at least this level (in e <sup>-</sup> per second) are  |
|                      | to be used for the dark subtraction. This is to avoid the low-level influ-               |
|                      | ence from residual stars that can affect the dark current determination,                 |
|                      | and to avoid adding noise.   |
| mask_bad_dark = True | If True, identify pixels with bad dark correction and mask them.                         |
| gain = None          | Gain is normally computed from the house-keeping data, but can be                        |
|                      | overridden if defined here (in e <sup>-</sup> per ADU) to e.g. 1.95 e <sup>-</sup> /ADU. |
| non_lin = True       | If True, apply a non-linear correction to the data.                                      |
| cti_corr = True      | As the detector ages, radiation damage produces charge traps that causes                 |
|                      | a charge transfer inefficiency (CTI). The CTI effect thus increases with                 |
|                      | time. With this parameter set to True, a simplified CTI model (devised                   |
|                      | by Göran Olofsson) is applied to correct the data. See Fortier et al.                    |
|                      | (2024).  |
| cti_t0 = 58800.0     | This is the zero epoch (in MJD) for start of CTI deterioration.                          |
| cti_scale = 0.0016   | This is a CTI scaling factor.  |
| cti_expo = -0.65     | The CTI exponent.  |
| cti_lim = 0.0333     | An empiric CTI limit.  |

# A.3 Background correction

 ${\bf Table~5:~Background~correction~parameters.}$ 

| PIPE parameter | Function  |
|----------------|---|
| bg_fit = 0     | This integer parameter decides if and how the diffuse background (due           |
|                | to e.g. the zodiac or scattered light from the Earth limb) should be esti-      |
|                | mated. $-1$ means that the background will not be fitted simultaneously         |
|                | with the PSF (there will still be background correction for the subarray        |
|                | estimated from pixels far outside the target PSF. The background for            |
|                | the imagettes will then be interpolated from the subarray background es-        |
|                | timates). <b>0</b> means that a constant will be simultaneously fitted with the |
|                | PSF and assumed to be due to the background. 1 and 2 are supposed to            |
|                | be simultaneously fitted bi-linear and parabolic background functions,          |
|                | but are not yet implemented.  |

Table 5: Background correction parameters.

