# PEARLS: Near-infrared Photometry in the JWST North Ecliptic Pole Time Domain Field ${ }^{*}$ 

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#### Abstract

We present near-infrared (NIR) ground-based $Y, J, H$, and $K$ imaging obtained in the James Webb Space Telescope (JWST) North Ecliptic Pole Time Domain Field (NEP TDF) using the MMT-Magellan Infrared Imager and Spectrometer on the MMT. These new observations cover a field of approximately $230 \operatorname{arcmin}^{2}$ in $Y, H$, and $K$, and $313 \mathrm{arcmin}^{2}$ in $J$. Using Monte Carlo simulations, we estimate a $1 \sigma$ depth relative to the background sky of $(Y, J, H, K)=(23.80,23.53,23.13,23.28)$ in AB magnitudes for point sources at a $95 \%$ completeness level. These observations are part of the ground-based effort to characterize this region of the sky, supplementing space-based data obtained with Chandra, NuSTAR, XMM, AstroSat, Hubble Space Telescope, and JWST. This paper describes the observations and reduction of the NIR imaging and combines these NIR data with archival imaging in the visible, obtained with the Subaru Hyper-Suprime-Cam, to produce a merged catalog of 57,501 sources. The new observations reported here, plus the corresponding multiwavelength catalog, will provide a baseline for time-domain studies of bright sources in the NEP TDF.


Unified Astronomy Thesaurus concepts: Galaxy evolution (594); Galaxy photometry (611); Galaxy counts (588); Catalogs (205); Surveys (1671)

## 1. Introduction

One of the greatest discoveries in the last few decades is the existence of dark energy, which was inferred from the observations of Type Ia supernovae ranging from the local Universe to redshifts of 1 and above. This discovery has prompted several large-scale surveys of galaxies, e.g., Sloan Digital Sky Survey (SDSS)-IV (Albareti et al. 2017), Palomar Transient Factory (Rau et al. 2009), J-PAS (Bonoli et al. 2021),

[^0][^1]and DESI (Dey et al. 2019) among others, to better characterize dark energy and understand its origin. Because the detection and follow-up of very distant supernovae from the ground becomes very challenging at redshifts of $z \sim 2$ and above, the use of space-based facilities becomes essential. However, a major limitation in using the space observatories is the ability to monitor regions of the sky on a somewhat regular time cadence, which allows identifying and following the changes transient sources undergo.

The combination of mirror size, sensitivity, and smaller Galactic attenuation in the infrared allow the James Webb Space Telescope (JWST) to reach much fainter flux levels than ever before, and this makes it an ideal instrument to detect sources presenting changes, be it through proper motions (e.g., outer solar system bodies and nearby stellar and substellar objects), or in light output (e.g., supernovae, variable stars, and active galactic nuclei).

Jansen \& Windhorst (2018) proposed taking advantage of a region of the sky that can be observed by JWST any time of the year to detect much fainter transient, variable, and moving objects than would be feasible from the ground even with near-future facilities like the Rubin Observatory. Examining the distribution of stars and mid-infrared emitters (bright stars and galactic cirrus), Jansen \& Windhorst (2018) selected a patch of sky close to the North Ecliptic Pole (NEP) within JWST's northern continuous viewing zone (CVZ), which is a position on the sky used for spacecraft housekeeping and will be frequently visited by the telescope. The position chosen in the northern CVZ is centered at (R.A., decl.) $\mathrm{J} 2000=(17: 22: 47.896,+65: 49: 21.54)$ and has a particularly low number density of bright ( $m_{K} \leqslant 15.5 \mathrm{mag}$ ) sources and low Galactic extinction $(E(B-V) \lesssim 0.03 \mathrm{mag})$ where such timemonitoring survey work will be feasible and efficient. This area was dubbed the JWST NEP Time Domain Field (NEP TDF) and will be a prime field for time-domain measurements during the lifetime of the telescope. It is also one of two CVZ targets of the Prime Extragalactic Areas for Reionization and Lensing Science (PEARLS) JWST/GTO program 2738 (PIs: Windhorst \& Hammel) (Windhorst et al. 2023), the other being the IRAC Dark Field (Yan et al. 2023) a few degrees away.

Prior to 2016, the NEP TDF was covered by imaging data from the SDSS (Alam et al. 2015) and Pan-STARRS (Tonry et al. 2012) in the visible, Two Micron All Sky Survey (2MASS; Skrutskie et al. 2006) in the near-infrared (NIR), and the Wide-field Infrared Survey Explorer (Wright et al. 2010) in the mid-infrared. While extremely valuable, these data were not deep enough to detect the very faint sources necessary to vet the astrometry and photometry of the JWST NEP TDF observations. This prompted a multiobservatory effort to collect visible and NIR data for the calibrations and to provide the first epoch of observations to identify bright transients and variable sources. These visible and NIR data also provide counterparts for observations obtained at other wavelengths, both from space, e.g., Chandra (PI: W. Maksym), NuSTAR (PI: F. Civano; Zhao et al. 2021), XMM (PI: F. Civano), eROSITA (PI: A. Merloni \& R. Sunyaev), AstroSat (PI: K. Saha), and Hubble Space Telescope (HST; PIs: R. Jansen \& N. Grogin), and from the ground, e.g., Large Binocular Telescope/Large Binocular Camera (PI: R. Jansen), J-NEP (PI: S. Bonoli \& R. Dupke; Hernán-Caballero et al. 2023), HiPeRCAM/Gran Telescopio de Canarias (PI: V.Dhillon), Hyper-Suprime-Cam/Subaru (HEROES, PI: G. Hasinger \& E. Hu; Songaila et al. 2018; Taylor et al. 2023), NOEMA (PI: S. Cohen), SCUBA2 (PIs: M. Im \& I. Smail; Hyun et al. 2023), Submillimeter Array (PI: G. Fazio), Very Large Array (VLA, PI: R. Windhorst; Hyun et al. 2023; Willner et al. 2023), and LOFAR (PI: R. van Weeren).

In this paper, we report the results of a NIR imaging survey in the NEP TDF. These observations increase the depth relative to 2MASS by more than 5 mag and provide baseline observations for brighter sources detected by JWST. The NIR observations cover a wider field than targeted by JWST and provide NIR counterparts for sources that are outside the JWST footprint but contained within the fields of view of the other space observatories. We supplement the NIR infrared catalogs with visible data obtained from our own reduction of archival Subaru Hyper-Suprime-Cam (HSC) data taken in this part of the sky. In Section 2, we describe the observations; in Section 3, the reduction and calibrations; in Section 4, the analyses of MMT-Magellan Infrared Imager and Spectrometer (MMIRS) and the archival HSC data; and in Section 5, our
conclusions. All magnitudes are quoted in the AB system (Oke \& Gunn 1983), and following Willmer (2018), we adopt the Vega to AB transformations (MMIRS - Y, MMIRS - J, MMIRS - H, MMIRS $-K, \quad$ MMIRS $\left.-K_{\text {spec }}\right)=(0.574, \quad 0.891,1.333, \quad 1.836$, 1.840) mag. These filters will be noted throughout the text as $Y, J, H, K$, and $K_{\text {spec }}$.

## 2. Observations

We imaged the JWST NEP TDF with the MMIRS (McLeod et al. 2012, Chilingarian et al. 2015) on the MMT. The observations were taken in queue mode over 31 nights between UT 2017 May 14 and UT 2019 June 21. In addition to operating MMIRS, the queue observers also recorded the sky conditions, overheads, and any anomalies for each executed group of exposures ("blocks"). A $\log$ of these observations is provided in Table 10 (Appendix A). MMIRS is equipped with a single HAWAII-2RG (H2RG) detector with $2040 \times 2040$ light-sensitive pixels. ${ }^{19}$ The detector pixel scale of $\sim 0!21$ pixel $^{-1}$ gives a field of view of $\sim 6!8 \times 6!8$, so that four pointings (a $2 \times 2$ mosaic) are required to cover the $14^{\prime}$ diameter NEP TDF footprint. MMIRS allows using different gain values-the default low gain of $2.68 \mathrm{e}^{-} \mathrm{DN}^{-1}$ and a high gain of $0.95 \mathrm{e}^{-} \mathrm{DN}^{-1}$, designed for the observations of faint sources. The detector is read out nondestructively using 32 amplifiers, generating "ramps" for each exposure ("data cubes," where the third dimension contains the individual readouts). The observations used the $Y, J, H$, and $K$ filters; however, by accident, during the instrument setup early in the survey, some observations were taken using the wider spectroscopic filter $K_{\text {spec }}$ and others with the (short) $K$ using the high gain value of $0.95 \mathrm{e}^{-} \mathrm{DN}^{-1}$ rather than the default $2.68 \mathrm{e}^{-} \mathrm{DN}^{-1}$. Because of the different throughputs of the $K$ and $K_{\text {spec }}$ filters and the different noise characteristics due to detector gain (which are discussed in greater detail in Section 4.2), we differentiate the observations according to filter and gain values; thus, the (short) $K$ observations are represented as $K_{2.68}$ and $K_{0.95}$. To keep the sky background and source flux within the roughly linear regime and minimize the number of saturated sources, the integration time of individual exposures ranged from $\sim 8 \mathrm{~s}$ in $K$ to $\sim 57 \mathrm{~s}$ in $Y$. The ramps for the NEP TDF observations used the LogGain/ramp_4.5s readout mode for $Y$ and $J$ and ramp_1.475 for $H$ and $K / K_{\text {spec }}$, generating data cubes with 14 , $12,(8,11),{ }^{20}$ and 7 frames in ( $Y, J, H, K+K_{\text {spec }}$ ). For each science image, a corresponding sky-camera exposure was stored, which allows monitoring changes in the sky conditions over the duration of each observing block. At the end of each night, a series of dark frames were taken with integration times identical to those used for the science exposures.
After each exposure, the telescope was moved using a pseudorandom dither pattern to a new position at least three times the expected FWHM (assumed as $0!$ ) away in order to mitigate both bad pixels and persistence. Table 1 shows the exposure times per dither, the total number of images in each filter combining all quadrants, and the average exposure time per pixel. The accumulated total on-target science exposure time was $\sim 3.9,7.1,5.9$, and 9.7 hr in $Y, J, H$, and $K / K_{\text {spec }}$, respectively.

