

Put me here!

Towards Solving the Catalogue Cross-Match Problem

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University of Exeter

Alternatively — Space Is Crowded, Or: How I Learned To Start Worrying and Hate the Mess



Exeter Seminar, 8/Apr/21

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Photometric Observations

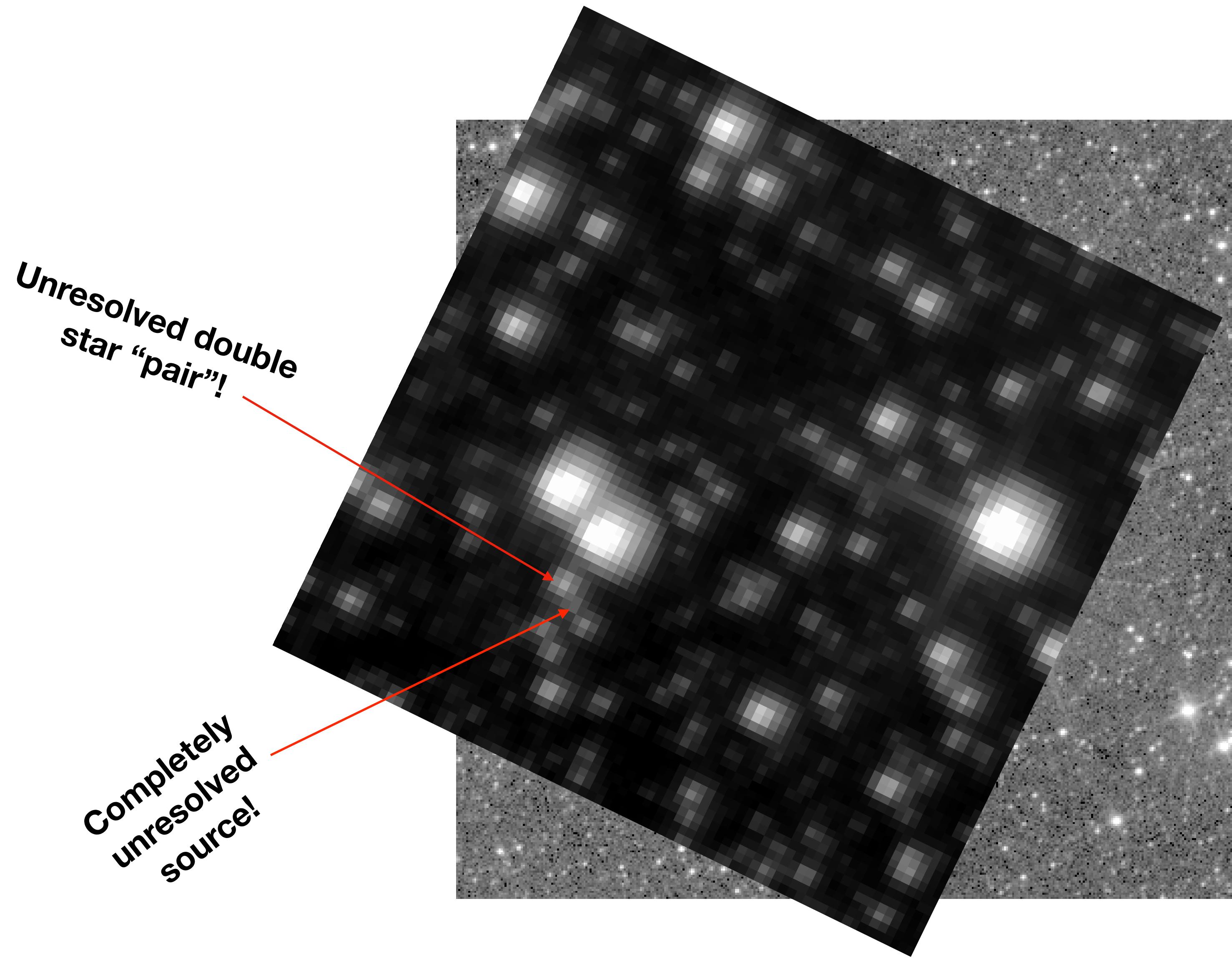


WISE - Wright et al. (2010)

WISE W1

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Photometric Observations



WISE - Wright et al. (2010)
TESS - Ricker et al. (2015)

TESS T
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Matching Constellations



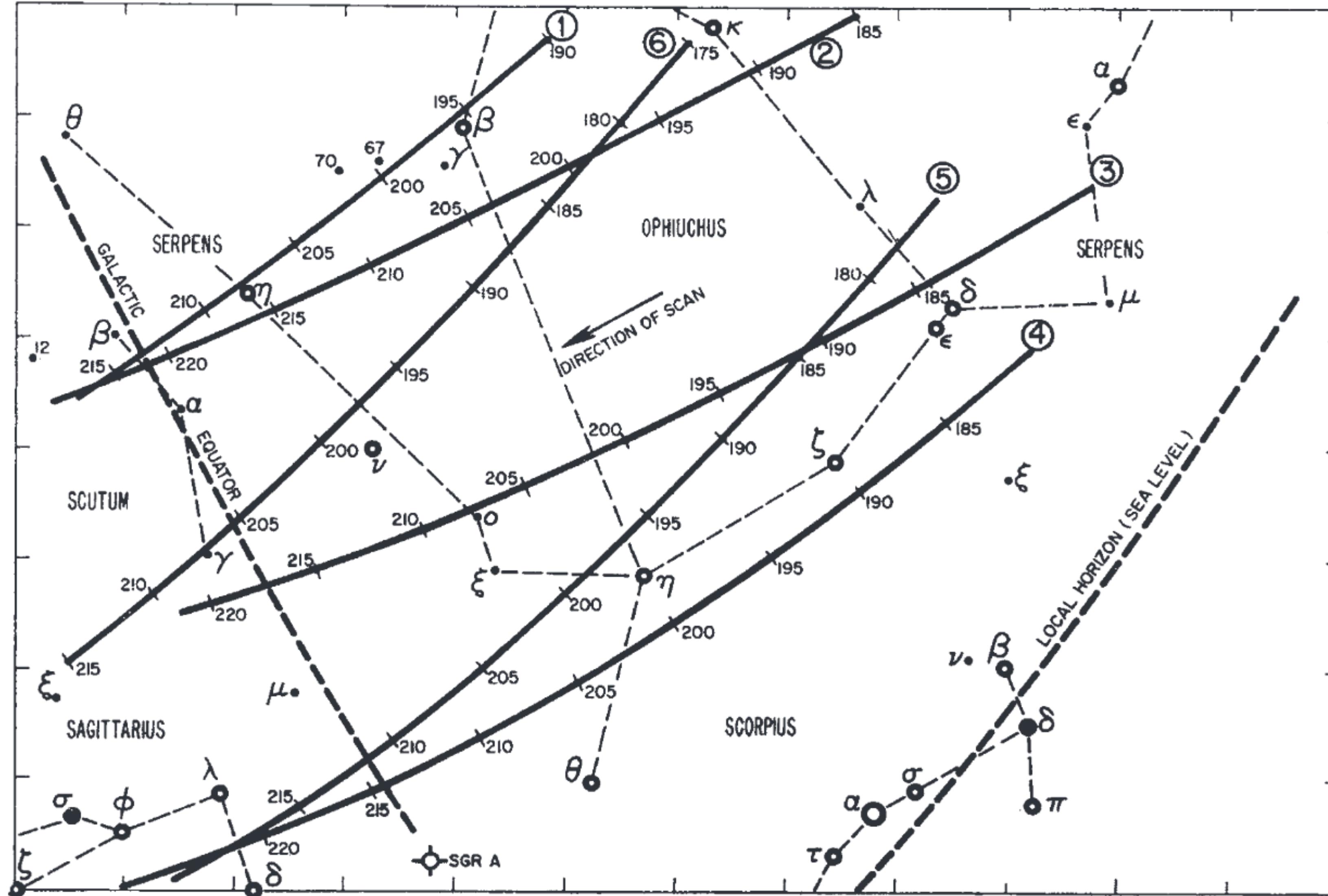
Image credit: Mouser, wikipedia

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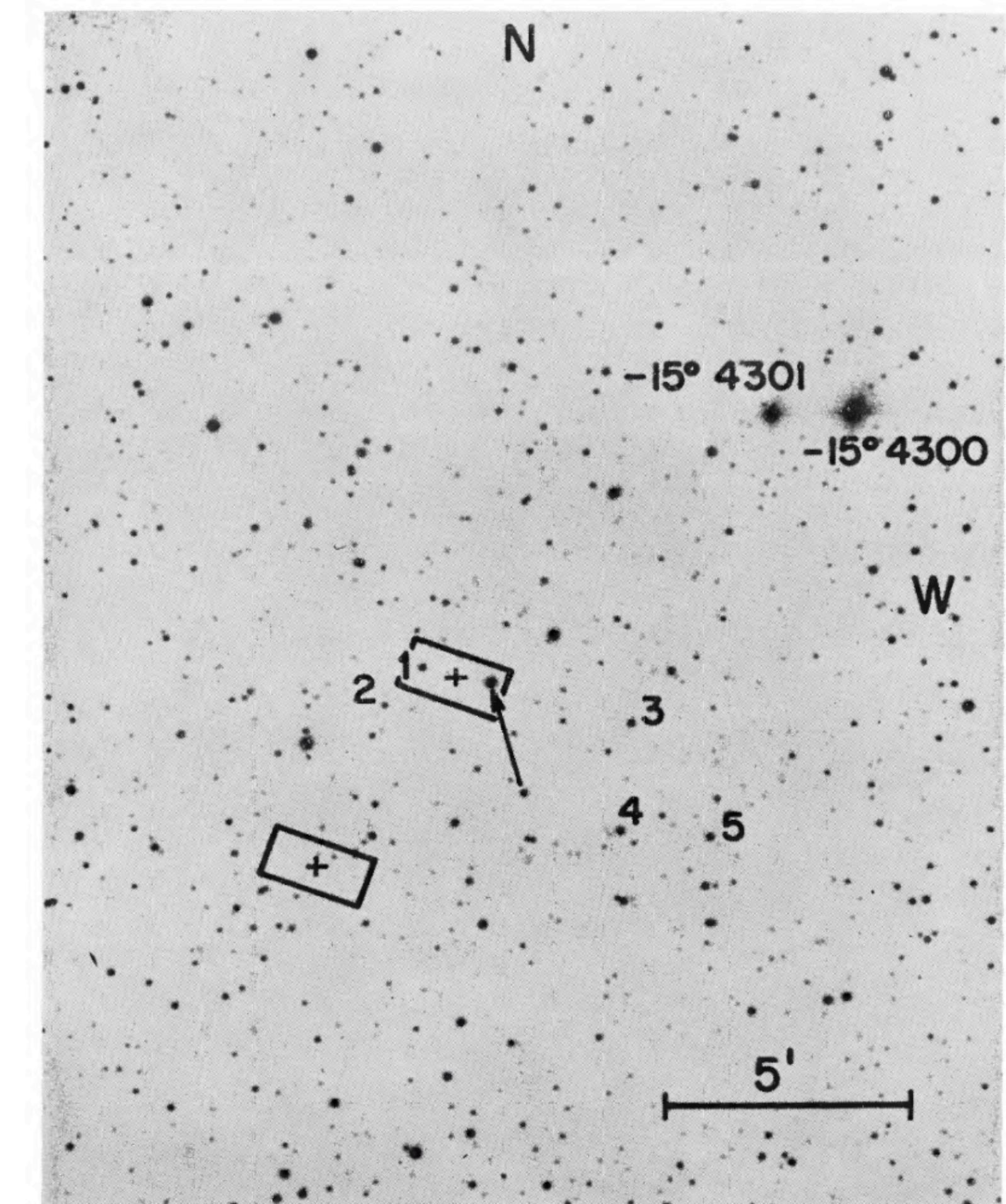
Technology Abounds

- Ancient lists of stars (Ptolemy, 150; Brahe, 1598)
- Galileo invents the telescope (1610)
- Greenwich Observatory catalogues (e.g. Bradley, 1798)
- Astrophotography invented (Bond & Whipple, 1850)
- Harvard Observatory surveys (8th magnitude, 1882-1886)
- Astrographic Chart (11th magnitude; 1887-1962)
- Carte Du Ciel (14th magnitude; 1880s-never finished)
- Invention of the CCD (Boyle & Smith, 1970)
- InfraRed detector invented (Forrest et al. (1985))
- 4- and 5-m class telescopes (1970s-1980s; e.g. LAT, MMT, UKIRT, CFHT, WHT)
- Space Telescopes (1980s-2010s; e.g. IRAS, ISO, AKARI, WISE, Spitzer)
- All-sky ground-based surveys (e.g. 2MASS, 1997-2001; SDSS, 2000-; Pan-STARRS, 2010-).