| PIPE parameter                 | Function   |
|--------------------------------|--|
| bgstars = True                 | This switch indicates if background stars (retrieved from the  |
|                                | EXT_PRE_StarCatalogue file) are to be modelled and subtracted from   |
|                                | the data cube. Their model brightness use the catalogued Gaia magni-   |
|                                | tudes directly scaled to the median brightness of the target, so this does   |
|                                | not fit the brightness of the background stars; see fit_bgstars below  |
|                                | for that.  |
| limflux = 1e-5                 | Only stars bright down to this fraction of the target will be modelled for   |
|                                | the background. Particularly useful in fields densely populated by very  |
|                                | faint stars that do not make a difference but take a very long time to model. The default limflux = 1e-5 thus means that only stars brighter           |
|                                | than $10^{-5}$ of the target will be taken into account when modelling the   |
|                                | star background.   |
| star_rad_scale = 1.0           | When modelling the background stars, there is a heuristic that deter-  |
|                                | mines how far the PSF of each background star should extend, depending   |
|                                | on star brightness. The PSF of brighter stars extend much further, up  |
|                                | to 100 pixels from the background star, while fainter stars extend much  |
|                                | less, down to 25 pixels radius. The default behaviour can be tweaked by  |
|                                | changing this parameter, > 1.0 for larger PSFs, and < 1.0 for smaller  |
|                                | PSFs. This can be used if it is suspected that a nearby background star  |
| fit_bgstars = True             | contaminates the extracted photometry more than expected.  Normally the background stars are blindly modelled after the Gaia DR2                       |
| TIO-DESCRIP - ILUE             | provided by the EXT_PRE_StarCatalogue file, using the tabulated bright-  |
|                                | ness and positions relative to the target (see bgstars above). If a back-  |
|                                | ground star is significantly variable, however, or if it is bright and closely   |
|                                | separated to the target, then it can be beneficial to make a PSF fit also  |
|                                | for the background star. Default is fit_bgstars = True, i.e. some back-  |
|                                | ground stars will be fit rather than merely blindly assigned a brightness.   |
| lim_fit = 0.01                 | It makes sense to fit only the brighter background stars, as attempting  |
|                                | to fit stars hidden in the noise may introduce noise into the extraction,  |
|                                | so better to use tabulated fluxes in those cases. The lim_fit parameter  |
|                                | sets the brightness fraction limit where stars brighter than the limit   |
|                                | will be fit. Note that fitting background stars significantly prolongs the execution time, as the work required to fit each background star is similar |
|                                | to the work required to fit the target.  |
| bg_star_inrad = 5.0            | For background stars or stellar companions at very close separation to   |
|                                | the target, attempting to fit their brightness may interfere with the fit of   |
|                                | the target itself, increasing the systematic noise; in those cases it is often   |
|                                | better to just use the tabulated flux from Gaia. This parameter gives a  |
|                                | limit (in pixels $\approx$ arcsec) of how close the background star is allowed to  |
| 1.7                            | be for fitting, even if bright. Defaulted to bg_star_inrad = 5.0 pixels.   |
| blur_res = 0.5                 | Because the CHEOPS field is rotating with time due to space craft or-  |
|                                | bital roll, background stars will show rotational motion blur, with longer arcs for stars more separated form the star. To model this behaviour,       |
|                                | the PSF for the background is added along the arc at steps defined by  |
|                                | this parameter (in pixels).  |
| mask_bg_stars = False          | As an alternative to subtract the background stars they can be com-  |
| _                              | pletely masked. A reason to do this could be that they are suspected   |
|                                | variable while fitting being unpractical. Any masked pixels in a frame   |
|                                | will not contribute to the PSF extraction of the target.   |
| mask_bg_star_sep = 30          | Since masking stars too close to the target star will also mask a large  |
|                                | fraction of pixels from the target, this gives a minimum separation (in  |
|                                | pixels) for background stars to be masked.   |
| mask_bg_stars_circle<br>= True | The default is to use a circular mask, but if set to False a PSF-shaped mask is used instead, with size determined by mask_bg_level.                   |
| mask_bg_radius = 20            | This is the radius of the circular aperture (in pixels).   |
| mask_bg_raurus — 20            | This is the factor of the circular aperture (in pixels).   |

 ${\bf Table~5:~Background~correction~parameters.}$ 

| PIPE parameter      | Function   |
|---------------------|--|
| mask_bg_level = 0.1 | If the mask is not circular but PSF shaped, then this value is used to   |
|                     | determine the mask size. Any pixel of the PSF that is brighter than this |
|                     | parameter (as a fraction of the peak brightness) is within the mask.     |

# A.4 Extraction

Table 6: Photometric extraction parameters.