[^2]

Figure 1. Footprint of the imaging for different MMIRS filters in the NEP TDF using a $\sim 23!9 \times 22!6$ Subaru HSC $g$, $i 2$, $z$ color composite as backdrop, with north up and east to the left, and with the grid showing coordinates in sexagesimal units. The NEP TDF is centered at (R.A., decl.) $J_{2000}=(17: 22: 47.896,+65: 49: 21.54)$. The white solid line outlines the $Y$ imaging, the green dashed line outlines $J$, the thick blue dashed-dotted line $H$, and the orange dots $K$. Most outlines overlap each other, with the exception of $J$. The $K$ outline combines observations taken with the (short) $K$ and $K_{\text {spec }}$ filters; the latter occupy the SE corner quadrant. The figure also includes the outline of the completed JWST/NIRCam observations as a thin red solid line. When the MMIRS observations were initiated, a preliminary NEP TDF pointing and set of NIRCam position angles were used. The NIRCam observations as executed use a field center offset to the W, such that the full NEP TDF is covered with MMIRS only in $J$.

Figure 1 shows outlines of the MMIRS NIR observations reported here, as well as outlines of the JWST observations in the NEP TDF that are now complete, overlaid on a background $g, i 2, z$ color composite constructed from our reduction of archival HSC observations taken in 2017 for the HEROES survey (Songaila et al. 2018; Taylor et al. 2023).

## 3. Reduction

### 3.1. Individual Images

The reduction of the MMIRS imaging followed a procedure similar to those used by Labbé et al. (2003), McCracken et al. (2012), and Pelló et al. (2018), taking into account some of the characteristics of the H2RG detector. The first step in the reduction was subtracting the dark current from each science image plane using the average of the corresponding plane in the set of dark images. Next, image counts were corrected for nonlinearity using lookup tables compiled by the MMIRS team, cosmic rays were removed and crosstalk subtracted, after which the count-rate images ("slopes") for each ramp in units of data numbers per second ( $\mathrm{DN} \mathrm{s} \mathrm{s}^{-1}$ ) were calculated. These reduction steps were carried out using the Chilingarian et al. (2015) IDL procedures ${ }^{21}$ with adaptations for use with imaging data by one of the authors (C.N.A.W.), as briefly described in Appendix B. The procedures as modified are publicly available

[^3]at https://github.com/cnaw/mmirs_imaging and in Willner et al. (2023).

The sky subtraction followed the IRAF script xdimsum, ${ }^{22}$ which was specifically designed for the reduction of NIR data; similar algorithms were used by Thompson et al. (1999) and Huang et al. (2003). The first step in the process was removing a baseline sky value from each image and storing the reciprocal of this value. ${ }^{23}$ The initial sky value per slope image was obtained taking a pixel-by-pixel average using the 16 (baselinesubtracted) images taken nearest in time (both before and after a given exposure) with no masking of sources or bad pixels. In the case of images at the beginning or end of a sequence, we used the nearest images in time following or preceding a given observation (still totalling 16). The next step calculated the ratio of the image divided by the sky, from which the median ratio (after outlier rejection) was obtained. The sky image multiplied by this median ratio was then subtracted from the slope image.
Initial source catalogs per (sky-subtracted) image were calculated using SExtractor (Bertin \& Arnouts 1996), and masks were created using the segmentation images, where each detection has its segmentation map extended radially by 2 pixels ( 0 ! 42 ) to mask any residual source light close to the detection-level isophote. The initial catalogs were also used to

[^4]Table 1
Summary of MMT/MMIRS Observations of the NEP TDF

| Filter | Gain |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\left(\mathrm{e}^{\left.-\mathrm{DN}^{-1}\right)}\right.$ | $\mathrm{ZP}_{\mathrm{AB}}$ | Exposure per <br> Image <br> $(\mathrm{s})$ | No. Exposures | Total Exposure <br> $(\mathrm{s})$ | Exposure per <br> Pixel <br> $(\mathrm{s})$ | Fields |

Notes.
${ }^{\mathrm{a}} 11.802 \mathrm{~s}$ exposures.
${ }^{\mathrm{b}} 16.228 \mathrm{~s}$ exposures.
identify "ghost" sources due to image persistence. This was done by examining the brightest sources and using the pixel coordinates between sequential images; the persistence images must fall $\lesssim 8$ pixels away from the pixel position in the previous exposure occupied by the bright star and be fainter by more that 2 mag. In general, after two ramps, the remaining signal is negligible compared to the sky background, though in the case of very bright stars, some residuals can be detected even after $\sim 200 \mathrm{~s}$ (in the case of the $Y$ exposures, which are the longest integrations we used). All persistence ghosts had their segmentation maps expanded radially by 5 pixels. Bad (i.e., dead or hot) pixels were also flagged, and their positions were noted in the segmentation images and added to the image masks for each slope image.

After the initial sky-subtraction step, we measured the seeing of individual images with the aim of optimizing the mosaic depth relative to the image resolution. To estimate the seeing, we used the sky-subtracted images to identify stellar sources in each image using distributions of the instrumental surface brightness of the brightest pixel ( $\mu_{\max }$ ) versus the detected area (isoarea_image) to remove hot pixels and residual cosmic rays, as well as the half-light radius (flux_radius) versus the "total" magnitude (mag_auto), all calculated by SExtractor. The total magnitudes used a minimum Kron radius of 3.5 pixels (1!!47 diameter) and a Kron factor of 2.5. The average FWHM for each image was calculated from the measurements of these candidate stars using the biweight estimator (Beers et al. 1990).

Figure 2 shows the number of images in each filter for a given FWHM. The left panel shows the distribution for the $Y$, $J$, and $H$ filters; the right panel shows the distribution for the observations using MMIRS with the $K_{\text {spec }}, K_{0.95}$, and $K_{2.68}$ configurations. The majority of observations peak under $\sim 1^{\prime \prime}$ values, and the distribution of $K_{\text {spec }}, K_{0.95}$, and $K_{2.68}$ reflects the varying conditions during the observations, taken over several nights between 2017 June and 2017 November (see Figure 10 in Appendix A for the variations over individual nights). In the case of $K_{0.95}$, data were taken in a single night, while those for $K_{\text {spec }}$ were taken over the course of two nights. Most of the $K$ values are $\leqslant 1!5$ but have a somewhat broad distribution, in particular in the case of the $K_{2.68}$ observations.

