

#### PSFEx:a generic tool for extracting the PSF in astronomical images

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## Outline

- Use of the PSF for detection, measurement and classification of astronomical sources
- Modeling the PSF with PSFEx
- Finding « prototype » stars
- Quality control at TERAPIX
- PSF-fitting with SExtractor



## **Detection and the PSF**

- Matched (optimum) filtering for detection
  - Stationary noise with power spectrum *P*(*k*) and isolated point-sources: convolve with

$$h = \phi^* * \mathsf{F}(P^{-1})$$



Irwin 1985



# Source-deblending and profile-fitting in crowded star fields

- The PSF profile  $\phi(\mathbf{x})$  can be quickly centered on isolated stars using a simple gradient descent
  - At each step, derive a profile offset  $\Delta \mathbf{x}$  by fitting

$$F.(\phi + (\nabla \phi).\Delta x)$$

 Clumps of overlapping stars can be fitted using the same simple technique with additional constraints (no negative flux, minimum distance between stars)



## Astrometry

- Effects of crowding
- The definition of a star position can be ambiguous for asymmetric PSFs
  - Flux-dependency when centroiding thresholded profiles



## Point-source photometry

• Profile-fitting photometry always optimum in terms of SNR:



- On photon-noise limited images with negligible background
  - $\sigma_i^2 \propto \phi_i$ : profile-fitting equivalent to integration of pixel values within an aperture
- On photon-noise limited images with dominant background
  - $\sigma_i^2 \propto cste$ : profile-fitting equivalent to a profile-weighted sum of pixel values



## Star/galaxy separation

 Local PSF used as a reference for computing the likelihoods p(y|S) and p(y|G) of a star/galaxy Bayesian classifier (Sebok 1979, Valdes 1982 and followers)



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## Morphology of extended sources

- Non-linear galaxy profile-fitting (e.g. GIM2D)
  - Reconvolution with the local PSF needed at each iteration
- Decomposition on basis functions (PCA, shapelets)
  - Basis functions are convolved with the local PSF before fitting

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### Parametric deconvolution of galaxies





### Measuring morphological parameters

- I<23
- 3h exposure with 0.7" seeing (ground-based) on a 3.6m telescope





## Building a model of the PSF

- Software written in 1998 for SExtractor
  Not publicly available yet
- Requirements:
  - Model variations across the field
  - Be able to deal with (moderate) undersampling
  - Number of degrees of freedom as small as possible while being capable of modeling any arbitrary (optical) PSF



### **PSF** models

- Analytical vs tabulated models
  - Analytical models are simpler to implement and can deal with undersampling "naturally"
    - BUT: simple (not instrument-dependent) models have trouble handling PSF features like diffraction effects (spikes and rings)
      - Such features can be tabulated provided that the data are correctly sampled, but this is not always the case (ex: WFPC2, NICMOS,...)
  - Tabulated models don't have these limitations
    - BUT: over- and under-sampling are not properly handled.



## A solution: "super-tabulation"

- The PSF is tabulated at a resolution which depends on the stellar FWHM (typically 3 pixels/FWHM)
  - Minimize redundancy in cases of bad seeing
  - Handle undersampled data by building a "super-tabulated" PSF model
  - Work with diffraction-limited images (images are band-limited by the autocorrelation of the pupil)
  - Find the sample values by solving a system using stars at different positions on the pixel grid
    - Intuitive approach: solve in Fourier space. Easy but suboptimum (no weighting)
    - Working in direct space would give much more robust results



## Solving in Fourier space











## Solving in direct space

 A resampling kernel h, based on a compact interpolating function (*Lanczos3*), links the "super-tabulated" PSF to the real data: the pixel i of star j can be written as

$$P_{ij} = \sum_{k} h_j \left( \vec{\mathbf{x}}_k - \vec{\mathbf{x}}_i \right) \varphi_k$$

- The  $\varphi_k$ 's are derived using a weighted  $\chi^2$  minimization.
  - Lots of computations involved:

Sparse matrix processing might prove useful for large models

The practice the oversampling of faint peripheral pixels can be dropped.