X-ray Detections: Hunting for Sco X-1



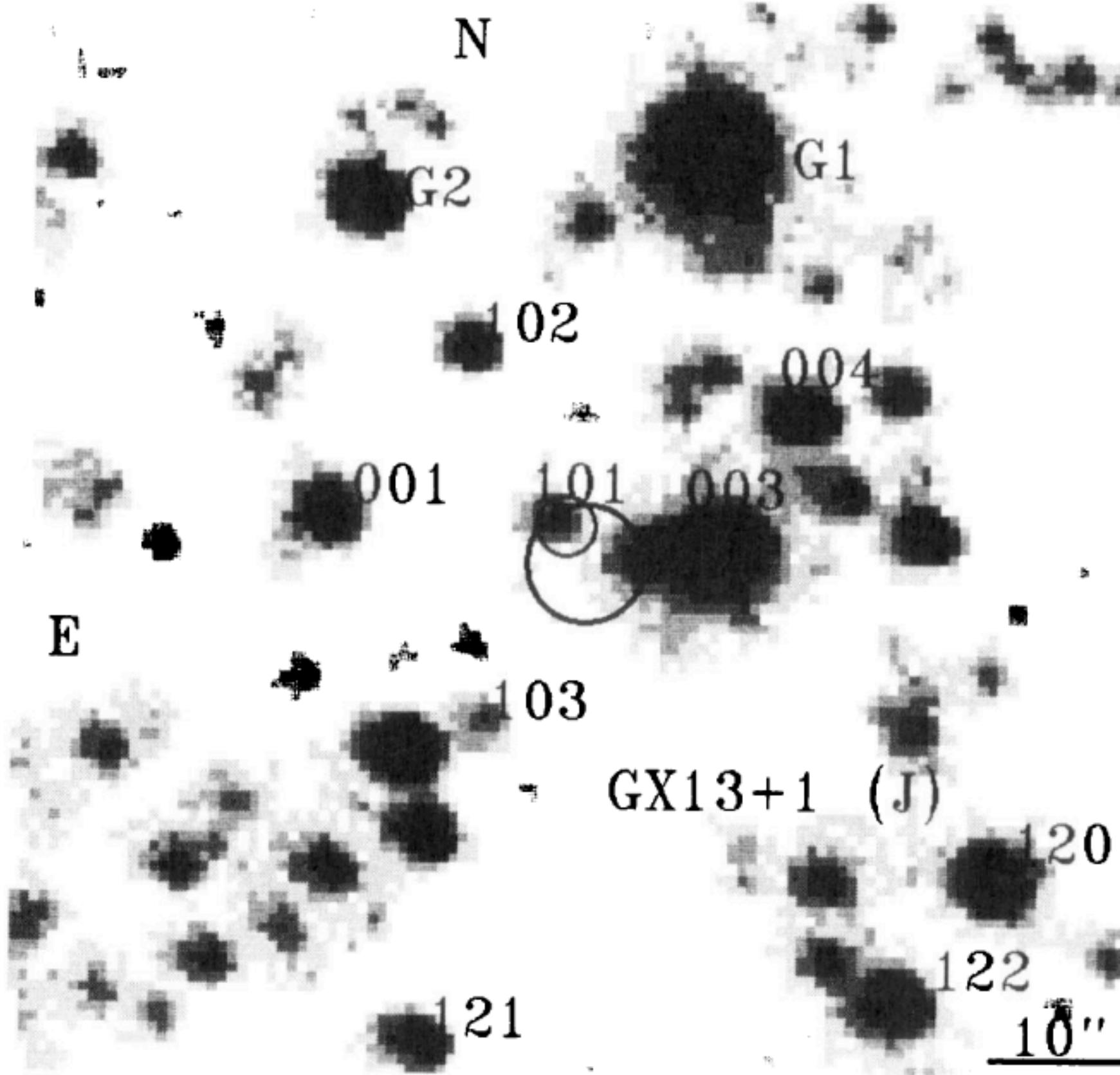
Giacconi, Gursky, & Waters (1964)



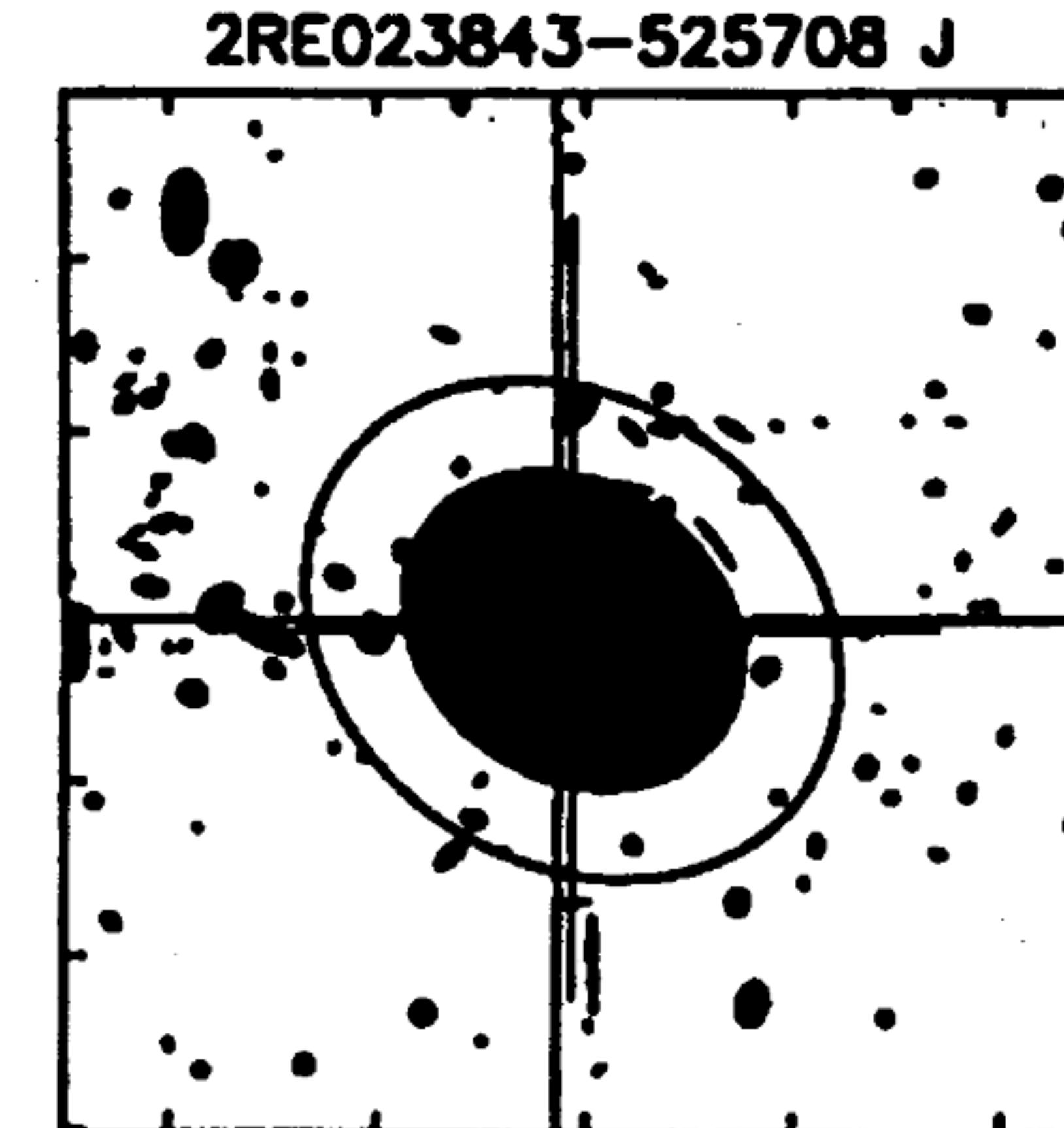
Sandage et al. (1966)

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The Brightest Star in the Sky



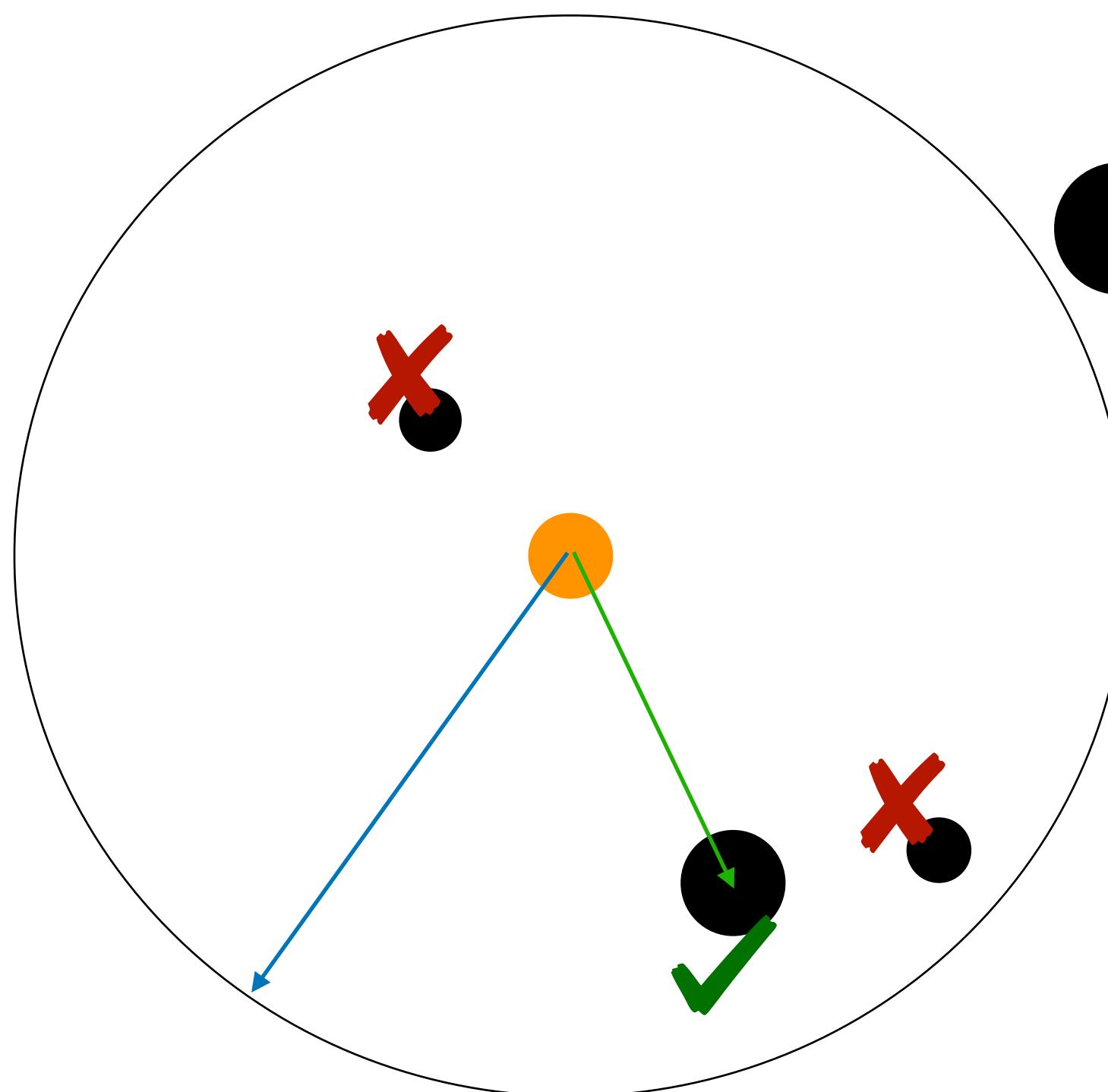
Naylor, Charles, & Longmore (1991)



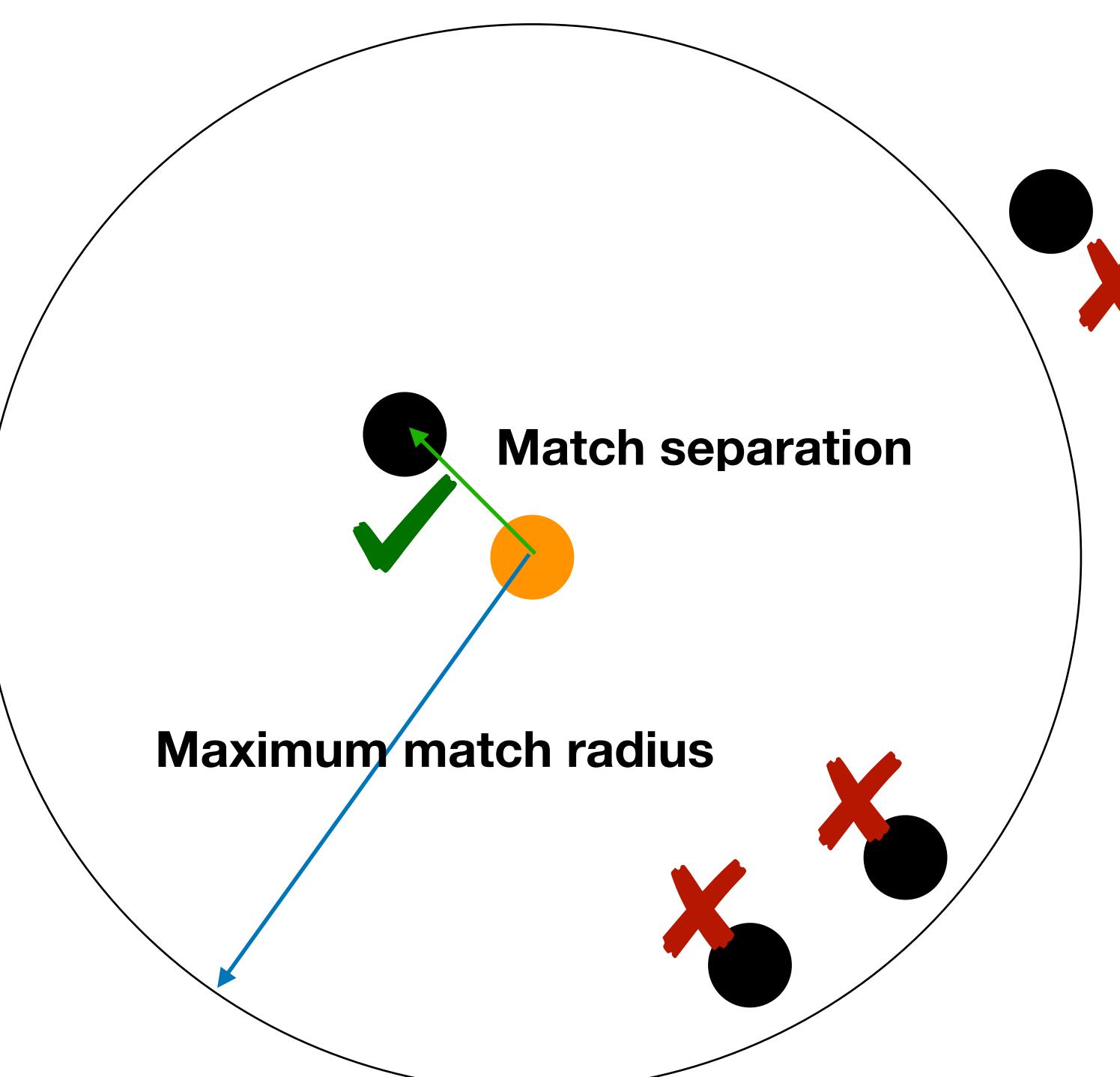
“...X-ray sources are rare events; bright optical sources are also rare events, so the observation of an X-ray source and a bright optical source in the same region of the sky is considered a non-random event”

Fotopoulou et al. (2016)