A second calculation of the sky background was then performed using the object masks, after which we refined the World Coordinate System (WCS) of individual images using the http:/ /astrometry.net software of Lang et al. (2010) and the

Gaia-DR3 catalog (Gaia Collaboration et al. 2023) as reference. The improved astrometry for individual images has typical rms residuals $\sim 0!$ ! 03 . During this phase of sky-subtraction, weights for individual images, described by the inverse variance due to the total number of photons in an exposure, the readout noise (3.14 $\mathrm{e}^{-}$), and the gain, have also aggregated a contribution due to the seeing:

$$
\begin{align*}
& \text { weight }_{i, j} \\
& \qquad=\frac{1}{\left(\text { count_rate }_{i, j} \times \text { exptime }+ \text { readnoise } / \text { gain }^{2}\right) \times \text { FWHM }^{2}} . \tag{1}
\end{align*}
$$

Masked pixels were assigned inverse-variance values of zero, so swarp (Bertin et al. 2002; Bertin 2010) ignores these pixels when assembling mosaics, while the lowest-weight values were set to $10^{-9}$ for the SExtractor parameter weight_thresh; pixels with associated weights lower than this value are ignored by SExtractor. With the image masks and weights in place, another round of object detection was carried out. These new catalogs were used by scamp (Bertin 2006) to include higherorder terms in the astrometry, providing WCS keywords compatible with swarp. ${ }^{24}$ The astrometric calibration of scamp also used the Gaia-DR3 catalog matched to sources classified as stellar on individual slope images. The zero-point magnitude offsets between individual slope images as well as initial ensemble zero-point offsets were calculated using sources matched to 2MASS (Skrutskie et al. 2003, 2006) for $J, H, K$, and $K_{\text {spec }}$, and to Pan-STARRS (Tonry et al. 2012) for the $Y$ band. These corrections by scamp were written into ASCII headers subsequently read by swarp when creating the full-field mosaics.

### 3.2. Mosaics

To create the full-field mosaics, we followed a procedure similar to those of Ashcraft et al. $(2018,2023)$ and McCabe et al. (2023), where images in a given filter are combined in two different ways: one to optimize the spatial resolution at the cost of depth, and the other to optimize the depth at the cost of resolution. In those works, the optimal resolution images result

[^5]

Figure 2. Distribution of the seeing FWHM in arcseconds measured for sources classified as stars in each individual image. The left panel shows the distributions for $Y, J$, and $H$, and the right panel the distributions for the low- and high-gain $K_{2.68}$ and $K_{0.95}$ short $K$ filters and $K_{\text {spec }}$, as indicated in the key. The median seeing values for the different configurations are $\left(Y, J, H, K_{0.95}, K_{2.68}, K_{\text {spec }}\right)=(0!!90,0!96,0!85,0!\prime 80,1!06,0!\prime 86)$. The distribution of $K_{2.68}$ values reflects the variation of sky conditions over the several nights the data were collected. When making mosaics for the whole field, a maximum seeing cut at 2 ! 1 was adopted as a compromise between attaining depth and maintaining resolution; this cut mainly affects the $K_{2.68}$ measurements.
from stacking the best-seeing images, comprising about $10 \%$ of the total number of images (Ashcraft et al. 2018), while for the optimal depth, the $5 \%-10 \%$ worst-seeing images are excluded (Ashcraft et al. 2018; McCabe et al. 2023); typically, the cutoff is made at $\sim 2!3$, with the deeper mosaic reaching $\sim 1 \mathrm{mag}$ fainter than the higher-resolution mosaic (McCabe et al. 2023).

Because the aim of the NIR imaging was to provide a first epoch of variability studies, we opted to reach as faint as possible, and following these works, we cut at 2 !! 1 when stacking the images. The greatest impact of this cut is on the $K_{2.68}$ images, where 213 out of 2571 ( $\sim 8 \%$ ) were excluded. These numbers are smaller for the remaining setups-26/796 ( $\sim 3 \%$ ) for $K_{\text {spec }}, 7 / 563$ for $K_{0.95}, 3 / 1502$ for $H$, and $1 / 248$ for $Y$. No $J$ images were excluded.

In addition to mosaics of individual images, stacked images combining the $Y+J, K+K_{\text {spec }}$, and $H+K+K_{\text {spec }}$ were created to have deeper (or in the case of $K+K_{\text {spec }}$, full-field) coverage. In addition to the image mosaics, swarp calculated a stacked weight image that includes the contribution from individual image weights. For the output mosaics, we adopted the pixel size of 0 !' 168 and center position (17:22:48.1298, $+65: 50: 14.853$, J2000) used in the Subaru/HSC imaging discussed in Section 4.3. The final mosaics were combined using swarp parameters combine_type $=$ clipped with clip_sigma $=3.0$.

The source catalogs were generated in a three-step process. The first step used SExtractor with a high detection threshold ( $5 \sigma$ ) per pixel relative to the average background, optimized to detect stars; these candidate sources were then processed by PSFEX (Bertin 2011) to calculate the average point-spread function (PSF) for each mosaic. The PSFs were used in a second pass of SExtractor to calculate the photometry and image classification using the parameters listed in Table 2. All runs used the dual-image mode, where the detection images were mosaics combining $J+Y$ for $J$ and $Y$, $H+K+K_{\text {spec }}$ for $H$, and combined $K+K_{\text {spec }}$ for the $K$ and $K_{\text {spec }}$ images. The positions of these catalogs are in the GaiaDR3 system, and using Gaia-DR3 stars we find that the MMIRS positions have uncertainties $\lesssim 0$.' 070 . The photometric zero-points reported in Table 2 were calculated following Almeida-Fernandes et al. (2022), who used external photometric measurements coming from wide-area surveys (e.g., Gaia and SDSS) with the stellar population models of Coelho (2014) to calculate the unknown zero-point offsets of

Table 2
MMIRS SExtractor Detection Parameters

| Parameter | Value | Note |
| :--- | :--- | :--- |
| DETECT_THRESH | 1.0 | $\sigma$ above mean <br> background |
|  |  | $\sigma$ above mean <br> background |
| ANALYSIS_THRESH | 1.0 | pixels |
|  |  |  |
| DETECT_MINAREA | 10 |  |
| DEBLEND_NTHRESH | 32 |  |
| DEBLEND_MINCONT | 0.005 |  |
| FILTER_NAME | tophat_5.0_5 $\times 5$. |  |
|  | conv |  |
| PHOT_FLUXFRAC | 0.5 |  |
| PHOT_AUTOPARAMS | $2.5,3.5$ |  |
|  |  |  |
| PHOT_AUTOAPERS | $0.0,0.0$ |  |
| CLEAN_PARAM |  |  |

uncalibrated filters. This procedure creates a grid of flux measurements by folding the stellar population models with the filter curves of the reference data and considering several values of Galactic extinction. By means of $\chi^{2}$ minimization, the best-fitting model is found that also provides an estimate of the Galactic (and atmospheric) extinction. From the ensemble of fitted stars, the best-fitting zero-point is then estimated.
In this calibration, we matched the MMIRS-derived catalogs with Gaia, SDSS (Abdurro'uf et al. 2022), Pan-STARRS, and 2MASS, and only used sources that were classified as stars in Gaia and also had measurements in all other catalogs. While the number


Figure 3. Completeness levels for the individual bands and for the combined $K+K_{\text {spec }}$ mosaic $(g)$ for point sources (orange points and lines) and for $n=1$ Sérsic profiles (blue). The vertical lines indicate (from left to right) the $95 \%, 80 \%$, and $50 \%$ completeness levels.
of sources used is small (ranging from 32 in the case of $K_{\text {spec }}$ to 136 for $J$ ), in particular compared to the thousands used by Almeida-Fernandes et al. (2022), the average values for the zeropoint estimates show uncertainties of only $\sim 1 \%-4 \%$. These are much smaller than the estimates calculated using scamp, which are on the order of $10 \%-30 \%$ (per image) when using 2 MASS measurements alone. These final zero-point values are shown in the third column of Table 1.

## 4. Results

### 4.1. Catalog Completeness

To create the catalogs for each MMIRS filter, we used exposure maps to remove sources contained in regions covered by fewer than (5, 4, 7, 9, 9, 9) images for $Y, J, H, K_{0.95}, K_{2.68}$, $K_{\text {spec }}$ because of the low signal-to-noise ratio and greater likelihood of including spurious detections. These regions are those outlined in Figure 1.
The completeness of individual bands and for the $K+K_{\text {spec }}$ mosaic was estimated following, e.g., Caldwell (2006) and Finkelstein et al. (2015), by adding simulated sources to cutouts of the science images. We used sections of $1000 \times 1000$ pixels-with the exception of the $K+K_{\text {spec }}$ mosaic, for which a $2000 \times 2000$ region was used-centered on the position where all $K$ imaging setups overlap. Using a grid of 0.1 mag steps, 100 simulations per magnitude bin were run, where in each simulation 100 sources were added at random positions and catalogs created using the same procedure as for the science catalogs, i.e., using a minimum area of 10 pixels with a detection threshold at $1 \sigma$ above the mean background and using the SExtractor mag_auto estimator for the total magnitude and the detection parameters presented in Table 2. Table 3 lists the completeness limits in each band estimated for point sources modeled by the PSF constructed by PSFEX from the mosaics and for sources with a Sérsic $n=1$ profile, using GalSim (Rowe et al. 2015) models convolved with these PSFs. The Sérsic models were sampled from a uniform distribution of position angle, half-light radius, and axial ratio ranging between $\left[0^{\circ}, 360^{\circ}\right],[0!8,3!3]$, and $[0.3,1)$, respectively. Table 3 shows the completeness limits