#### Lanczos interpolation kernel





#### Testing on simulated, undersampled data



Diffraction-limited FWHM ≈ 1pixel Moderately crowded



#### Automatic candidate selection

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#### Recovered PSF with simulated, undersampled data







#### Residuals on simulated, undersampled data





#### Simulated, defocused data



Diffraction-limited FWHM ≈ 7 pixels Moderately crowded



#### Results with simulated, defocused data



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## Using different basis functions

• The array of "super-pixels" can be replaced by a combination of ad-hoc basis functions  $\psi_b$  (the  $c_b$  are the parameters to determine)

$$P_{ij} = \sum_{b} \sum_{k} h_j \left( \vec{x}_k - \vec{x}_i \right) c_b \psi_{bk}$$

- Should be more robust in many cases
- One might use PCA components of the theoretical PSF aberrations for diffraction-limited instruments.



## Handling PSF variations

- PSF variations are assumed to be a smooth function of object coordinates
  - The variations can be decomposed on a polynomial basis  $X_{I}$

$$\boldsymbol{P}_{ij} = \sum_{l} X_{l} \sum_{k} h_{j} (\boldsymbol{x}_{k} - \boldsymbol{x}_{i}) \varphi_{kl}$$

- A third order polynomial (*I* = 10) is generally sufficient to describe the variation of the PSF with position in the field
- Different basis functions, with arbitrary parameters (flux, instrumental context, etc.) can be used for  $X_I$
- In our case a KL decomposition (e.g. Lupton et al. 2001) was not beneficial (and in fact it makes the rejection of « bad » PSF prototypes harder).



## Example of $\varphi_{lk}$ PSF components for a UH8k image





#### **Reconstructed UH8k PSF**





#### Testing on real, non-linear data



Schmidt-plate exposures in the galactic plane FWHM ≈ 3pixel Second order polynomial of FLUX\_AUTO





#### Star subtraction on Schmidt-plate data



Schmidt-plate exposures in the galactic plane FWHM ≈ 3pixel Second order polynomial of FLUX\_AUTO



## Finding prototype stars

- Basically we are looking for something we don't know yet
  - PSF variability makes the stellar locus "fuzzy" in feature space
  - Problems due to crowding at low galactic latitude
  - Confusion with galaxies in cluster areas
- Empirically designed automatic selection based on magnitude,half-light radius, ellipticity, crowding and saturation flags seems to work fine
  - Remaining configuration parameters for selection essentially consist of acceptable FWHM range and ellipticity
  - Iterative rejection procedure based on similarity between samples and a rough PSF estimate



#### Half-light radius/magnitude diagram





## QualityFITS

- AstroWISE project developed at TERAPIX by F. Magnard
- Provides quality control for FITS images
  - Background homogeneity
  - PSF and variability
  - Source counts
  - Weight maps
- Diagnostic generated automatically for all incoming and outgoing MEGACAM survey images
  - FITS and XML formats
  - Access from <u>Spica</u>





## Fitting the PSF model

- Identify star "clusters", like in DAOPhot (Stetson 1987) and proceed interatively:
  - First a unique star is fitted
    - The basic centering algorithm is a modified gradient descent
  - The star is subtracted from the cluster and a local maximum sufficiently distant from the peak of the first star is identified
  - Two stars are fitted and subtracted, and a new maximum is found
    - Iterate up to 11 stars/cluster or
    - Stop if stars coalesce during the centering process



## Current Performance

- Processing speed:
  - For building the PSF model: ~130 stars/second (Athlon 2GHz)
  - For the PSF-fitting: ~100-500 stars/second (Athlon 2GHz)
- Measurement accuracy:
  - Slightly better than DAOPhot on properly sampled, non-crowded fields
  - Slightly worse than DAOPhot (one pass) on properly sampled, crowded fields
  - Significantly better than DAOPhot on undersampled images
- Poor completeness (~99% for "obvious" detections) because of the underlying SExtractor detection scheme



Application: Comparison with DAOPhot on NGC 6819 (CFH12k)



Kalirai et al. 2001a

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Application: Photometric accuracy in NGC 6819 (CFH12k)



Kalirai et al. 2001b

E. Bertin



Application: Colour-magnitude diagrams in NGC 6819 (CFH12k)



Kalirai et al. 2001b



## Conclusions

- The PSFEx approach to PSF modeling gives reliables results
  - Undersampled data (down to 1 pixel FWHM)
  - Variability across the field
  - Moderately crowded fields
- Currently available as an external module: "PSFEx"
  - Soon to be publicly released (together with QualityFITS)
  - But not for PSF fitting in SExtractor
    - Mostly completeness issues
- Wait for SExtractor3
  - New detection scheme
  - Handling of variable noise ACF