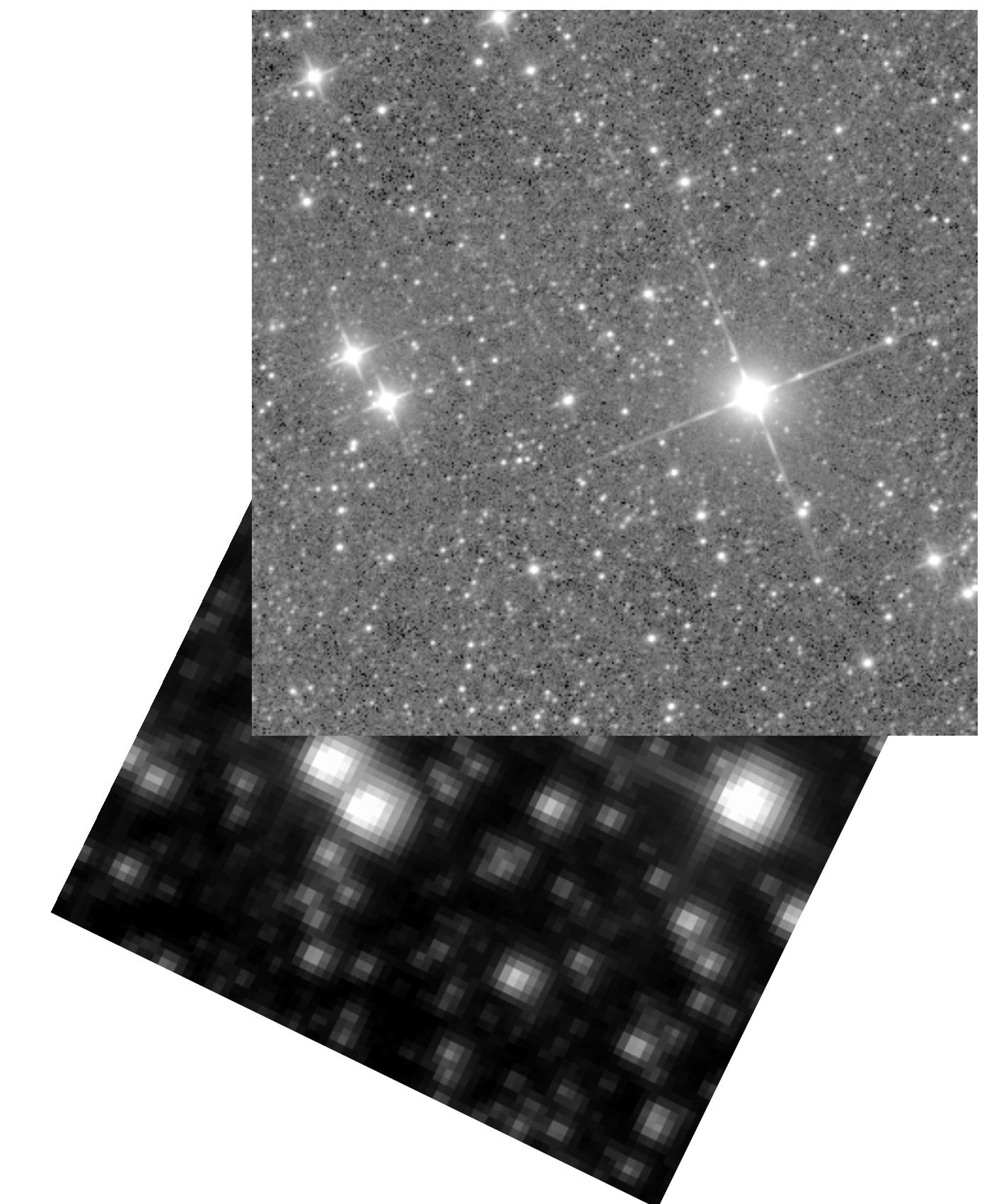
“Traditional” Cross-Matching



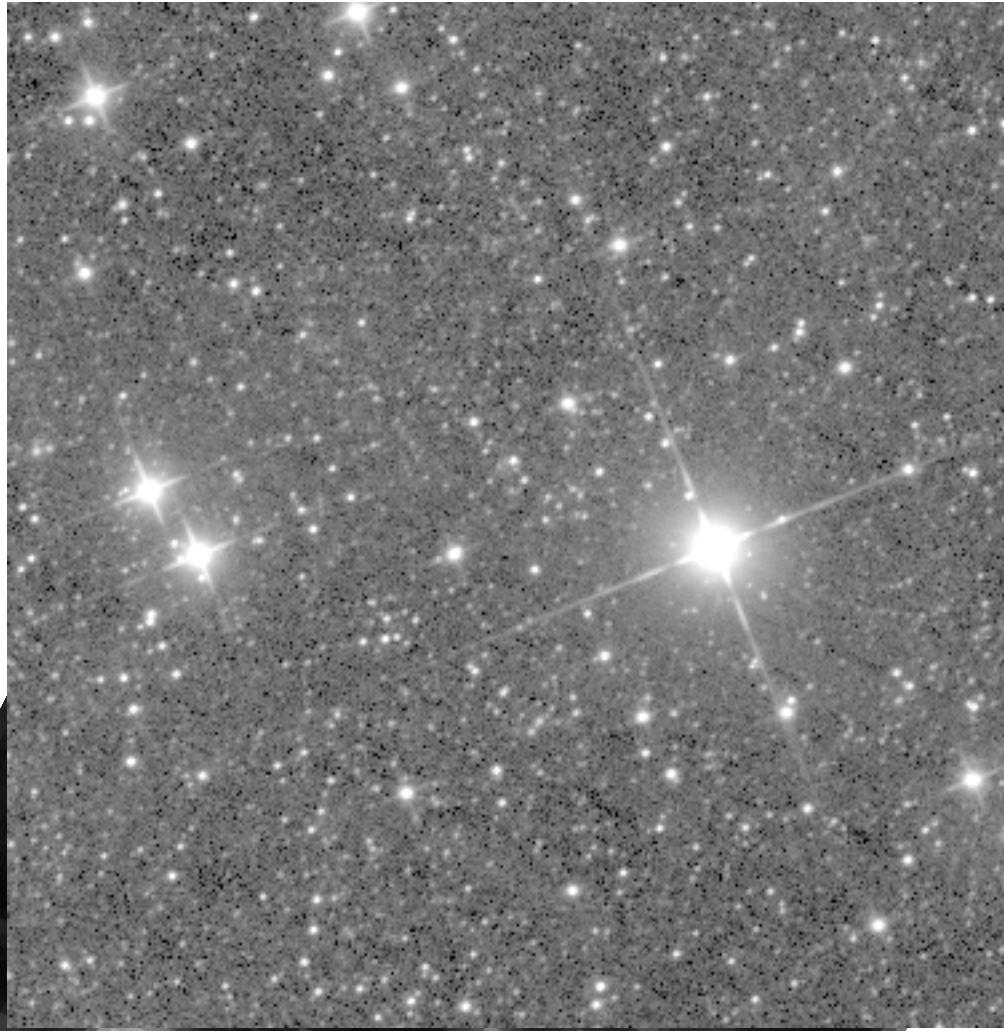
Declination / degrees



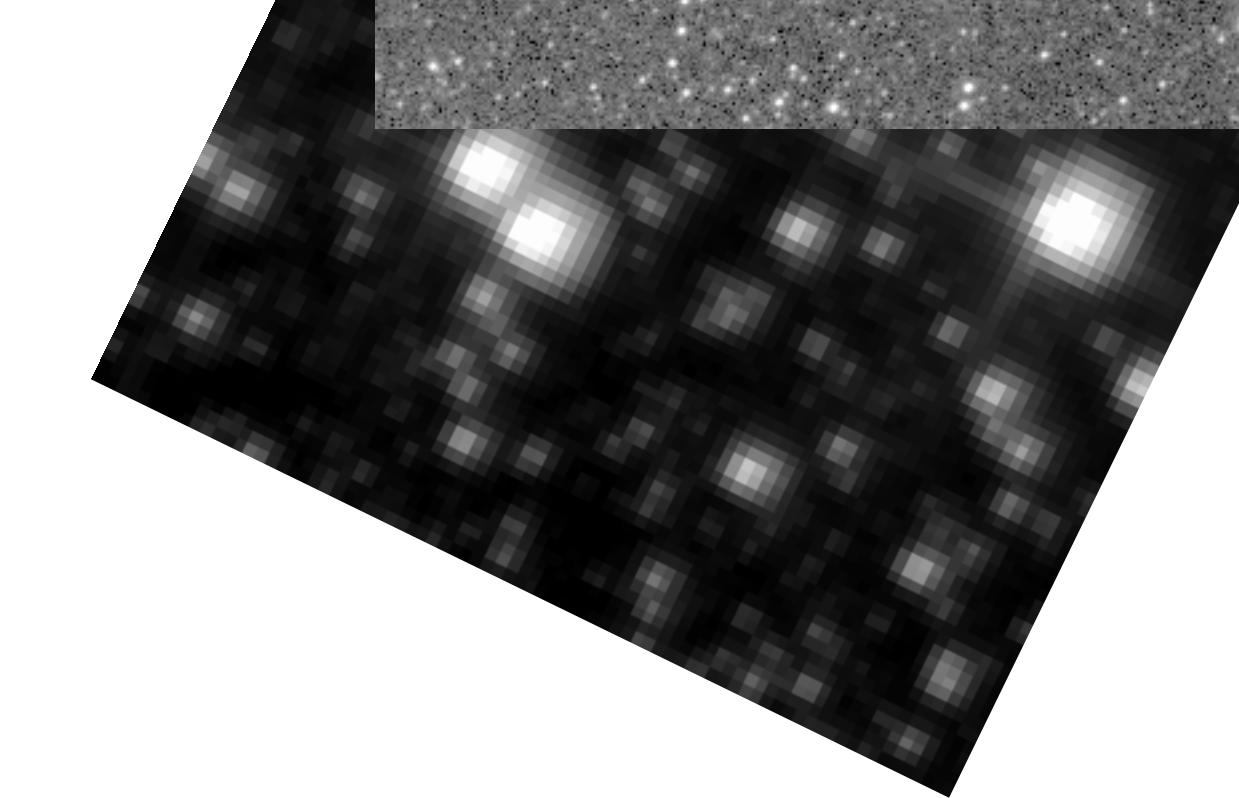
Right Ascension / degrees



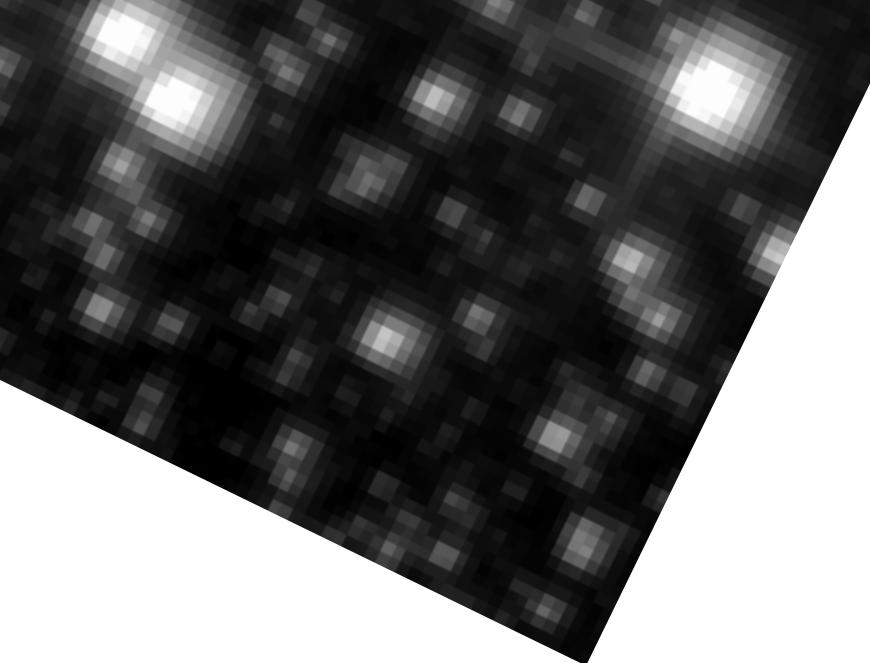
The Astronomy Error Function



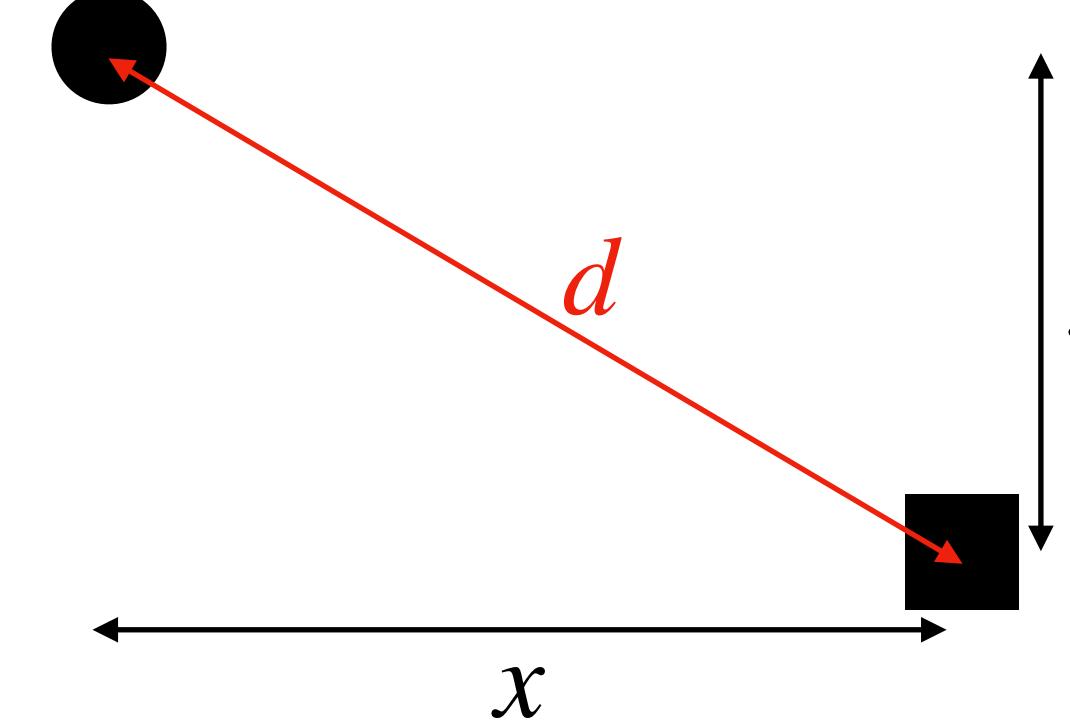
- 1) $p(x \text{ and } y) = p(x)p(y)$
- 2) $p(x)$ decreases as x increases
- 3) $p(x) = p(-x) \Rightarrow p(x^2)$