Table 3
Completeness Levels

| Filter | $m_{\text {AB }}(95 \%)$ | $m_{\text {AB }}(80 \%)$ | $m_{\text {AB }}(50 \%)$ | Source Type, Minarea ${ }^{\text {a }}$, Threshold ${ }^{\text {b }}$ |
| :---: | :---: | :---: | :---: | :---: |
| $Y$ | 23.80 | 24.22 | 24.59 | point source, 10, 1 |
| Y | 22.32 | 22.82 | 23.23 | Sérsic $n=1,10,1$ |
| $J$ | 23.53 | 24.03 | 24.38 | point source, 10, 1 |
| $J$ | 21.80 | 22.57 | 23.01 | Sérsic $n=1,10,1$ |
| H | 23.13 | 23.52 | 23.88 | point source, 10, 1 |
| H | 21.63 | 22.13 | 22.55 | Sérsic $n=1,10,1$ |
| $K_{0.95}$ | 22.71 | 22.97 | 23.32 | point source, 10, 1 |
| $K_{0.95}$ | 21.26 | 21.59 | 22.00 | Sérsic $n=1,10,1$ |
| $K_{2.68}$ | 22.91 | 23.34 | 23.68 | point source, 10, 1 |
| $K_{2.68}$ | 21.39 | 21.91 | 22.35 | Sérsic $n=1,10,1$ |
| $K_{\text {spec }}$ | 23.33 | 23.58 | 23.93 | point source, 10, 1 |
| $K_{\text {spec }}$ | 21.83 | 22.19 | 22.58 | Sérsic $n=1,10,1$ |
| $K+K_{\text {spec }}$ | 23.28 | 23.62 | 24.02 | point source, 10, 1 |
| $K+K_{\text {spec }}$ | 21.79 | 22.25 | 22.69 | Sérsic $n=1,10,1$ |

## Notes.

${ }^{\text {a }}$ SExtractor minimum detection area (in pixels ${ }^{2}$ of $0!168$ per pixel).
${ }^{b}$ Sextractor detection threshold in terms of standard deviations above the background level.
(in AB magnitudes) at the $95 \%, 80 \%$, and $50 \%$ levels for each of the filters (and gain value in the case of $K$ ) and for the combined $K+K_{\text {spec }}$ mosaic. The completeness curves from which these limits were computed are presented in Figure 3.

### 4.2. Building a Combined K Sample

The different filter throughputs and detector gain values have a small but detectable effect on the data quality of the mosaics. We show in Figure 4 the distribution of the standard deviation of the background flux measured from 100,000 randomly placed 2 !" 1 apertures in each of the $K$-band mosaics. The $K_{2.68}$ measurements (orange distribution) show the smallest dispersion, though there is a large amount of overlap between these and the measurements using the wider $K_{\text {spec }}$ filter (gray) and the $K_{0.95}$ (green) setups. Because of this, we created catalogs for


Figure 4. Distribution of the dispersion of measured background fluxes using 100,000 random apertures of 12.5 pixels ( 2 !! 1 ) calculated for the final science-grade mosaics in $K_{2.68}$ (orange), $K_{0.95}$ (green), and $K_{\text {spec }}$ (gray). The numbers of apertures contributing to the histogram for each filter are proportional to the areas with exposure times greater than 600 s ; these comprise $78 \%, 53 \%$, and $31 \%$ of the random variates for $K_{2.68}, K_{0.95}$, and $K_{\text {spec }}$, respectively. The higher noise level for the $K_{0.95}$ is immediately apparent; the different peaks are associated with the large contiguous regions of the mosaics with very similar exposure times. Shorter exposures shift the peaks toward higher noise values, as expected; the peaks seen for $K_{2.68}$ (from left to right) correspond to total exposures $\sim 9100 \mathrm{~s}, 8600 \mathrm{~s}$, and 3800 s , while for $K_{0.95}$ the peaks correspond to $\sim 4000 \mathrm{~s}$ and 1000 s , and for $K_{\text {spec }}$ the counts peak at $\sim 6000 \mathrm{~s}$.

Table 4
$K$ Magnitude Differences

| Filter and Gain Setup | Mean Difference | Uncertainty $^{\mathrm{a}}$ | $\sigma$ | Matching Sources | Bright (B) or Full (F) Sample |
| :--- | :---: | :---: | :---: | :---: | :---: |
| $K_{2.68}-K_{0.95}$ | -0.031 | 0.007 | 0.282 | 1613 | Quadrants |
| $K_{2.68}-K_{\text {spec }}$ | 0.073 | 0.017 | 0.418 | B |  |
| $K_{0.95}-K_{\text {spec }}$ | 0.122 | 0.010 | 0.340 | SE |  |
| $K_{2.68}-K_{0.95}$ | -0.010 | 0.007 | 0.307 | 1284 | SW |
| $K_{2.68}-K_{\text {spec }}$ | 0.130 | 0.018 | 0.513 | 1921 | B |
| $K_{0.95}-K_{\text {spec }}$ | 0.212 | 0.011 | 0.442 | 776 | F |

Notes.
${ }^{\text {a }}$ The uncertainty is calculated by resistant_mean as $\sqrt{\sigma} /(N-1)$.
${ }^{\mathrm{b}}$ Both sets use gain of $2.68 \mathrm{e}^{-} \mathrm{DN}^{-1}$.
the individual instrument settings. The different peaks seen for the $K_{2.68}$ and $K_{0.95}$ setups are due to a large number of pixels that have very similar exposure times and are concentrated at the center of the mosaics in a given quadrant.

The regions where each of these settings overlap (see Figure 5) were used to estimate average offsets between the different imaging samples (Table 4 and Figure 6). In this calculation, we used the IDL task resistant_mean with a cutoff at $3.5 \sigma$ and restricted the sample to sources with Sextractor flags $=0$. We considered two cases: in the first one, we used sources with magnitudes in the range $18 \leqslant$ mag_auto $\leqslant 23 \mathrm{mag}$ to exclude bright saturated sources and faint sources where the photometry becomes uncertain; the second case repeated the same calculation but without magnitude cuts. From Table 4, using the bright sample, we found an offset of $-0.031 \pm 0.007$ mag between $K_{2.68}$ and $K_{0.95}$. The offset between $K_{0.95}-K_{\text {spec }}$ is $0.122 \pm 0.010 \mathrm{mag}$. Combining the $K_{2.68}-K_{0.95}$ and $K_{0.95}-K_{\text {spec }}$ differences implies that $K_{2.68}-K_{\text {spec }} \sim 0.091$ mag. The direct measurement of $K_{2.68}-K_{\text {spec }}=0.073 \mathrm{mag}$ suggests that constructing a
single-magnitude sample can have systematic uncertainties on the order of $\sim 0.02 \mathrm{mag}$.

To create a uniform sample, we used $K_{2.68}$ as reference magnitudes, subtracted 0.031 mag from $K_{0.98}$, and added 0.073 mag to the $K_{\text {spec }}$ measurements. The final magnitude for each source is an inverse-variance-weighted average using the corrected fluxes in each modality with weights derived from the estimated flux uncertainty measured by SExtractor.

### 4.3. Archival Subaru Hyper-Suprime-Cam Data

To expand the wavelength coverage and provide visible counterparts to the NIR sources, we used archival Subaru Hyper-Suprime-Cam (HSC) data that were available in early 2020 from the SMOKA ${ }^{25}$ archive maintained by the National Astronomical Observatory of Japan. These images were obtained as part of HEROES (Songaila et al. 2018; Taylor et al. 2023) and comprise data of a single field: $N E P$-wide-A05.

[^6]

Figure 5. Distribution in R.A. and decl. of sources in the regions where the three modes of $K$-band imaging overlap. Pink circles show sources in common between all modalities ( $K_{2.68}, K_{0.95}$, and $K_{\text {spec }}$ ). Blue points represent sources with $K_{2.68}$ and $K_{0.95}$ imaging; green points are sources with $K_{0.95}$ and $K_{\text {spec }}$. Sources detected in $K_{\text {spec }}$ and $K_{0.95}$ only are represented by black " + " and green "x" symbols, respectively. Gray diamonds represent sources with $K_{\text {spec }}$ and $K_{2.68}$ measurements.


Figure 6. Comparisons between photometric measurements of sources for the different $K$-band filters and gain values. The full sample of galaxies, with no magnitude cut, is plotted as black diamonds; green dots represent sources that remain after a $3.5 \sigma$ cut. The orange diamonds represent the subsample of galaxies after the magnitude cuts, and black plusses the same sample after a $3.5 \sigma$ cut. The dashed lines represent the mean calculated using the resistant_mean procedure in IDL, and the horizontal dotted-dashed lines the $\pm 1 \sigma$ range for the full sample (green) and the magnitude-limited one (blue). The vertical dashed lines represent the bright and faint magnitude limits used to calculate the average offsets.