| PIPE parameter        | Function   |
|-----------------------|--|
| klip = 5              | This is the number of principal components to use when fitting the tar-  |
|                       | get PSF in each frame. The more components, the more closely the   |
|                       | observed PSF can be modelled. Too many components results in over-   |
|                       | fitting, introducing more noise to the photometry. To few components   |
|                       | may also not be optimal if the observed PSF is not sufficiently accurately   |
|                       | modelled. In general, bright stars are better served with more compo-  |
|                       | nents (up to 10 for the brightest stars, $G < 7 \mathrm{mag}$ ), while faint stars   |
| -i                    | (G > 10  mag) are generally better extracted with a single component.  |
| sigma_clip = 15       | This is the factor used for clipping bad pixels in a first iteration. Since  |
|                       | the noise estimates can be uncertain and biased at this point, the clipping factor is rather high to avoid masking out signal. |
| sigma_iter = 2        | This is the number of iterations to use in the sigma-clipping loop.  |
| empiric_noise = True  | Instead of using a theoretical noise estimate based on Poisson statistics  |
| ompilionorse - lide   | of the detected signal, background and dark noise, readout noise etc.,   |
|                       | estimate the noise empirically from the statistics of the residuals of each  |
|                       | pixel. Importantly, this also takes care of noise due to PSF mismatch  |
|                       | ("PSF noise"), which is difficult to asses by other means.   |
| empiric_sigma_clip =  | This is the sigma-clipping factor when using the more accurate empiric   |
| 4                     | noise to identify and mask outliers.   |
| centfit_rad = 23      | Find target and fit centroid within this radius of the frame centre.   |
| centfit_subrad = 3    | Compute flux centroid within this radius in deconvolved image.   |
| fitrad = 30           | Fit the PSF inside this radius of the target centre. In principal the  |
|                       | photometry should not be sensitive to this parameter since the fit in  |
|                       | the PSF wings are increasingly less significant. In practice a larger fit  |
|                       | domain can be useful to better constrain the (diffuse) background. This  |
|                       | also depends on if there are complications with contaminating stars. It  |
|                       | can be worth trying out different radii (between, say, 25 and 60 pixels)   |
|                       | and inspect the results, using a metric such as the MAD. This can be   |
| normrad = 25          | automated (see §A.5).  To put PSFs derived from different stars and different circumstances                                    |
| Hormiau – 25          | onto a somewhat similar footing, they are normalised by adding up all  |
|                       | flux within the radius defined by this parameter.  |
| centre = True         | Switch to decide whether to compute the centre of the target in each   |
|                       | frame. The pointing jitter is typically less than a pixel, so an alternative   |
|                       | is to not do centring at all, if e.g. there is a problem due to a nearby   |
|                       | companion biasing the centring computation.  |
| centre_psf_filename = | Centring is done using a PSF to deconvolve the image and then finding  |
| None                  | the centre through centre-of-flux of the deconvolved image. This pa-   |
|                       | rameter defines the filename of a specific PSF to use for deconvolution.   |
|                       | Normally the PSF is automatically selected to match the star, but this   |
|                       | allows to have a fixed PSF to be used for different visits and targets, po-  |
|                       | tentially improving astrometric consistency (since the position is a weak  |
|                       | function of PSF). Defaulted to centre_psf_filename = None meaning  |
|                       | that this PSF is automatically assigned.   |

Table 6: Photometric extraction parameters.

| PIPE parameter        | Function  |
|-----------------------|---|
| source_window_radius  | The target is assumed to be within this radius from the centre of the                       |
| = 30                  | frame in order for the frame to not be flagged as source-less. This can                     |
|                       | happen if the pointing is wrong, if there is a very bright transiting satel-                |
|                       | lite, or if there is a deep occultation (in particular from solar system                    |
|                       | objects, such as Quaoar). Default is source_window_radius = 30 pix-                         |
|                       | els.  |
| centre_off_x = -0.710 | This is the typical x-coordinate offset from the exact centre of the frame,                 |
|                       | used as search window centre but also useful as a default value when not                    |
|                       | determining the centre. Default centre_off_x = -0.710 pixels.                               |
| centre_off_y = 1.055  | Same as for $x$ , but in the $y$ -direction. Default centre_off_y = 1.055                   |
|                       | pixels.   |
| mask_badpix = True    | Set to True if bad pixels, as defined in the bad pixel map and by sigma-                    |
|                       | clipping, are to be masked.   |
| mask_level = 2        | The bad pixel maps are not binary but have pixel behaviour encoded.                         |
|                       | This parameter defines what pixel behaviour should be masked out, with                      |
|                       | values to the right filter out all before: <b>0</b> no pixels are masked, <b>-2</b> filters |
|                       | only dead pixels, -1 filters also half-dead pixels, 3 filters also telegraphic              |
|                       | pixels, <b>2</b> filters also saturated pixels, <b>1</b> filters also hot non-saturated     |
|                       | pixels. Default is mask_level = 2, i.e. all pixels defined to be bad except                 |
|                       | regular hot pixels are masked.  |
| smear_corr = True     | Since there is no shutter on CHEOPS, the detector is exposed during                         |
|                       | read out. For bright stars this results in vertical streaks in the read-out                 |
|                       | direction. This switch is set to True to correct this smearing from both                    |
|                       | the target and bright background stars. The smearing is computed from                       |
|                       | model stars, where all stars in the vertical column of the detector (even                   |
|                       | outside subarray) are considered.   |
| smear_resid_sa =      | For bright stars, the smearing model may not remove the observed smear-                     |
| False                 | ing completely. This can be because the smearing is determined by the                       |
|                       | instantaneous position of the star during readout, while the subarray                       |
|                       | image shows the superposition of the PSF positions during 30–60 sec-                        |
|                       | onds. If this parameter is set to True, there is an attempt to remove                       |
|                       | the residuals by vertical median filtering. The default smear_resid_sa                      |
|                       | = False since this rarely helps the photometric precision in practice.                      |
| smear_resid_im =      | The same as previous, but for imagettes. Since each imagette typically                      |
| False                 | is of shorter duration than the subarrays, residual smearing is less of                     |
|                       | a problem. In addition, the small size makes it difficult to properly                       |
|                       | measure the vertical smearing. Thus it will almost certainly never be                       |
|                       | well advised to use residual smearing correction on imagettes.                              |
| remove_static = True  | Some dark current and flat field may not be fully corrected for, and this                   |
|                       | can show up as a static pattern in the frame residuals, i.e. when remov-                    |
|                       | ing the fitted model av the target and the background stars from the                        |
|                       | frame. By median filtering over all frames, the static residual is deter-                   |
|                       | mined. If remove_static = True, this static image is then subtracted                        |
|                       | from the data cube before extraction in a second iteration. Doing this                      |
|                       | often improves the extracted photometric precision.   |
| pos_static = False    | Since the static image correction is the median of the residual frames, it                  |
| -                     | can be both positive and negative. By setting this parameter to True,                       |
|                       | the static image is enforced to be non-negative.  |
|                       |   |