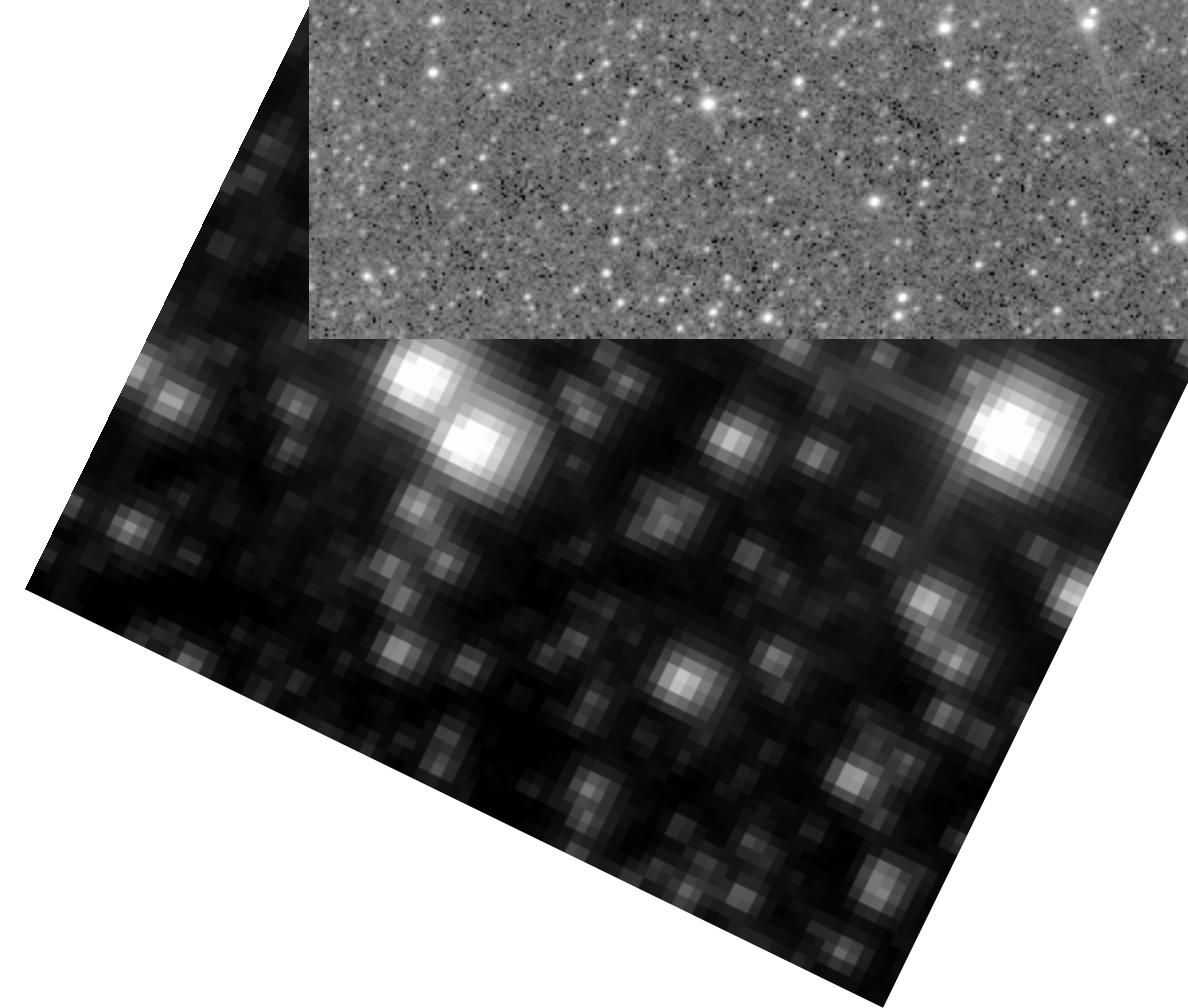
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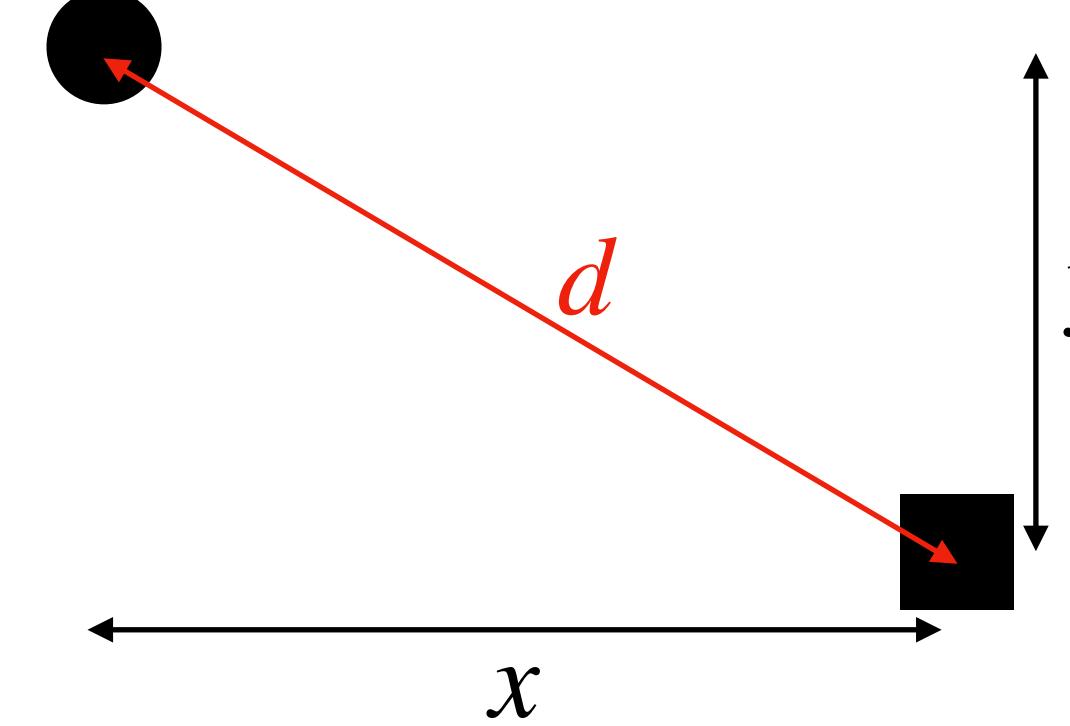
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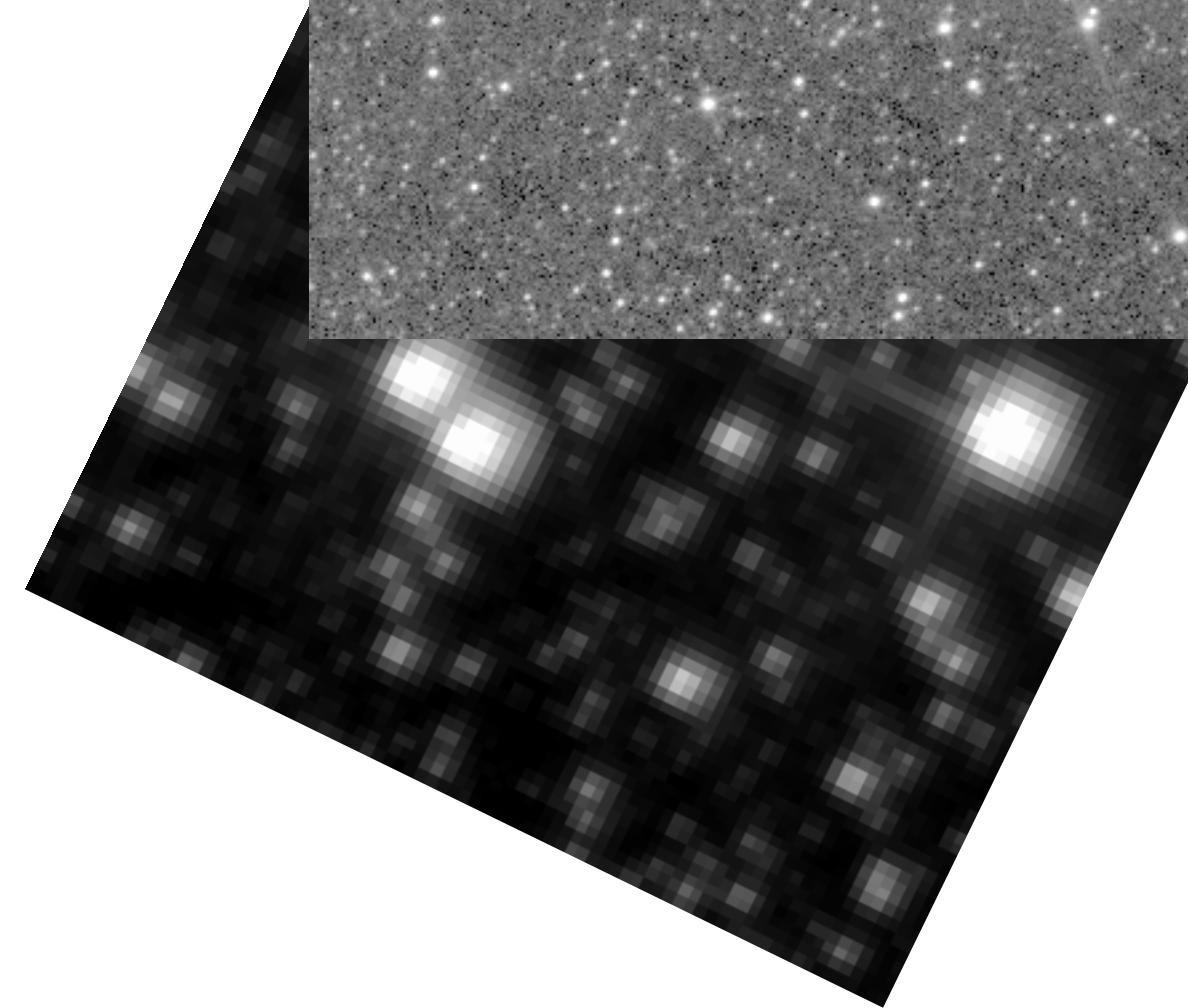


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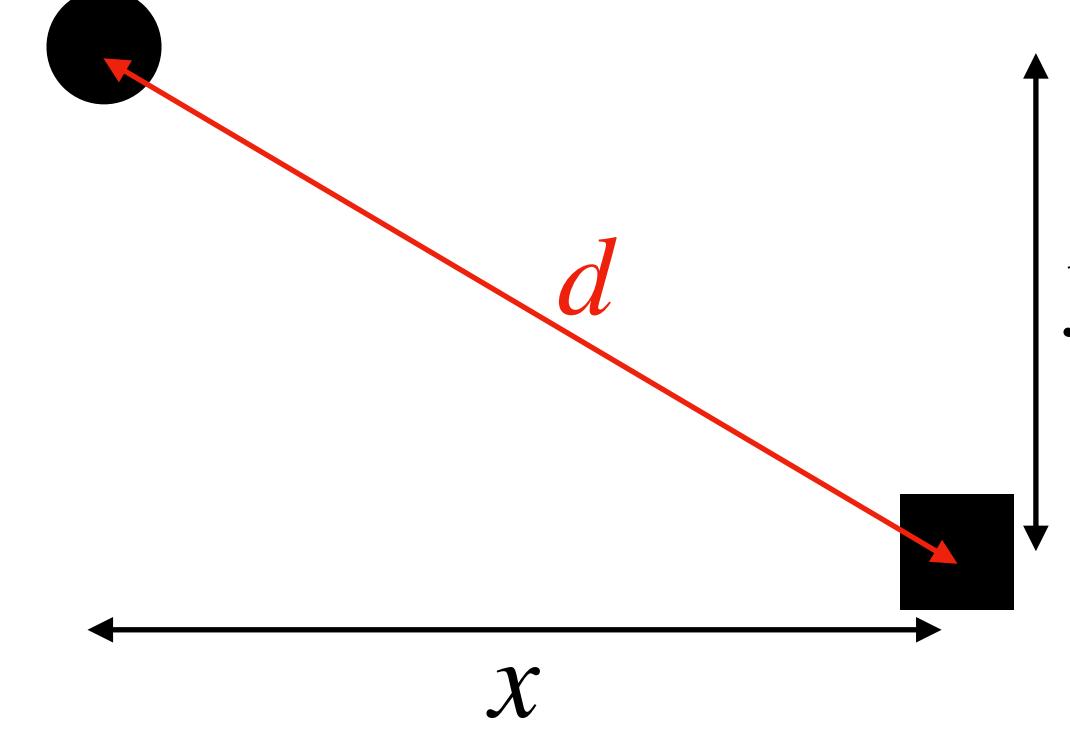


$$p(d^2) = p(x^2 + y^2) = p(x^2)p(y^2)$$

The Astronomy Error Function



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$$p(d^2) = p(x^2 + y^2) = p(x^2)p(y^2)$$

$$g(x, y, \sigma) = (2\pi\sigma^2)^{-1} \exp\left(-\frac{1}{2} \frac{x^2 + y^2}{\sigma^2}\right)$$

Probabilistic Cross-Matching

The Likelihood Ratio

$$dp(r|id) = r \times e^{-r^2/2} dr.$$

$$dp(r|c) = 2\lambda r \times e^{-\lambda r^2} dr$$

$$LR(r) = dp(r|id)/dp(r|c) = \frac{1}{2\lambda} \exp\left\{\frac{r^2}{2}(2\lambda - 1)\right\}$$

de Ruiter, Willis, & Arp (1977)

$$dp_{id} = Qr \exp\left(\frac{-r^2}{2}\right) dr. \quad dp_{uo} = 2\lambda r dr$$

$$LR(r) = \frac{dp_{id}}{dp_{uo}} = \frac{Q \exp(-r^2/2)}{2\lambda}$$

Wolstencroft et al. (1986)