Table 5
HSC Observation Log

| Filter | Exposure <br> $(\mathrm{s})$ | $m_{\text {lim }}$ <br> $($ AB Magnitude $)$ | $\lambda_{\text {pivot }} \mathrm{a}$ <br> $(\mu \mathrm{m})$ | $d \lambda^{\mathrm{a}}$ <br> $(\mu \mathrm{m})$ | $F_{\nu 0^{\mathrm{a}}}$ <br> $\left(\mathrm{erg} \mathrm{cm}^{-2} \mathrm{~Hz}^{-1} \mathrm{~s}^{-1}\right)$ |
| :--- | :---: | :---: | :---: | :---: | :---: |
| $g$ | 525 | 25.1 | 0.4780 | 0.1208 | $4.06004 \times 10^{-20}$ |
| $i 2$ | 1200 | 24.6 | 0.7735 | 0.1495 | $2.58121 \times 10^{-20}$ |
| $z$ | 1200 | 24.2 | 0.8910 | 0.0778 | $2.31604 \times 10^{-20}$ |
| NB0816 | 2000 | 24.7 | 0.8177 | 0.0113 | $2.40826 \times 10^{-20}$ |
| NB0921 | 1200 | 24.1 | 0.9214 | 0.0135 | $2.25802 \times 10^{-20}$ |

Note.
${ }^{\text {a }}$ Values from Willmer (2018); https://mips.as.arizona.edu/~cnaw/sun.html.

The observation log is shown in Table 5, which lists the filter name, total exposure time, the limiting magnitude estimated from the peak of the magnitude distribution in 0.5 mag bins, filter pivot wavelength and bandwidth (e.g., Willmer 2018), the flux corresponding to a count of 1 photon per second, and dates of observation.
The data were reduced using the HSC pipeline (Bosch et al. 2018), and for the NEP TDF analyses, a region of $\sim 30!80 \times 33!59$ centered at (17:22:48.1298, $+65: 50: 14.853$, J2000) was excised, which contains the footprints of JWST (PI: R. Windhorst \& H. Hammel) and HST (PI: R. Jansen \& N. Grogin) imaging and most of the other ancillary data, e.g.,

JCMT/SCUBA2 and VLA, (Hyun et al. 2023); Chandra (PI: W. Maksym); NuSTAR (PI: F. Civano; Zhao et al. 2021), and XMM-Newton (PI: F. Civano). The source catalog was created using SExtractor (Bertin \& Arnouts 1996) in dual-imaging mode with a detection image stacking the $i 2, z, N B 0816$, and NB0921 images following Szalay et al. (1999). The detection was run in two steps, "cold" and "hot," where some of the detection parameters are changed to minimize the fragmentation of large and bright galaxies; these parameters are presented in Table 6 and those that change according to the detection mode are noted. The final PEARLS HSC catalog is a single list concatenating the "cold" and "hot" output from the five individual bands, and it contains data for

56,786 sources. The astrometric and photometric calibration used the same process described by Bosch et al. (2018) and Aihara et al. (2022) with the Pan-STARRS DR2 as reference, and it is tied to the Gaia DR1 astrometric reference frame (Gaia Collaboration et al. 2016a, 2016b). The HSC images are calibrated into maggies, ${ }^{26}$ which are in the AB system (Oke \& Gunn 1983), and the flux value corresponding to a magnitude of zero is specified by the FLUXMAG0 image header keyword. ${ }^{27}$ The conversion from maggies into Jy or erg cm ${ }^{-2} \mathrm{~s}^{-1} \mathrm{~Hz}^{-1}$ requires multiplying the image values by $10^{8.9 / 2.5}$ or $10^{-48.6 / 2.5}$, respectively. Because of the flux units adopted by the HSC pipeline, the SExtractor magnitudes are calculated directly in the AB system. However, the expression used by SExtractor to calculate the magnitude uncertainties,

$$
\begin{align*}
& \text { magerr }=\frac{2.5}{\ln (10)} \\
& \quad \times \frac{\sqrt{\text { area } \times \text { variance }(\text { background })+\text { flux } / \text { gain }}}{\text { flux }}, \tag{2}
\end{align*}
$$

assumes flux measurements in $\mathrm{DN} \mathrm{s}^{-1}$ (data numbers per second), and it will produce incorrect uncertainties (by several orders of magnitude) because the statistical uncertainty in photon counts has to be separated from the conversion of photon counts into flux units. The magnitude uncertainties we report are obtained by converting the AB magnitudes into
where area is the total number of pixels covered by the source, and the flux as well as the background variance are values measured by SExtractor. Using the filter parameters in Table 5 and an effective diameter of 820 cm for the Subaru telescope, ${ }^{28}$ this expression produces estimated uncertainties that are on the same order of magnitude as those available in the SDSS database for objects in common and are slightly more conservative than the uncertainty estimates in the full HEROES catalog (Taylor et al. 2023), as discussed in the next paragraph. Both flux and flux uncertainties are converted into nJy by multiplying each value by $10^{31.4 / 2.5}$.
A catalog for the full HEROES survey was published by Taylor et al. (2023), who also used the pipeline described by Bosch et al. (2018) to produce a band-merged aperturematched catalog for the full HEROES region. The HEROES catalog divides the survey into several patches, four of which $(17,9,17,10,18,9$, and 18,10$)$ cover the NEP TDF.

The catalog of unique sources (HEROES_Full_Catalog.fits) was used to recover some of the photometric measurements only available in the patch-level catalogs, e.g., aperture magnitudes. We restricted this catalog to the NEP TDF region: $260.0583 \quad(17: 20: 13.992) \leqslant$ R.A. $\leqslant 261.3350 \quad(17: 25: 20.400)$ and $65.550(65: 33: 00) \leqslant$ decl. $\leqslant 66.125$ (67:07:30). It contains 150,216 sources after setting the flags
is_primary = 1,
and for each filter

$$
\begin{align*}
\{g, i 2, z, & \text { NB816, NB921\}_base_PixelFlags_flag_edge } \\
& =\text { false } \tag{6}
\end{align*}
$$

photon numbers per $\mathrm{cm}^{2}$ per second using

$$
\begin{equation*}
\text { photon_numbers }=\frac{f_{\nu 0} \times d \lambda}{h \times \lambda} \tag{3}
\end{equation*}
$$

where $f_{\nu} 0$ is the flux measured in erg $\mathrm{cm}^{-2} \mathrm{~Hz}^{-1} \mathrm{~s}^{-1}$ corresponding to one event per second in a given filter, $\lambda$ and $d \lambda$ are the filter's characteristic wavelength and bandwidth, and $h$ is Planck's constant. We make the assumption that the instrument gain is already contained in the calibration, and we use the pivot wavelength (e.g., Bessell \& Murphy 2012, Equation (A13)) as the characteristic wavelength. Assuming Poisson errors both for background and source photons, the expression above becomes
and

$$
\begin{align*}
\{g, i 2, z & N B 816, N B 921\} \_ \text {base_PixelFlags_bad } \\
& =\text { false, } \tag{7}
\end{align*}
$$

following Taylor et al. (2023). Matching the latter with the PEARLS catalog of HSC sources produces a list of 55,796 objects in common. To assess the quality of the PEARLS photometry, we selected stellar sources using the distribution of HEROES MAG_PSF magnitudes versus the MAG_PSF - MAG_CMODEL magnitude difference following Bosch et al. (2018). In the comparison that follows, we

$$
\begin{align*}
& \text { magerr }=\frac{2.5}{\ln (10)} \\
& \quad \times \frac{\sqrt{[\text { area } \times \text { variance }(\text { background })+\text { flux }] \times \text { photon_numbers } \times \text { mirror area }}}{\text { flux } \times \text { photon_numbers } \times \text { mirror area }}, \tag{4}
\end{align*}
$$

[^7][^8]Table 6
HSC SExtractor Detection Parameters

| Parameter | Value | Note |
| :---: | :---: | :---: |
| DETECT_THRESH | 5.0 | $\sigma$ above mean background cold mode |
| ANALYSIS_THRESH | 5.0 | $\sigma$ above mean background cold mode |
| DETECT_MINAREA | 10 | pixels |
| DEBLEND_NTHRESH | 32 |  |
| DEBLEND_MINCONT | 0.001 | cold mode |
| DETECT_THRESH | 1.10 | $\sigma$ above mean background hot mode |
| ANALYSIS_THRESH | 1.10 | $\sigma$ above mean background hot mode |
| DETECT_MINAREA | 10 | pixels |
| DEBLEND_NTHRESH | 32 |  |
| DEBLEND_MINCONT | 0.0001 | hot mode |
| FILTER_NAME | $\begin{aligned} & \text { Gauss_1.5_3 } \times 3 \text {. } \\ & \text { conv } \end{aligned}$ |  |
| PHOT_FLUXFRAC | 0.5 |  |
| PHOT_AUTOPARAMS | 2.5,3.5 | Kron factor, minimum radius |
| PHOT_AUTOAPERS | 0.0,0.0 |  |
| CLEAN_PARAM | 0.3 |  |
| WEIGHT_TYPE | BACKGROUND |  |
| RESCALE_WEIGHTS | N |  |
| WEIGHT_THRESH | 1.e-08 |  |
| SEEING | 0! 7 |  |
| SATUR_LEVEL | $8.0 \mathrm{e}-09$ | maggies |

restricted sources to having

$$
\begin{equation*}
\left\|M A G \_P S F-M A G \_C M O D E L\right\| \leqslant 0.01 \tag{8}
\end{equation*}
$$

for all wide-band filters and only considered sources with $17.5 \leqslant i 2 \leqslant 23.5$.