## A.5 Optimisation

Some parameters affect the quality of extraction while at the same time being difficult to constrain *a priori*. A solution is to extract photometry for a range of different parameters and pick the best. This can be time consuming, so to do this in a systematic way PIPE gives the option to optimise over a range of selected parameters that have proven to be of this nature. They are the number of principal components in fit (klip),

the radius of the fitting region (fitrad), whether or not to simultaneously fit for background (bg\_fit), correct for dark current (darksub), and remove the static image (remove\_static).

To test all possible combinations of parameters would potentially be extremely time consuming; instead a simple gradient-like search algorithm is implemented. The metric used for the quality of extraction is reduced point-to-point scatter as measured by the mean absolute deviation (MAD). In PIPE, the point-to-point MAD is used, defined as:

```
import numpy as np
def mad(x):
    """Given np array x, returns point-to-point MAD
    """
    return np.median(np.abs(np.diff(x)))
```

For normally distributed noise, MAD  $\sim 0.95\sigma$ , with  $\sigma$  the standard deviation.

The optimisation starts with extraction using the nominal values, as defined by the parameters. It then continues to vary each of the parameters one at a time, as defined by the optimisation parameters below, and picks the best value for each parameter in turn. As an example, assume that the nominal klip is 5, and we want to test the values 1, 3, 5, and 10, and similarly the nominal fitrad is 30 and we want to test the values 25, 30, 40, 50, 60. then the search will start by varying only klip keeping fitrad constant at 30, and then continue with keeping klip constant at 30 varying fitrad instead. The best value for each parameter is then saved as the new nominal to be again varied in the same way. E.g. let us assume klip = 3 and fitrad = 50 were found to be best. Then the full range of klip values will be tested for a fitrad kept contant at 50, and next klip will be kept constant at 3 for the full range of fitrad values to be tested. All evaluations are saved so that no parameter combination needs to be repeated. This procedure continues until no improvement is found or until optimise\_tree\_iter have been reached. The procedure then restarts with the best combination found so far as the new nominal. This is repeated optimise\_restarts times. If this explanation of the path followed by optimisation is hard to follow, a better way is perhaps to just try it out and see how it works in specific cases.

Some steps, like data reduction, is identical for extraction with different parameters, so they are not repeated. This results in a shorter total execution time than if PIPE would be run repeatedly from scratch with the different set of parameters.