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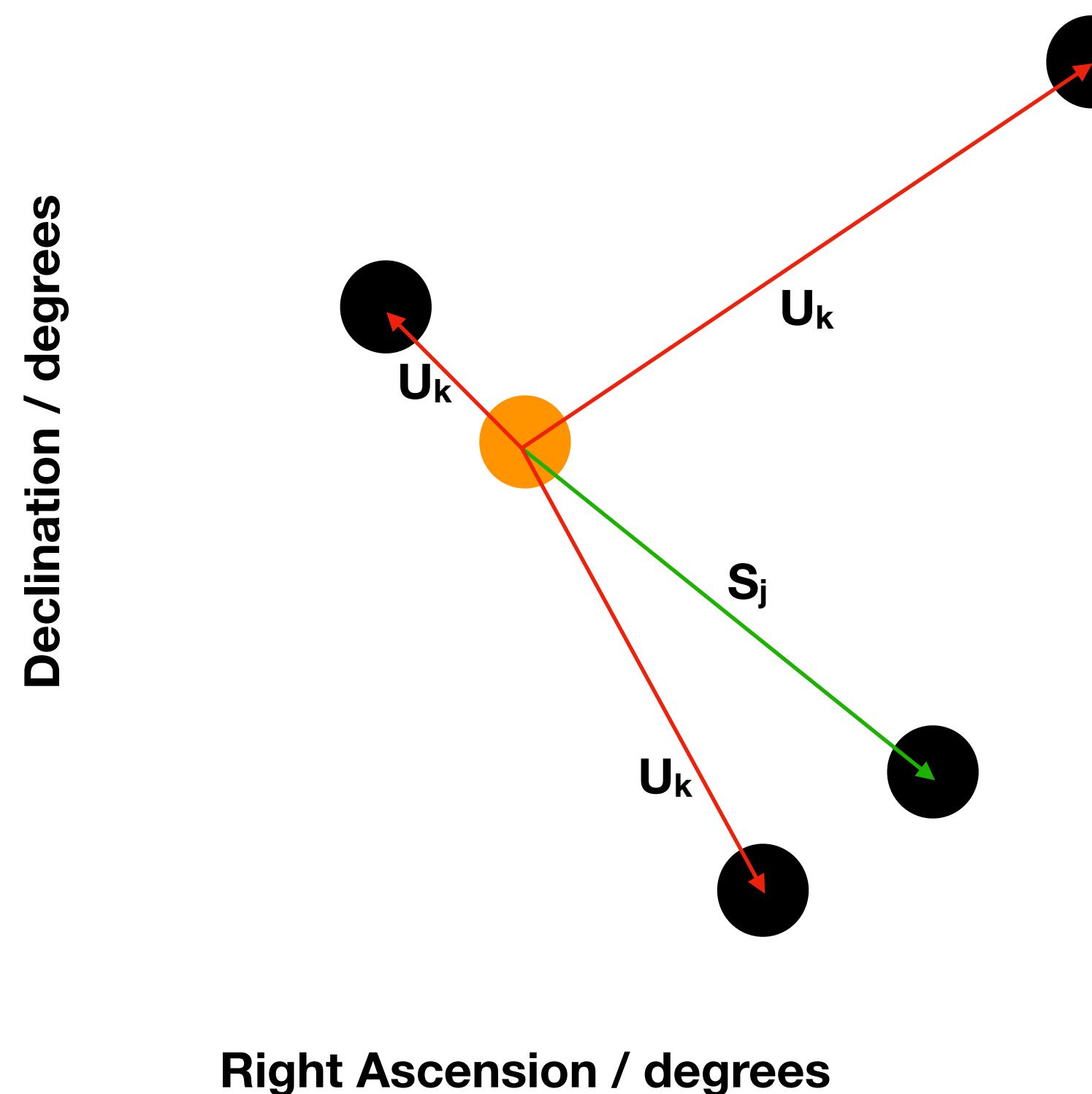
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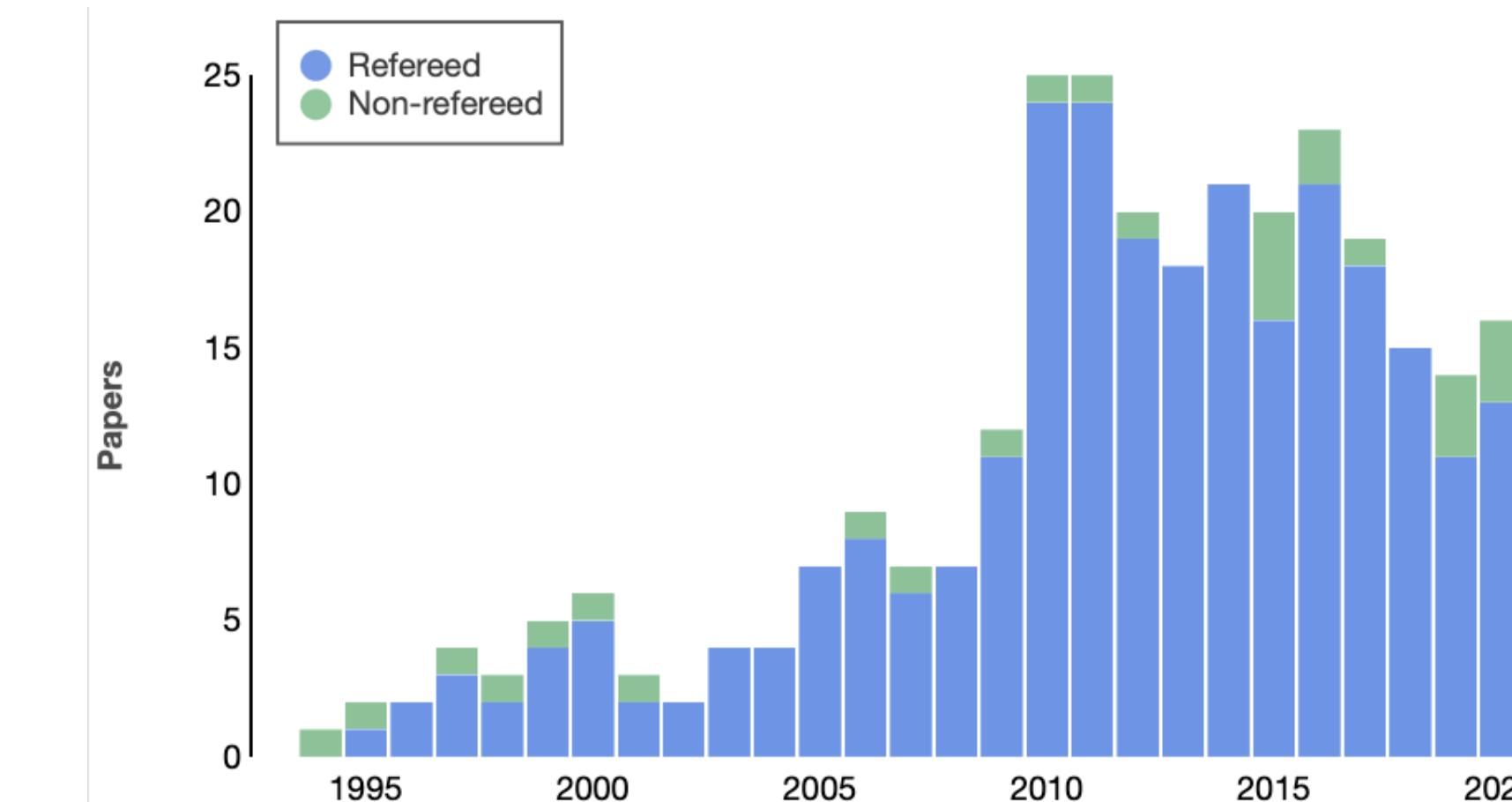
Wolstencroft et al. (1986)

The “Reliability” – Sutherland & Saunders (1992)



$$R_j = \frac{\Pr\left[S_j \cap \left(\bigcap_{k \neq j} U_k\right) \cap \left(\bigcap_{k'} E_{k'}\right)\right]}{\sum_i \Pr\left[S_i \cap \left(\bigcap_{k \neq i} U_k\right) \cap \left(\bigcap_{k'} E_{k'}\right)\right] + \Pr\left[(m_s > m_{lim}) \cap \left(\bigcap_k U_k\right) \cap \left(\bigcap_{k'} E_{k'}\right)\right]} = \frac{L_j}{\sum_i L_i + (1 - Q)}$$

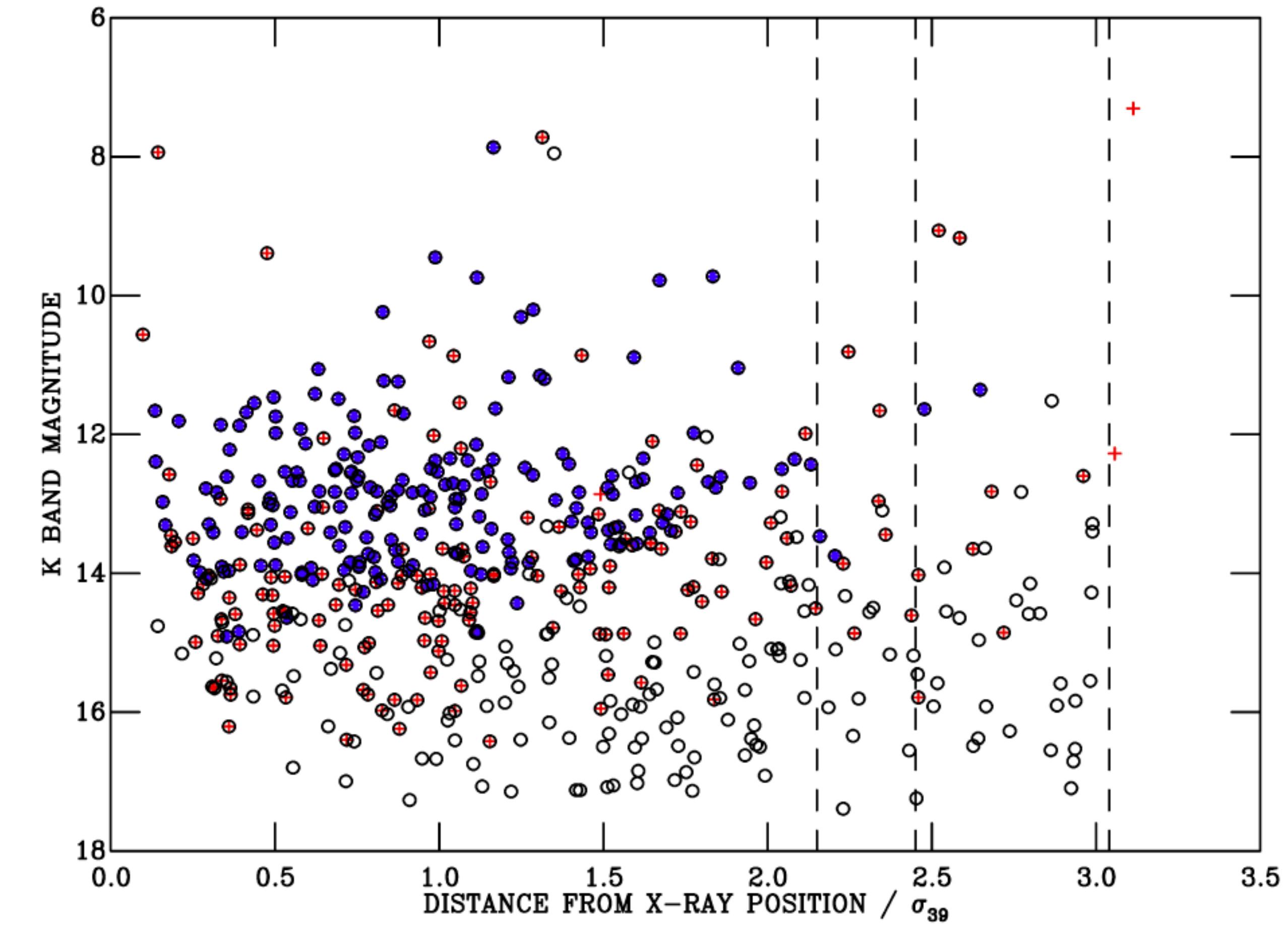
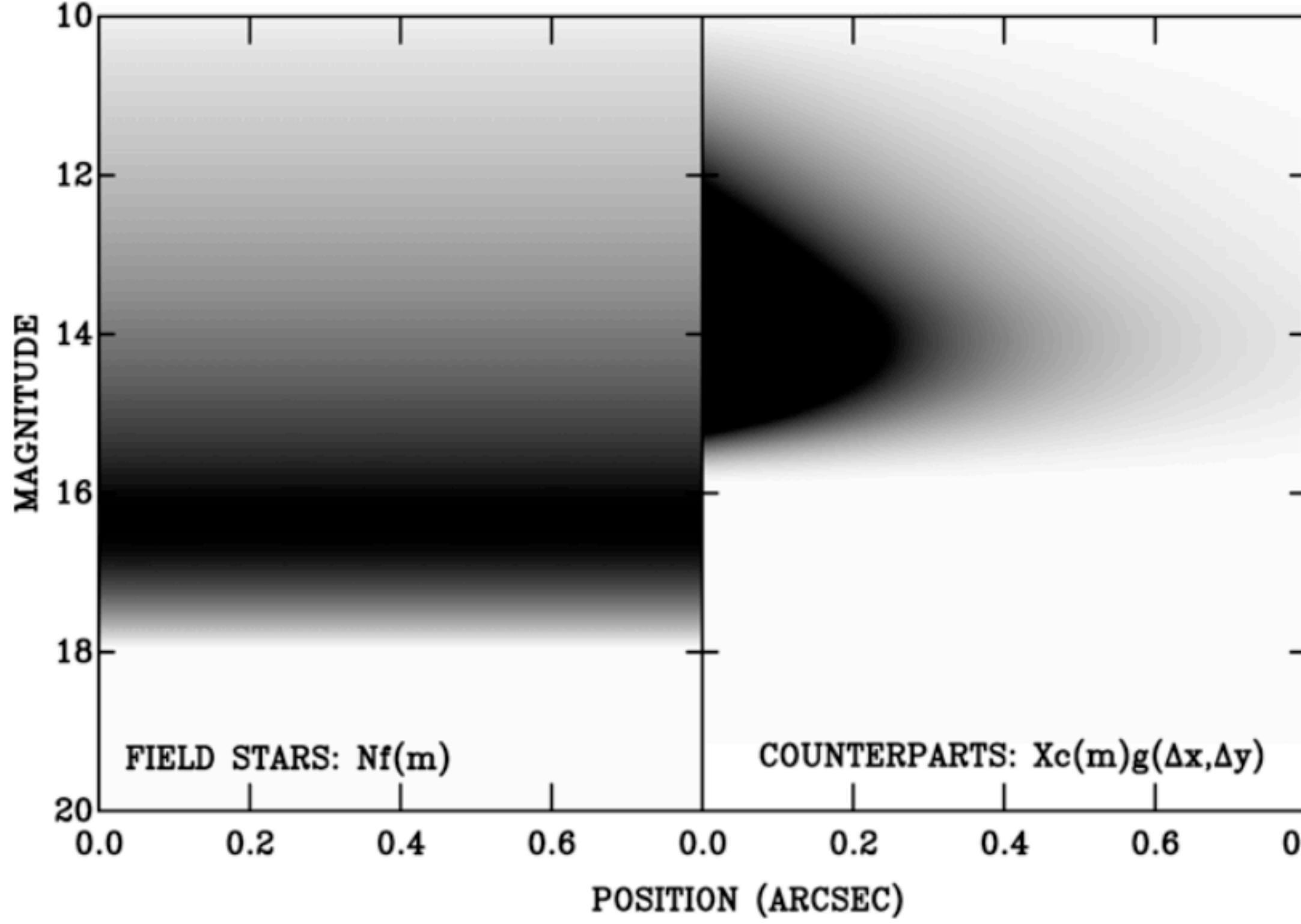
$$L = \frac{q(m, c) f(x, y)}{n(m, c)}$$