This excised catalog was used to verify the SExtractor photometry after accounting for some of the procedure differences. For example, the HSC pipeline uses pixel radii to define the aperture photometry, whereas SExtractor uses diameters. The HSC pipeline also adds an aperture correction when producing the final magnitudes for stars and galaxies, but it is not used in the aperture magnitudes (Bosch et al. 2018). The photometry comparison used the aperture magnitude measured at a radius of 12 pixels, which corresponds to a $4!!03$ aperture diameter, as this showed the best compromise between the amount of light being measured and the dispersion between both sets of measurements. The results from this comparison are shown in Table 7, where the mean value and estimated uncertainty of the differences in magnitude are calculated using the IDL resistant_mean procedure. The last column shows the multiplicative factor that corrects the fluxes measured by SExtractor to be consistent with the HEROES values.

### 4.4. Merged Visible-NIR Catalog

To create a merged catalog joining the PEARLS HSC and MMIRS photometry, sources were matched using a search radius of $0!18$ between PEARLS HSC and the NIR sources and between the different NIR catalogs. While the position uncertainties of the MMIRS and PEARLS HSC are smaller than this- $0!\prime 070$ and $0!030$, respectively-the larger search radius allows matching the larger and brighter objects. In the cases where there were multiple

Table 7
Magnitude Differences between HEROES and This Work

| Filter | $N_{\text {stars }}$ | Mean Difference | Uncertainty | Flux Correction |
| :--- | :---: | :---: | :---: | :---: |
| $g$ | 2219 | -0.022 | 0.052 | 1.02043 |
| $i 2$ | 2219 | +0.087 | 0.025 | 0.92318 |
| $z$ | 2219 | -0.012 | 0.025 | 1.01125 |
| NB816 | 2219 | -0.009 | 0.033 | 1.00789 |
| NB921 | 2219 | -0.009 | 0.030 | 1.00790 |

matches, these were visually inspected and resolved in the cases of obvious mismatches. The final list of unique sources includes those with matches as well as sources with detections in single NIR bands. The total number of sources in the final merged catalog is 57,501 (including 56,752 PEARLS HSC sources and respectively $(Y, J, H, K)=(8128,10,558,7612,6361)$. Of the sources identified only in the MMIRS imaging, 354 were found to be spurious and removed from the sample. The number of sources with a detection in a single NIR band is $(50,146,151,73)$ for $Y, J, H, K$, respectively. As the number of single detections is small, these were visually inspected to remove spurious sources due to hot pixels, residual cosmic rays, background fluctuations, and stellar diffraction spikes. The final tally is $(1,32,82,35)$ sources in each of the NIR filters, but we note that several sources have PEARLS HSC counterparts that are below the catalog detection threshold. The larger number of single detections in $H$ is due to the use of the combined $H$ and $K$ images in the source detection and the better image quality of the observations (e.g., Figure 2). Once the spurious detections are removed, the remaining number of sources is $(Y, J, H, K)=(7990,10355$, 7535,6316 ), and the total number of bona fide sources is 57,467 , of which 715 are NIR-detected without counterparts in the PEARLS HSC catalog.

The number of sources without PEARLS HSC detections ${ }^{29}$ but present in all NIR ( $Y, J, H, K$ ) catalogs (" $z$-dropouts") is 118; the number of those with $H$ and $K$ (" $J$-dropouts") is 323 . The latter comprise candidate objects close to bright sources that were not deblended correctly in the PEARLS HSC catalog ( 20 sources, $\sim 6 \%$ of the unmatched subsample), sources with PEARLS HSC counterparts below the detection level (96 sources, $\sim 29.5 \%$ ), and legitimate higher-redshift sources (210 galaxies, $\sim 64.5 \%$ ). Given the relatively bright limits of these samples, these "dropouts" are most likely caused by the Balmer break shifting through the HSC and MMIRS filters. To better understand the properties of these sources without PEARLS HSC matches, we matched the PEARLS HSC-MMIRS catalog with a preliminary catalog of the NEP TDF comprising the first two spokes, which contains 24,119 sources. Using a $0!50$ matching radius, we identify 1858 counterparts, 1419 of which have MMIRS measurements in at least one filter. Figure 7 shows the 174 sources with no PEARLS HSC counterparts but detected in at least one MMIRS filter. Of these, eight have only $K$ measurements (red circles), 66 have $H$ and $K$ (orange triangles), and the remaining 29 are detected in all NIR filters (blue triangles). The gray circles represent the remaining 69 sources that have no HSC detection and have one or two NIR detections but not necessarily in contiguous filters; in one case, this is a source contaminated by a diffraction spike. The visual inspection of these objects suggests the majority are higher-

[^9]

Figure 7. Color-color plot of sources matching the PEARLS HSC-MMIRS with a preliminary JWST-TDF catalog measured from the first two NEP TDF spokes. The magnitudes of the latter were measured using SExtractor with a fixed aperture determined from the F444W filter. The 174 plotted NIR sources have no HSC counterparts, and the key identifies the different characteristics, i.e., sources that have at least one NIR measurement (gray dots; sources with measurements in all NIR filters (blue upward-pointing triangles); sources with measurements in $H$ and $K$ (orange downward-pointing triangles) and only in $K$ (red circles). The five galaxies that match SCUBA2 sources are noted by open black squares, and the 28 MMIRS-detected galaxies with no HEROES counterparts are represented by green open diamonds.


Figure 8. Star-galaxy separation using the total magnitude (mag_auto) vs. half-light radius measured from the HSC $z$-band photometry. Green plus signs represent sources classified as stars; small blue squares represent galaxies; brown asterisks represent saturated stars or merged images where one or more sources is a star. For sources without an HSC match within $0!!80$, the NIR half-light radius is used, and these are represented by black diamonds. The vertical lines represent the completeness for extended sources for the $95 \%$ level (dashed line), $80 \%$ level (dotted-dashed line), and $50 \%$ level (dotted line).
redshift ( $z \gtrsim 1.5$ ) or dusty galaxies-five of these galaxies are also SCUBA2 sources discovered by Hyun et al. (2023). One source is a quiescent galaxy at $z \sim 2$ (N. Adams 2023, private communication).

We used the HEROES (Taylor et al. 2023) subcatalog discussed in Section 4.3 to have an independent assessment of the 715 galaxies without PEARLS HSC matches. HEROES includes additional photometry specifically obtained in the

Table 8
Number Counts ${ }^{\text {a }}$


## Note.

${ }^{\text {a }}$ Italics refer to counts affected by incompleteness estimated from the Monte Carlo simulations of Section 4.1


Figure 9. Differential galaxy number counts $\left(N_{c}(m)\right)$ measured in this work (black open circles) compared to other measurements in the literature (symbols as noted in the key) for the NIR filters. The vertical line segments indicate the location of the $95 \%, 80 \%$, and $50 \%$ completeness levels of Table 3 for extended sources as dashed, dotted-dashed, and dotted lines, respectively. Data being plotted come from Bielby et al. (2012), Conselice et al. (2008), Cristóbal-Hornillos et al. (2009), Davies et al. (2021), Fontana et al. (2014), Gardner et al. (1993), Iovino et al. (2005), Huang et al. (1997), Jeon et al. (2014), Keenan et al. (2010), Laigle et al. (2016), Metcalfe et al. (2006), and Windhorst et al. (2011). For references where Table 9 tabulates magnitude bins $=0.5$, the number counts in galaxies $0.5 \mathrm{mag}^{-1}$ per unit solid angle were converted into galaxies per unit magnitude per unit solid angle.