As the parameter ranges can be different for subarrays and imagettes, each have their own set of optimisation parameters.

| PIPE parameter          | Function  |
|-------------------------|---|
| sa_optimise = False     | If set to True, this initiates a search for optimal parameters for subarray |
|                         | extraction.   |
| im_optimise = False     | If set to True, this initiates a search for optimal parameters for imagette |
|                         | extraction.   |
| optimise_tree_iter =    | The number of iterations to use in search before giving up. If no im-       |
| 5                       | provement is made by varying parameters, the search is concluded early.     |
| optimise_restarts =     | The number of search restarts made with the best parameter set so far       |
| 3                       | set as nominal value.   |
| sa_test_klips =         | klip values to be tested for subarray extraction.                           |
| [1,3,5,10]              |   |
| sa_test_fitrads =       | fitrad values to be tested for subarray extraction.                         |
| [25,30,40,50,60]        |   |
| sa_test_BG = True       | Test to fit background or not?  |
| sa_test_Dark = True     | Test to correct for dark or not?  |
| $sa\_test\_Stat = True$ | Test to remove static image or not?   |
| im_test_klips =         | klip values to be tested for imagette extraction.                           |
| [1,3,5,10]              |   |
| im_test_fitrads =       | fitrad values to be tested for imagette extraction (no point in testing     |
| [25,30]                 | radii larger than the imagette).  |
| im_test_BG = True       | Test to fit background or not?  |
| im_test_Dark = True     | Test to correct for dark or not?  |
| im_test_Stat = True     | Test to remove static image or not?   |

Table 7: Optimisation parameters.

## A.6 Save switches

These parameters decide whether to save specific diagnostic output files, typically as fits image cubes or tables in the output directory. The log text file and extracted photometry fits tables are always saved and have no switches.

Table 8: Save switches.

| PIPE parameter        | Function  |
|-----------------------|---|
| save_mask_cube = True | Saves a binary cube where all pixels in each frame that has been masked     |
|                       | is marked.  |
| save_resid_cube =     | Saves an image cube of residuals where the models for the target, back-     |
| True                  | ground stars and their smearing have been subtracted from the reduced       |
|                       | data. If defined, the static image is also subtracted.                      |
| save_bg_mask_cube =   | If a background star mask is defined and this oparameter is True, then      |
| True                  | a binary image cube with all pixels masked from background stars is         |
|                       | saved.  |
| save_bg_cube = False  | Save an image cube with a model for the target star removed, thus           |
|                       | keeping all the background stars in the cube.                               |
| save_bg_models =      | Saves an image cube with the modelled background, including stars,          |
| False                 | smearing and the static image.  |
| save_static = False   | Saves the static image.   |
| save_psfmodel =       | Saves an image cube with the fitted PSF model for each frame.               |
| False                 |   |
| save_psf_list = True  | Saves a text file with a list of the PSF models used in extraction. This    |
|                       | file can be used as input to ensure that the same PSF models are used       |
|                       | for extraction. This can be useful when extracting photometry from          |
|                       | multiple visits and inter-visit photometry is of interest, since it removes |
|                       | the photometric effect from switching PSF models. This, however, may        |
|                       | also result in less than optimal PSF models being used for extraction.      |
|                       | See §4.4 for more details on how to use it.                                 |
| save_psf_pc = False   | This saves an image cube of all derived PSF principal components with       |
|                       | one component centred in each frame.  |
| save_motion_mat =     | Because the star moves during an exposure, the PSF will be slightly         |
| False                 | blurred due to jitter. To match the observed PSF, the model PSF is          |
|                       | therefore blurred. The motion matrix is a 2D matrix where each side         |
|                       | corresponds to a sub-pixel offset in $x$ and $y$ due to motion.             |
| save_noise_cubes =    | The noise (std) estimated for every pixel and frame is saved in an image    |
| False                 | cube.   |
| save_gain = False     | Save a table with estimated gain (with columns MJD and gain).               |
| save_bg_star_phot =   | If photometry is extracted from background stars, save                      |
| True                  | their photometry in a table (bg_star_photometry_sa.fits or                  |
|                       | bg_star_photometry_im.fits, see §4.5).                                      |