Probabilistic Cross-Matching

$$P(0) = \frac{1 - X}{1 - X + \sum_j \frac{Xc(m_j) g(\Delta x_j, \Delta y_j)}{Nf(m_j)}}$$

$$P(i) = \frac{\frac{Xc(m_i) g(\Delta x_i, \Delta y_i)}{Nf(m_i)}}{1 - X + \sum_j \frac{Xc(m_j) g(\Delta x_j, \Delta y_j)}{Nf(m_j)}}$$



Probabilistic Cross-Matching

$$p(D|H) = \int p(\mathbf{m}|H) \prod_{i=1}^n p_i(\mathbf{x}_i|\mathbf{m}, H) d^3\mathbf{m}$$

$$p(D|K) = \prod_{i=1}^n \left[\int p(\mathbf{m}_i|K) p_i(\mathbf{x}_i|\mathbf{m}_i, K) d^3\mathbf{m}_i \right]$$

$$B(H, K|D') = \frac{\int p(\boldsymbol{\eta}|H) \prod_{i=1}^n p_i(\mathbf{g}_i|\boldsymbol{\eta}, H) d^r\boldsymbol{\eta}}{\prod_{i=1}^n \left[\int p(\boldsymbol{\eta}_i|K) p_i(\mathbf{g}_i|\boldsymbol{\eta}_i, K) d^r\boldsymbol{\eta}_i \right]}$$

Budavári & Szalay (2008)

Includes SED model fitting to all sources

Probabilistic Cross-Matching

Nearest neighbour or brightest neighbour: one-to-one, either astrometry OR photometry

Likelihood ratio: one-to-one matches, mostly just astrometry (e.g., Wolstencroft et al. 1986)

Reliability: One-to-many matches, uses photometry from one dataset (e.g. Naylor et al. 2013)

Budavári & Szalay (2008): one-to-one-to-one-to... matches, include SED fitting

e.g. Pineau et al. (2017): many-to-many-to-many-to... matches, no photometry implemented

Probabilistic Cross-Matching

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e.g. Pineau et al. (2017): many-to-many-to-many-to... matches, no photometry implemented

One assumption made in all of these works: positional errors of sources are Gaussian!

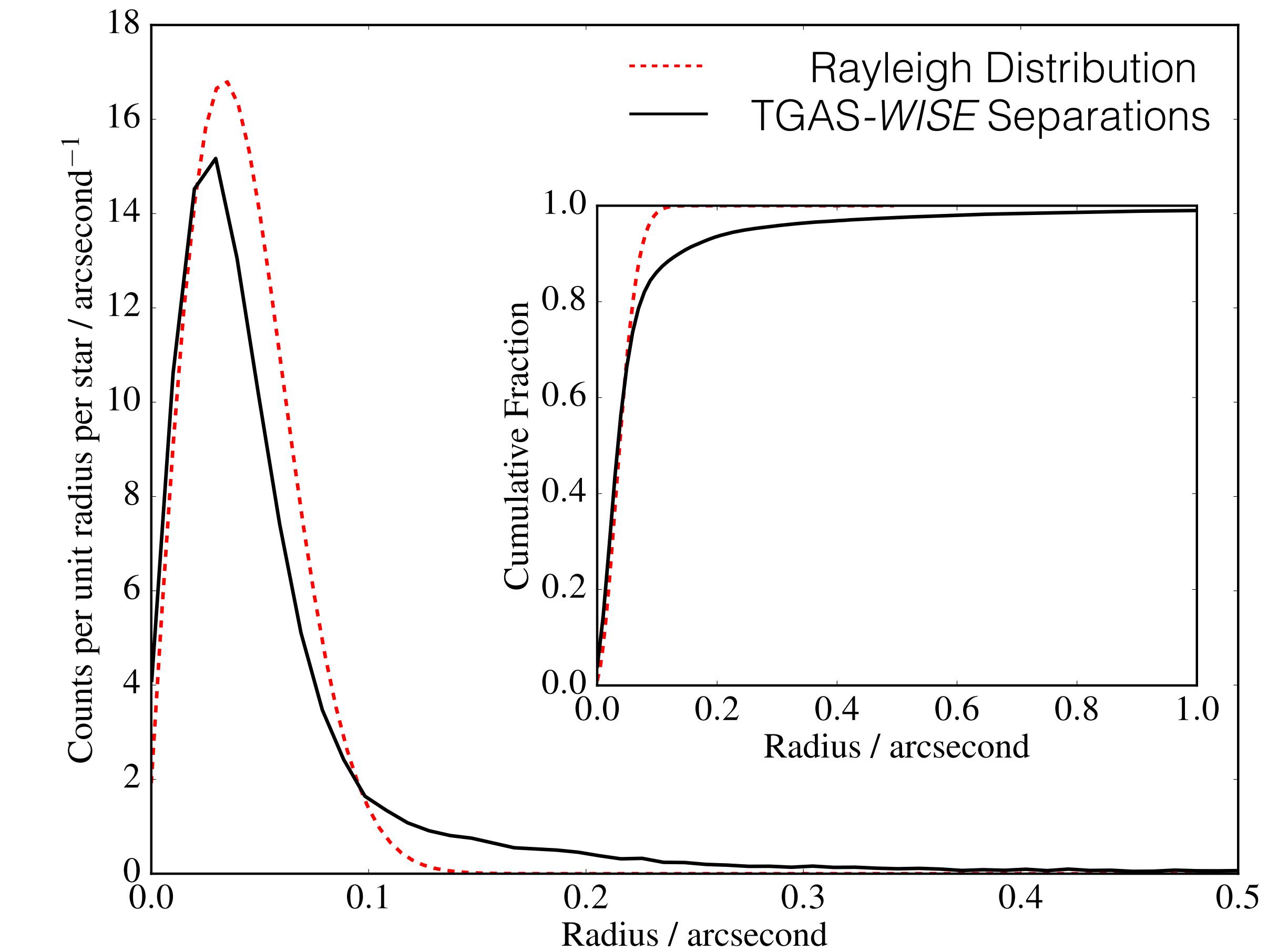
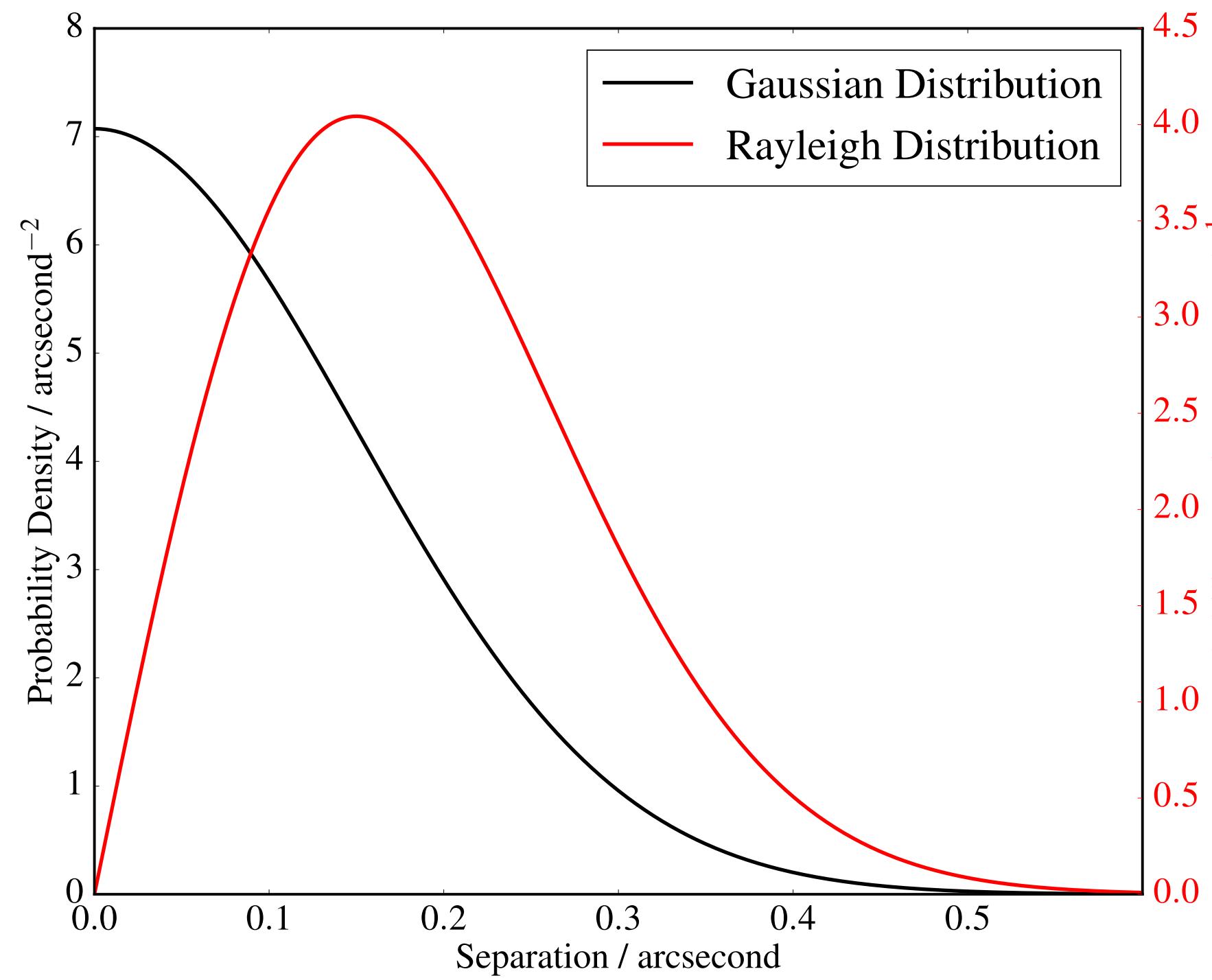
$$dp(r|id) = r \times e^{-r^2/2} dr. \quad P(i) = \frac{\frac{Xc(m_i) g(\Delta x_i, \Delta y_i)}{Nf(m_i)}}{1 - X + \sum_j \frac{Xc(m_j) g(\Delta x_j, \Delta y_j)}{Nf(m_j)}} \quad p(D|H) = \int p(m|H) \prod_{i=1}^n p_i(x_i|m, H) d^3m$$

The Astrometric Uncertainty Function

$$g(x, y, \sigma) = (2\pi\sigma^2)^{-1} \exp\left(-\frac{1}{2}\frac{x^2 + y^2}{\sigma^2}\right)$$



$$g(r, \sigma) = \frac{r}{\sigma^2} \exp\left(-\frac{1}{2}\frac{r^2}{\sigma^2}\right)$$



Wilson & Naylor (2017)
WISE - Wright et al. (2010)

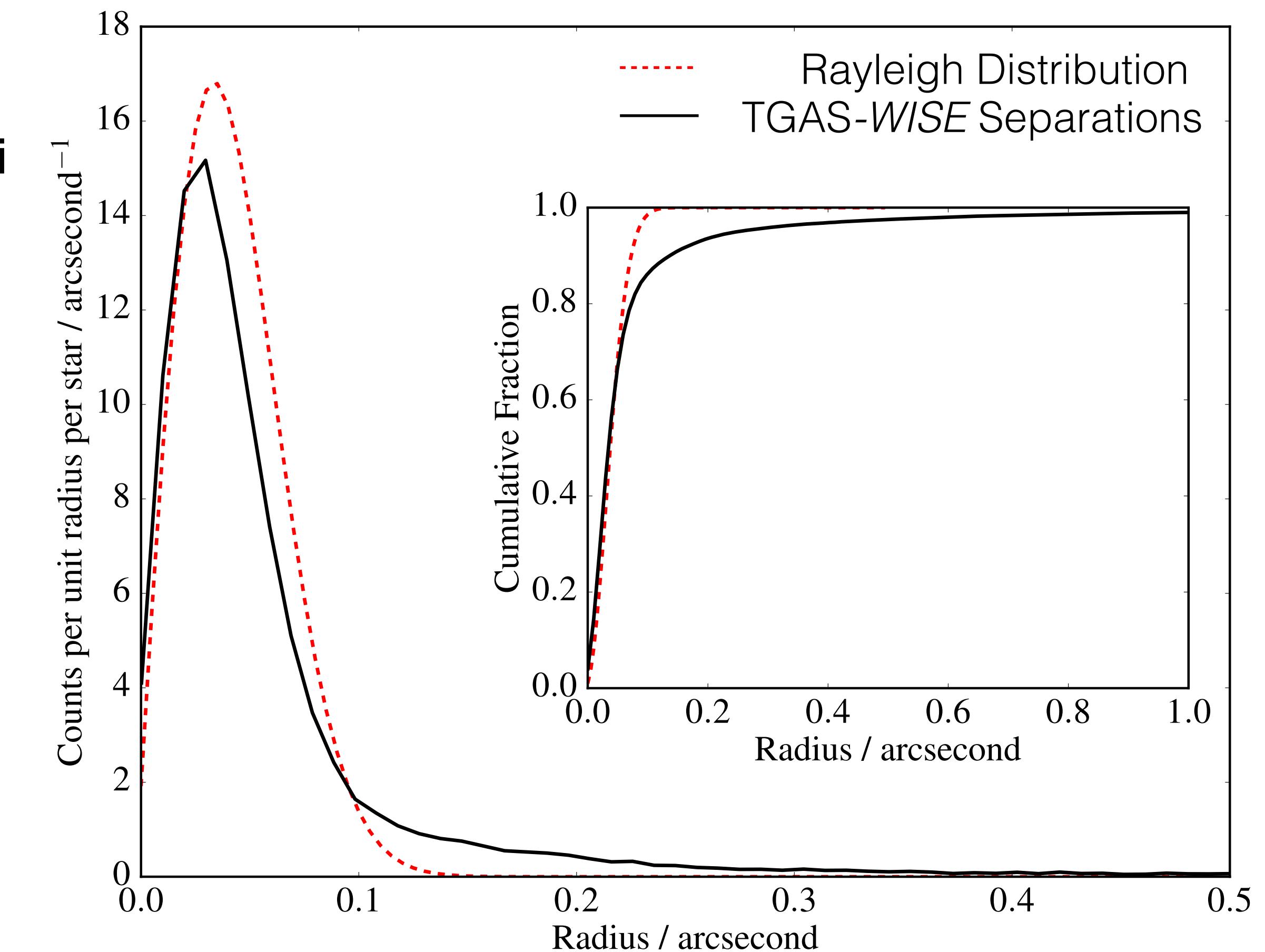
TGAS - Michalik, Lindegren, & Hobbs (2015)
Gaia - Gaia Collaboration, Brown A. G. A., et al. (2016)