NEP TDF with significantly deeper imaging in the $z$ band and additional data in the HSC-r2 and HSC-y bands. We find 562 sources with HEROES counterparts within 0!! 5, while 153 have no HEROES match. Of these, 12 are detected in all MMIRS bands, 93 in $H$ and $K$, and 20 in $K$ only; five are detected in $J$, where there are no MMIRS observations in the other filters. There are 17 objects detected in $Y$ and $J$ only, and most of these are due to imperfect matches, because of merged images or sources close to bright stars. We also matched this catalog to the JWST catalog of spokes 1 and 2, and we find 28 objects in common, which are identified in Figure 7 as green diamonds. Four of these are detected in $K$ only, and the majority (16) are detected in both $H$ and $K$.

The matched PEARLS HSC-MMIRS-JWST catalog was also used to verify if any sources show evidence of variability, comparing measurements in the closest pair of filters (e.g., F090W $-z$, F115W $-J$, F150 $-H$, and F200W $-K$ ). When considering only the unsaturated bright sources, there is no evidence of variability. The sources presenting large magnitude differences are all due to contamination either by diffraction spikes or very close bright objects.

### 4.5. Star/Galaxy Separation and Number Counts

The PEARLS HSC+MMIRS merged catalog was used to classify sources detected in the NIR bands as stars or galaxies (or more correctly, compact or extended sources). For this, we used the total magnitude versus the half-light radius, e.g., Kron (1980), measured from the HSC $z$ band and the total magnitude (mag_auto) in each NIR band. We used the radii measured for the PEARLS HSC because of the greater depth and overall better image seeing. In the case of MMIRS sources without PEARLS HSC half-light radius measurements, we used the corresponding NIR-band half-light radius values. The results from this classification are shown in Figure 8, where sources of different types are distinguished by symbol shapes and colors as noted in the key of each panel. We also show the $95 \%, 80 \%$, and $50 \%$ completeness levels for extended sources as dashed, dotted-dashed, and dotted lines. The plots show that, below the $80 \%$ completeness level, the star/galaxy classification starts breaking down, though this changes with the filter being used.

In addition to characterizing the large-scale properties of a sample (e.g., Koo \& Kron 1992; Smail et al. 1995), the number counts (shown in Table 8) provide an independent check on the

Table 9
Depth and Area of Literature Number Counts

| References | Filter | Magnitude Bin | Area | Limiting Magnitude [ AB$]^{\text {a }}$ |
| :---: | :---: | :---: | :---: | :---: |
| Davies et al. (2021) | $Y$ | 0.5 | $0.6 \mathrm{deg}^{2}$ | 25.5 |
| Windhorst et al. (2011) | $Y$ | 0.5 | $\sim 45 \mathrm{arcmin}^{2}$ | 26.0 |
| Bielby et al. (2012) | $J$ | 0.5 | $2.1 \mathrm{deg}^{2}$ | 23.8 |
| Conselice et al. (2008) | $J$ | 1.0 | $1.53 \mathrm{deg}^{2}$ | 23.4 |
| Cristóbal-Hornillos et al. (2009) | $J$ | 1.0 | $0.5 \mathrm{deg}^{2}$ | 23.3 |
| Davies et al. (2021) | $J$ | 0.5 | $6 \mathrm{deg}^{2}$ | 25.2 |
| Iovino et al. (2005) | $J$ | 0.5 | $400 \mathrm{arcmin}^{2}$ | 22.4 |
| Jeon et al. (2014) | $J$ | 0.5 | $5.1 \mathrm{deg}^{2}$ | 20.75 |
| Keenan et al. (2010) | $J$ | 1.0 | $2.77 \mathrm{deg}^{2}$ | 23.0 to 24.4 |
| Windhorst et al. (2011) | J | 0.5 | $\sim 45 \mathrm{arcmin}^{2}$ | 26.0 |
| Bielby et al. (2012) | H | 0.5 | $2.1 \mathrm{deg}^{2}$ | 23.8 |
| Cristóbal-Hornillos et al. (2009) | H | 1.0 | $0.5 \mathrm{deg}^{2}$ | 22.6 |
| Davies et al. (2021) | H | 0.5 | $6 \mathrm{deg}^{2}$ | 25.0 |
| Jeon et al. (2014) | H | 0.5 | $5.1 \mathrm{deg}^{2}$ | 20.75 |
| Keenan et al. (2010) | H | 1.0 | $3.6 \mathrm{deg}^{2}$ | 22.5 to 24.0 |
| Metcalfe et al. (2006) | H | 0.5 | $49 \mathrm{arcmin}^{2}$ | 22.5 |
| Windhorst et al. (2011) | H | 0.5 | $\sim 45 \mathrm{arcmin}^{2}$ | 26.0 |
| Bielby et al. (2012) | K | 0.5 | $2.1 \mathrm{deg}^{2}$ | 23.8 |
| Conselice et al. (2008) | K | 1.0 | $1.53 \mathrm{deg}^{2}$ | 23.8 |
| Cristóbal-Hornillos et al. (2009) | K | 1.0 | $0.5 \mathrm{deg}^{2}$ | 21.8 |
| Davies et al. (2021) | K | 0.5 | $6 \mathrm{deg}^{2}$ | 24.8 |
| Fontana et al. (2014) | K | 1.0 | $340.58 \mathrm{arcmin}^{2}$ | 26.5 |
| Gardner et al. (1993) | K | 0.5 | $1.16 \mathrm{arcmin}^{2}$ to $41.56 \mathrm{deg}^{2}$ | 24.3 |
| Huang et al. (1997) | K | 1.0 | $1.58 \mathrm{deg}^{2}$ | 16.3 |
| Iovino et al. (2005) | K | 0.5 | $400 \operatorname{arcmin}^{2}$ | 22.6 |
| Keenan et al. (2010) | K | 1.0 | $3.94 \mathrm{deg}^{2}$ | 23.0 to 24.8 |
| Laigle et al. (2016) | K | 0.5 | $0.62 \mathrm{deg}^{2}$ | 24.7 |

Note.
${ }^{\text {a }}$ Values quoted by the original references.
quality of the photometry, the effectiveness of the star/ galaxy separation (e.g., Conselice et al. 2008), and the completeness of the catalog (e.g., Windhorst et al. 2023). Figure 9 shows the number counts $N_{c}(m)=\frac{N(m)}{\Delta m \times \text { area }}$ measured in the NEP TDF for the $Y, J, H$, and combined $K$ samples. The uncertainties in Table 8 assume Poisson statistics $\delta N_{c}(m)=(\sqrt{N(m)} \times$ area $-N(m) \times \Delta$ area $) /(\Delta m \times \text { area })^{2}$ and are underestimated at larger numbers of galaxies because no cosmic variance is included. Vertical lines representing the completeness levels for extended sources from Table 3 are also shown and indicate that the turnover due to incompleteness is well characterized by the Monte Carlo simulations. The figures also show a selection of published measurements noted in Table 9, and when required, the latter were converted into AB magnitudes by adding the Vega to AB offsets of $(J, H, K)=(0.870,1.344,1.814) \mathrm{mag}$ derived for the 2MASS (Skrutskie et al. 2006) filters by Willmer (2018). The published counts in $N_{\mathrm{gal}} 0.5 \mathrm{mag}^{-1}$ per unit solid angle were converted into $N_{\text {gal }} \mathrm{mag}^{-1} \mathrm{deg}^{-2}$. The areas and limiting magnitudes come from the faintest magnitudes in the original references. ${ }^{30}$ Given the small area covered by the NEP TDF MMIRS imaging- $(0.0635,0.0870,0.0643,0.0634) \operatorname{deg}^{2}$ for $(Y, J, H, K)$, respectively)-some differences relative to other measurements are expected, though the agreement is good for magnitudes of 18 and fainter in all four bands.
The deepest data in $Y, J$, and $H$ come from the HST imaging in GOODS-S by the Early Science Release using the Wide-

[^10]Field Camera 3 of Windhorst et al. (2011). In $K$, the deepest coverage comes from the ground-based observations of Fontana et al. (2014) in the UDS and GOODS-S fields Because of the large number of samples in $K$, it is also the band showing the greatest variation between samples. The greatest differences are seen for the $K$ number counts at bright magnitudes by Bielby et al. (2012) and Fontana et al. (2014), because of small number statistics. Fainter than $K \sim 18$, the largest differences are seen for the Iovino et al. (2005) and Conselice et al. (2008) counts. However, the $J$ counts of both references are good agreement with other works. For $J, H$, and $K$, the MMIRS counts show good agreement with the multifield number counts of Keenan et al. (2010) throughout the common magnitude range.

## 5. Conclusion

We present the results of NIR imaging in the JWST North Ecliptic Pole Time Domain Field in the $Y, J, H$, and $K$ bands. These observations reach a $95 \%$ completeness level for AB magnitudes of $(Y, J, H, K)=(22.3,21.8,21.6,21.8)$ estimated from simulations of extended sources described by an exponential Sérsic profile. For point sources, the corresponding $95 \%$ completeness levels are $(Y, J, H, K)=(23.8,23.5,23.1$, 23.3) in AB magnitudes. These data are combined with catalogs derived from archival Subaru Hyper-Suprime-Cam observations in the ( $g, i 2, z, N B 816, N B 921$ ) bands to create a matched visible-NIR catalog, with positional precision with a dispersion $\sim 20$ mas. A comparison with a catalog of sources from the first two JWST observations of the NEP TDF suggests that a small number of HSC-MMIRS sources are galaxies at
redshifts $z \gtrsim 1.5$ and above or dusty galaxies. The expected number of galaxies at redshifts $\gtrsim 2$ is estimated to be $\sim 10$ using the redshift versus number density diagram of Conselice et al. (2008) (their Figure 3), who reach a comparable depth in the $K$ band imaging, though this number is very uncertain, particularly given the small area the MMIRS imaging covers. The examination of bright unsaturated sources in the joint HSC-MMIRS-JWST shows no source with significant ( $\Delta m \geqslant$ 0.1 mag ) variability. These observations provide the first epoch of deep NIR photometry in this part of the sky and will continue being used in the search for transient and variable objects in the NEP TDF. This region of the sky has already amassed in very few years an extensive coverage in wavelength space, ranging from X-rays to radio frequencies.

The mosaics and catalogs are publicly available at https://sites. google.com/view/jwstpearls/data-products and on Zenodo under an open-source Creative Commons Attribution license: doi:10.5281/zenodo. 7934393 .

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Facilities: MMT (MMIRS), Gaia, HST (ACS, WFC3), Subaru, IRSA (2MASS), NED, MAST (Pan-STARRS), Sloan, ADS

Software: Fortran, Perl, IDL, Astrometry.net (https:// astrometry.net/, Lang et al. 2010), GalSim (Rowe et al. 2015), HSC pipeline (Bosch et al. 2018), IDL Astronomy User's Library (Landsman 1995), IDL wrapper scripts for MMIRS reduction (Willmer 2023), Matplotlib (Barrett et al. 2005), MMIRS pipeline (Chilingarian et al. 2015), MPFIT (Markwardt 2009), PGPLOT (https://sites.astro.caltech.edu/tjp/pgplot), PSFEX (Bertin 2011), SAOimage DS9 (Joye 2011), SCAMP (Bertin 2006), SExtractor (Bertin \& Arnouts 1996), SWARP (Bertin et al. 2002), TOPCAT (Taylor 2005).

## Appendix A MMIRS Observation log

Table 10 is the MMIRS observation log, which identifies the date of observation (UT), the modified Julian date, the mean seeing value and dispersion measured from all images taken during the night, the quadrant(s) and filter(s), and in parenthesis the gain values. When no gain value is shown, the default $2.68 \mathrm{e}^{-} \mathrm{DN}^{-1}$ gain was used. Figure 10 shows the distribution of the average seeing FWHM value and dispersion measured for each filter (as noted by the color in the key) and each night the filter was used.

Table 10
MMIRS Observation Log

| Date | MJD | Seeing (arcseconds) | Uncertainty (arcseconds) | Field, Filter |
| :---: | :---: | :---: | :---: | :---: |
| 2017-05-14 | 57,887 | 1.117 | 0.155 | SE $J$, SW $J$ |
| 2017-06-11 | 57,915 | 1.085 | 0.183 | SW $J$, NE $J$ |
| 2017-06-12 | 57,916 | 1.094 | 0.228 | NW $J$ |
| 2017-06-13 | 57,917 | 0.843 | 0.503 | SE $K_{\text {spec }}$ |
| 2017-06-14 | 57,918 | 1.015 | 0.324 | SE $K_{\text {spec }}$ |
| 2017-06-15 | 57,919 | 0.871 | 0.354 | SE $K_{0.95}$, SW $K_{0.95}, K_{2.68}$ |
| 2017-06-16 | 57,920 | 1.293 | 0.507 | SW $K_{2.68}$ |
| 2017-06-18 | 57,922 | 1.994 | 0.193 | NE $K_{2.68}$ |
| 2017-06-19 | 57,923 | 1.942 | 0.667 | NE $K_{2.68}$ |
| 2017-07-01 | 57,935 | 0.769 | 0.234 | NE $K_{2.68}$ |
| 2017-07-02 | 57,936 | 0.605 | 0.093 | NE $K_{2.68}$ |
| 2017-07-06 | 57,940 | 1.856 | 0.430 | NE $K_{2.68}$ |
| 2017-10-17 | 58,043 | 1.433 | 0.445 | NE $K_{2.68}$, NW $K_{2.68}$ |
| 2017-10-19 | 58,045 | 1.183 | 0.228 | NW $K_{2.68}$ |
| 2017-10-20 | 58,046 | 0.745 | 0.126 | NW $K_{2.68}$ |
| 2017-11-03 | 58,060 | 0.880 | 0.083 | NW $K_{2.68}$ |
| 2017-11-07 | 58,064 | 0.878 | 0.244 | NW $K_{2.68}$ |
| 2018-08-28 | 58,358 | 0.855 | 0.089 | SE $H$, SE $Y$, NE $Y$ |
| 2018-08-29 | 58,359 | 0.811 | 0.138 | SW $Y$, NW $Y$ |
| 2018-08-31 | 58,361 | 0.837 | 0.123 | SE $H$ |
| 2018-11-17 | 58,439 | 0.926 | 0.166 | NW H |
| 2018-11-19 | 58,441 | 0.708 | 0.091 | NW H |
| 2018-11-20 | 58,442 | 0.999 | 0.189 | NW H |
| 2018-11-22 | 58,444 | 1.165 | 0.371 | NW $H$ |
| 2018-11-25 | 58,447 | 0.966 | 0.144 | NW $H$, NE $H$ |
| 2018-11-28 | 58,450 | 1.026 | 0.227 | NE $H$ |
| 2019-06-14 | 58,648 | 1.154 | 0.191 | NE $H$ |
| 2019-06-15 | 58,649 | 0.732 | 0.097 | SW H |
| 2019-06-16 | 58,650 | 0.584 | 0.051 | Center J |
| 2019-06-18 | 58,652 | 0.770 | 0.049 | SW1 J |
| 2019-06-21 | 58,655 | 1.042 | 0.150 | NW1 J |



Figure 10. Distribution of the average seeing FWHM and dispersion measured from all images during a night presented in Table 10. The colors represent different filters as noted in the key.

## Appendix B IDL Imaging Wrapper

A wrapper to run the initial imaging reduction scripts is available at https://github.com/cnaw/mmirs_imaging (also available at Willmer 2023), though currently there is limited documentation on how to use them. Most of the procedures are those of Chilingarian et al. (2015) but with small modifications
in the initial reduction that converts data cubes into slopes. Also in the repository is a perl script, pre_reduction.pl, that will read the contents of the directory where raw images are stored, sort images according to the DARKTIME keyword value, and check whether they are science or calibration data. This script attempts to emulate the perl script described in the Chilingarian et al. (2015) paper. The prereduction script sorts imaging from spectroscopic data (this feature was not tested extensively) and creates a batch file that should be run under IDL. The final product of this step is a set of dark-subtracted, linearity-corrected slope images.

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[^0]:    * Partly based on observations taken with the MMT, a joint facility operated by the University of Arizona and the Smithsonian Institution.

[^1]:    

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[^2]:    19 The remaining pixels ("reference pixels") are used to monitor the detector thermal variations.
    ${ }^{20}$ Due to improvements on the detector readout implemented in 2019, the exposure times for $H$ observations were increased from $\sim 11$ to 16 s , improving the observation efficiency.

[^3]:    ${ }^{21}$ https://bitbucket.org/chil_sai/mmirs-pipeline/wiki/Home

[^4]:    22 https://github.com/iraf-community/iraf-xdimsum
    ${ }^{23}$ xdimsum used a median with iterations to remove outliers; here, we used the outlier-resistant biweight estimator of Beers et al. (1990).

[^5]:    ${ }^{24}$ In particular, the Tan-SIP transform used by Lang et al. (2010) is not recognized by Astromatic software (SExtractor, scamp, and swarp).

[^6]:    ${ }^{25}$ https://smoka.nao.ac.jp/

[^7]:    ${ }^{26}$ https://www.sdss4.org/dr12/algorithms/magnitudes/
    ${ }^{27}$ https://hsc.mtk.nao.ac.jp/pipedoc/pipedoc_8_e/tips_e/read_catalog.html

[^8]:    ${ }^{28}$ https://subarutelescope.org/Observing/Telescope/Parameters/

[^9]:    ${ }^{29}$ That is, a counterpart may exist in the PEARLS HSC imaging but below the catalog detection threshold.

[^10]:    ${ }^{30}$ In the case of Gardner et al. (1993), the authors also included earlier measurements that cover the range of areas noted in the table.