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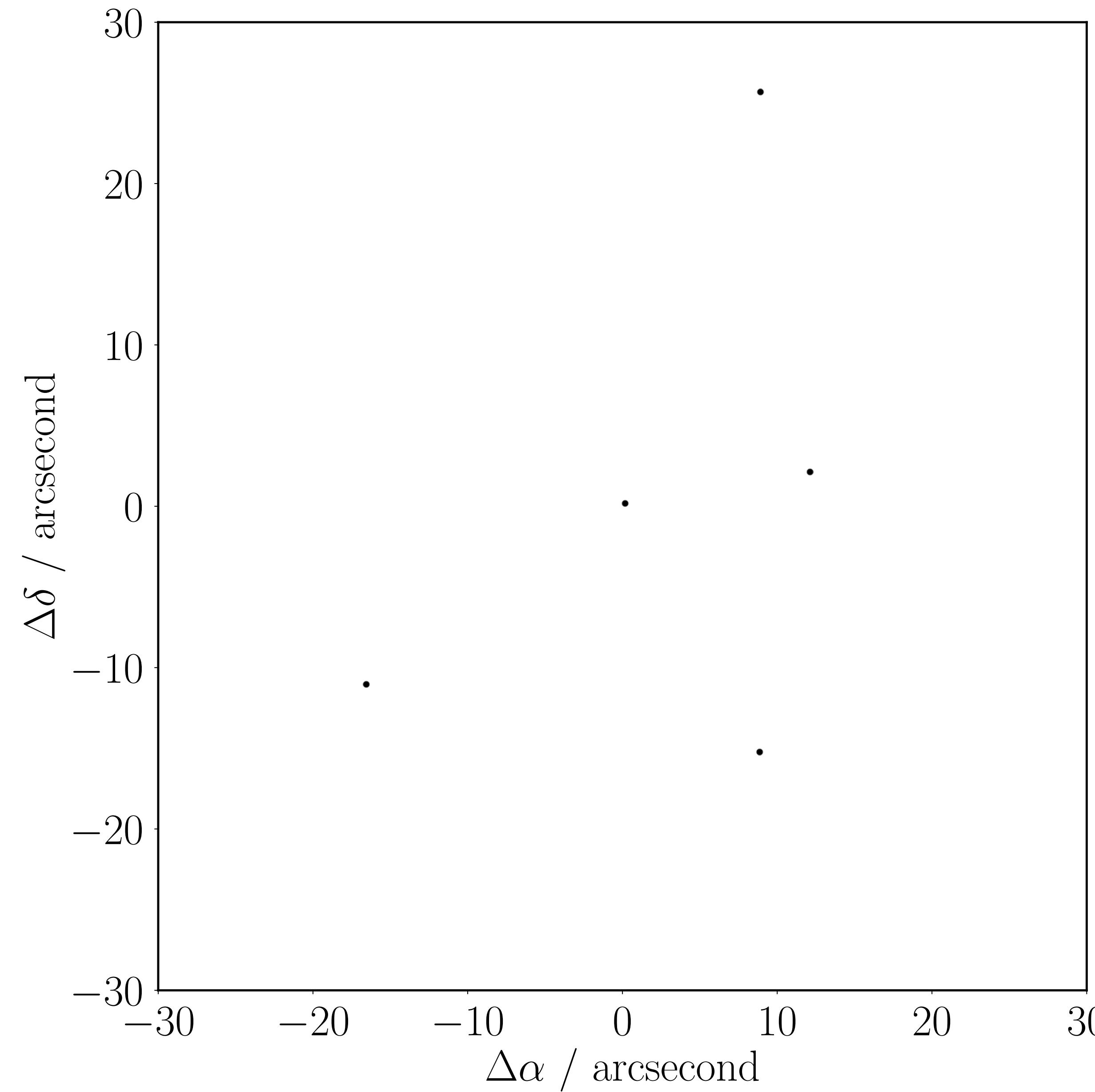
The Astrometric Uncertainty Function

Reasons for large separations:

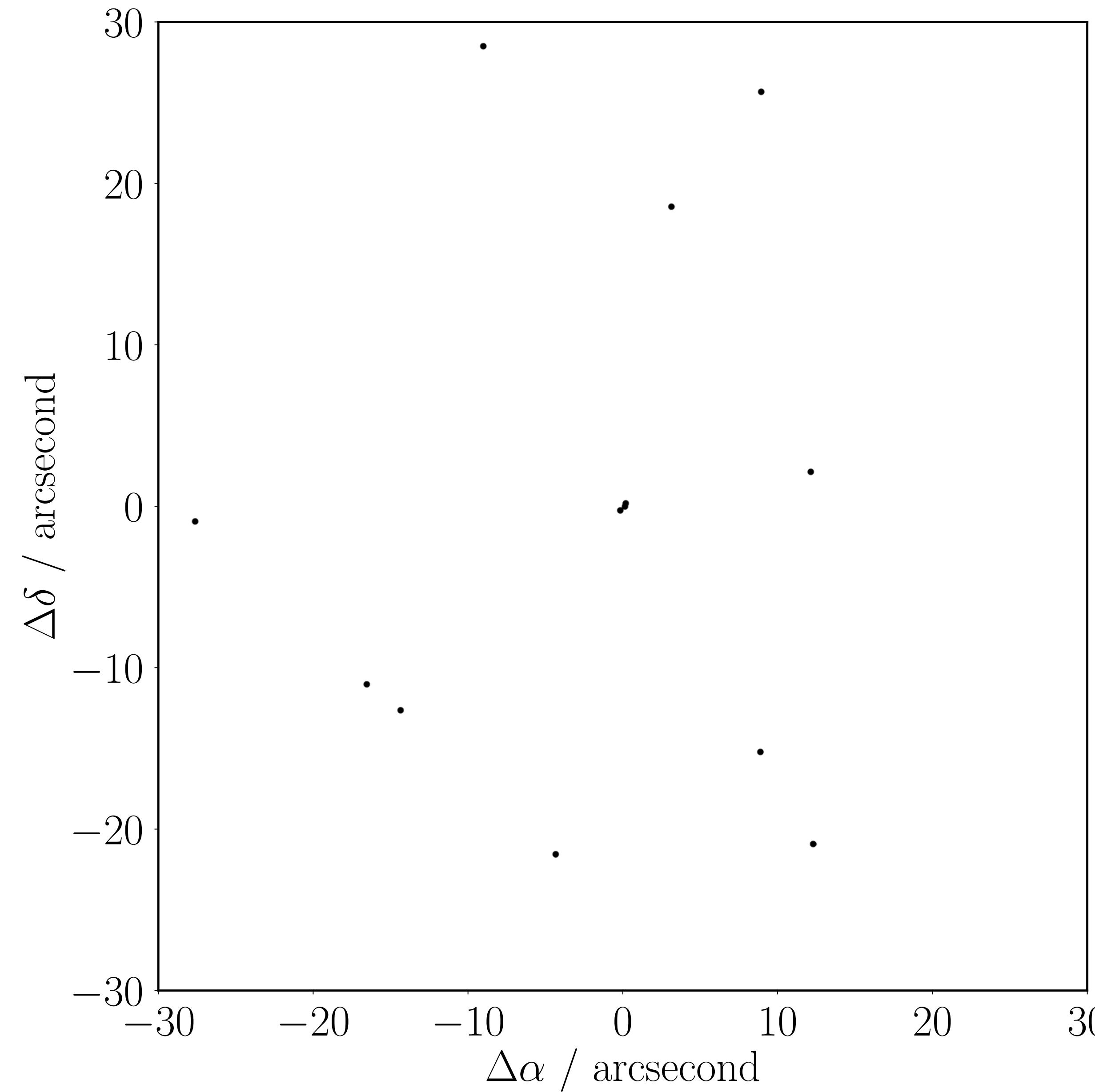
- 1) proper motions (e.g. AllWISE Supplement 6.4, Cutri et al. 2012) – no, TGAS provided for all sources
- 2) false matches – no, 0.1% chance of random match within 0.5 arcseconds
- 3) What else could it be?



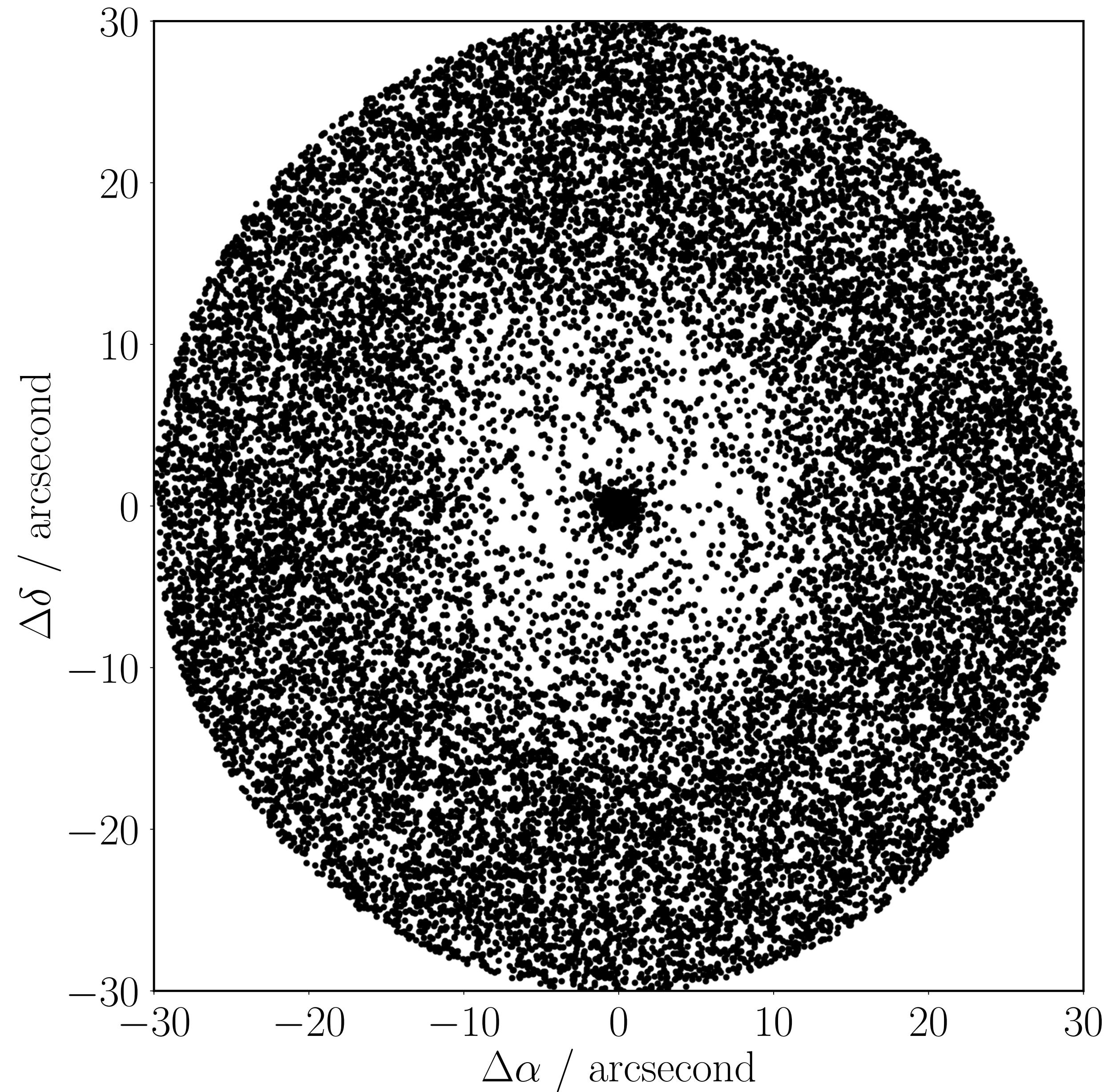
The AUF: Crowding



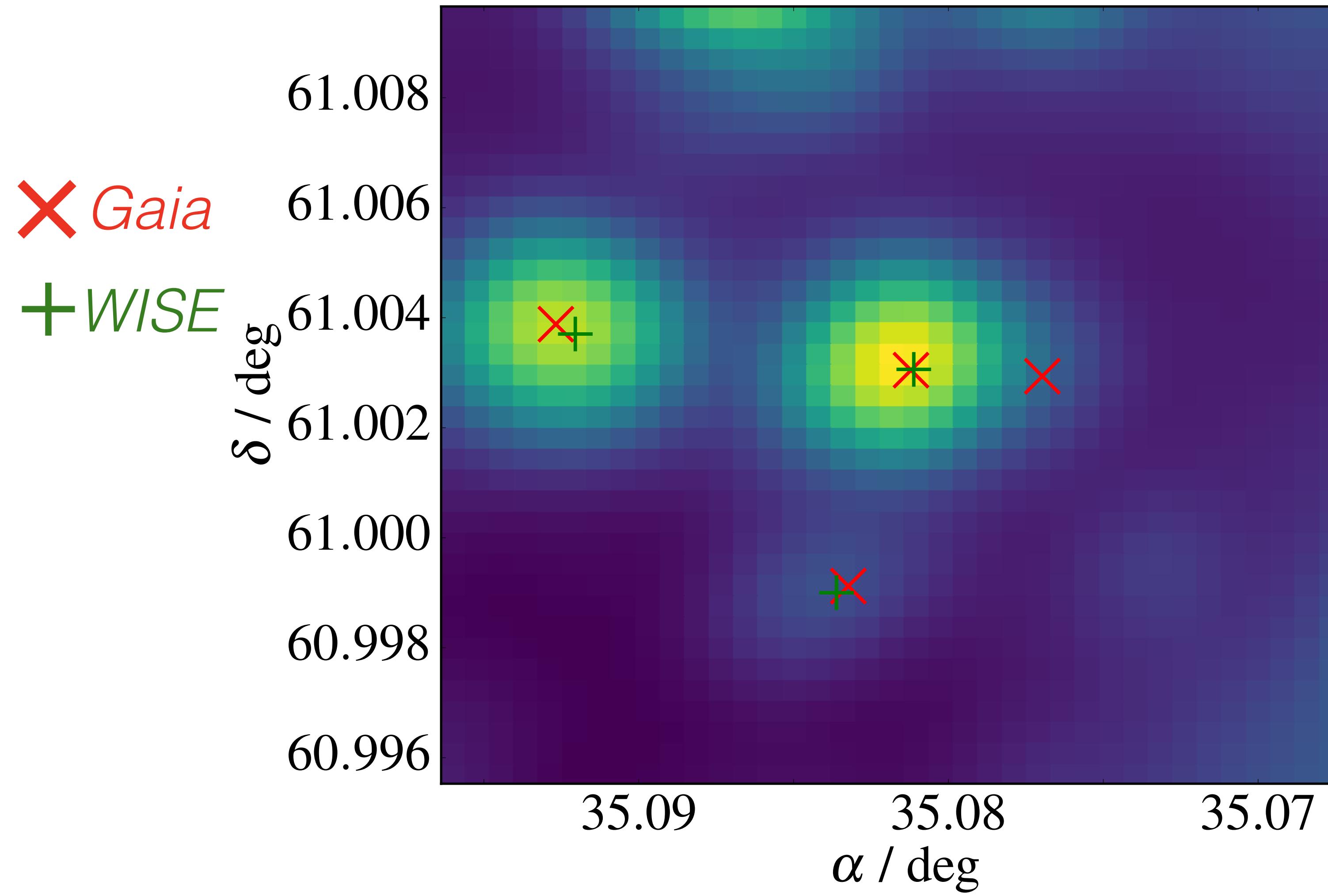
The AUF: Crowding



The AUF: Crowding



Resolving *Gaia*-*WISE* Blends



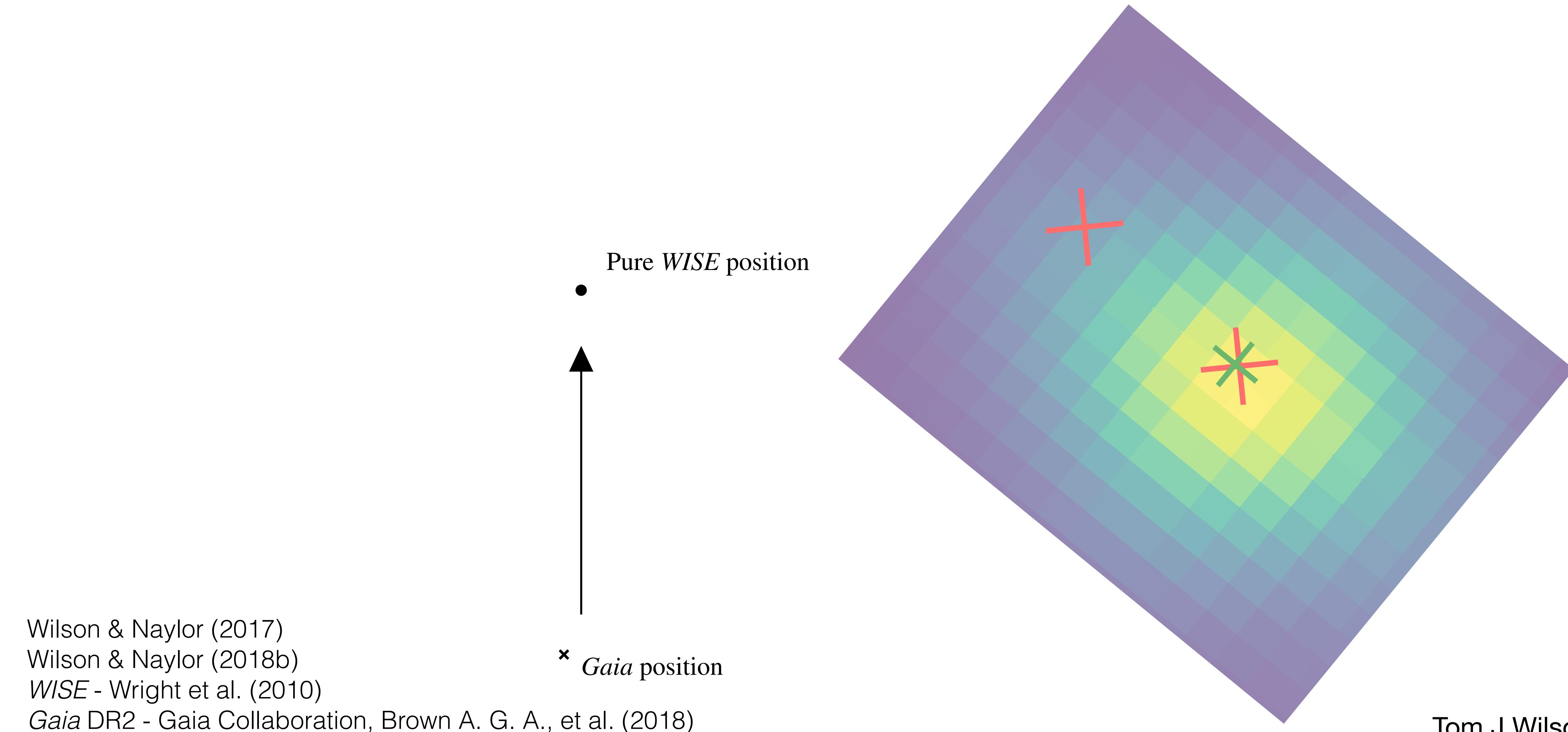
Wilson & Naylor (2018b)

WISE - Wright et al. (2010)

Gaia DR2 - Gaia Collaboration, Brown A. G. A., et al. (2018)

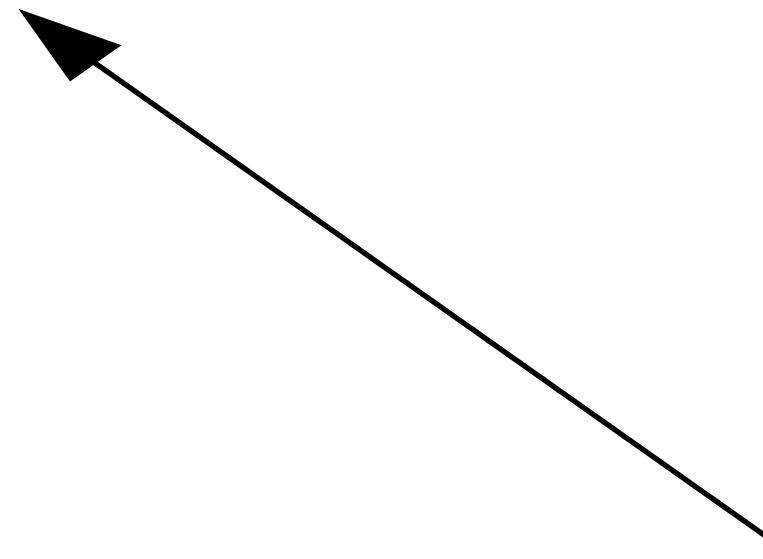
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The AUF: Perturbation



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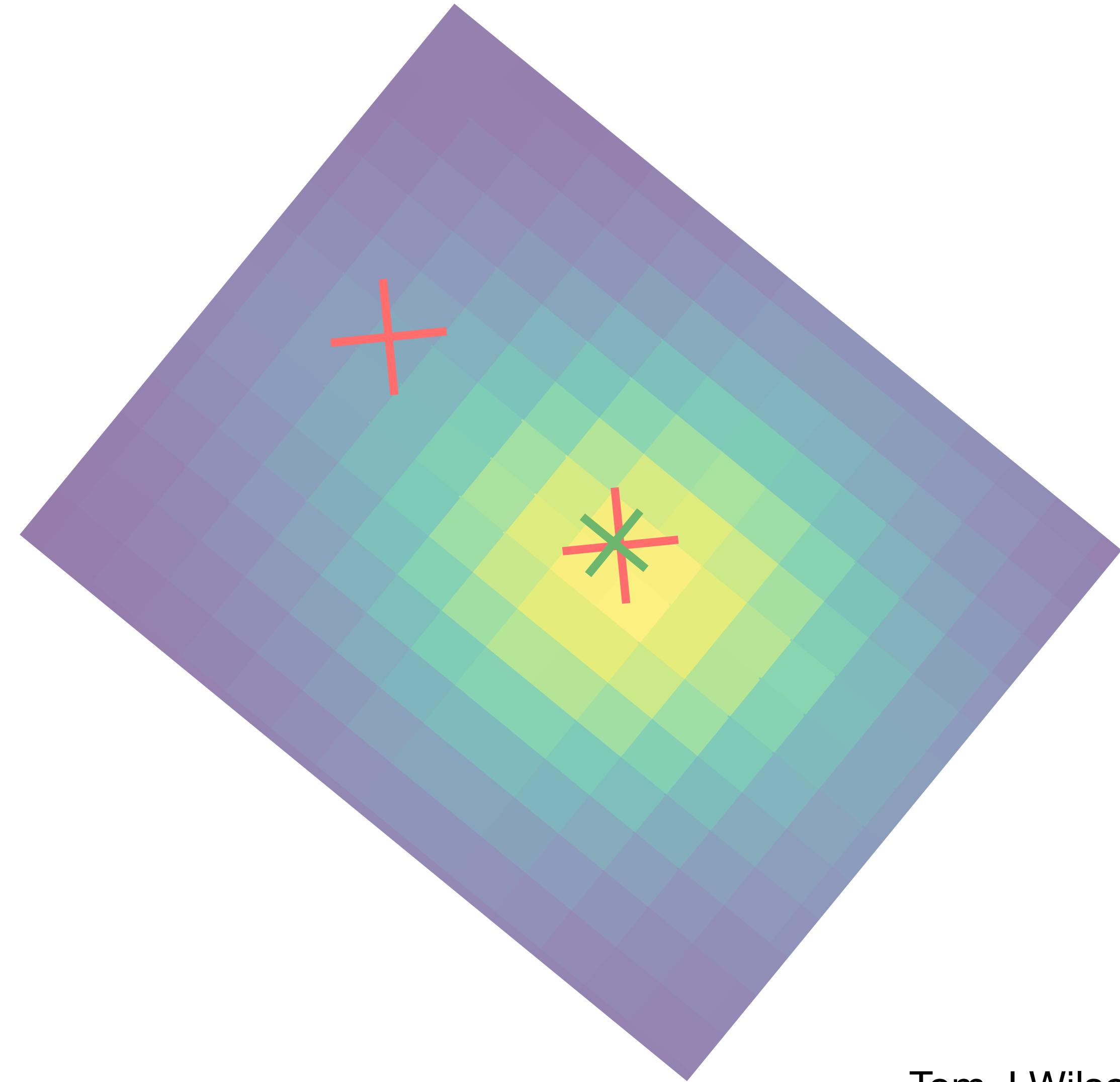
To *WISE* contaminant



Pure *WISE* position



× *Gaia* position



Wilson & Naylor (2017)

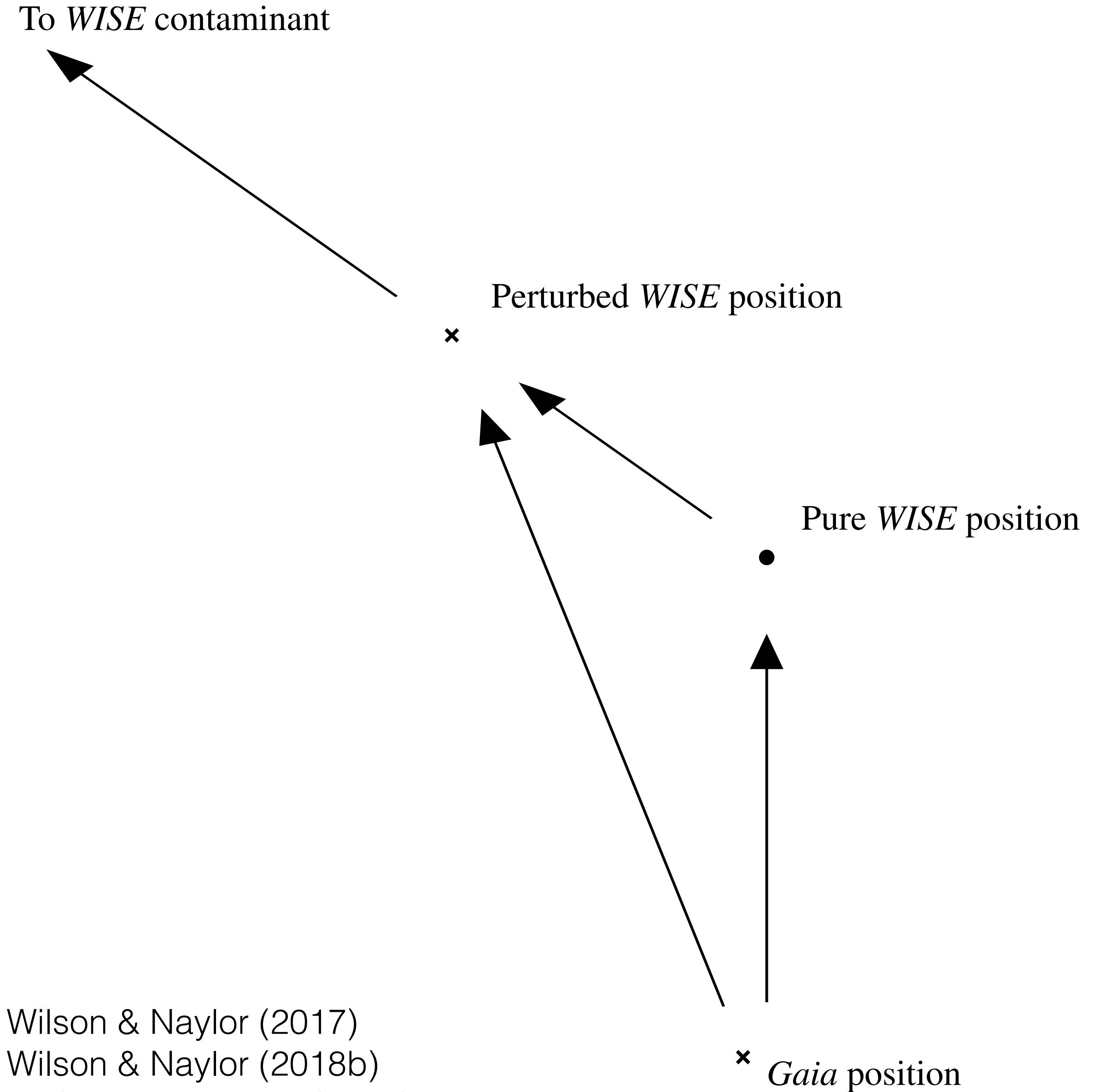
Wilson & Naylor (2018b)

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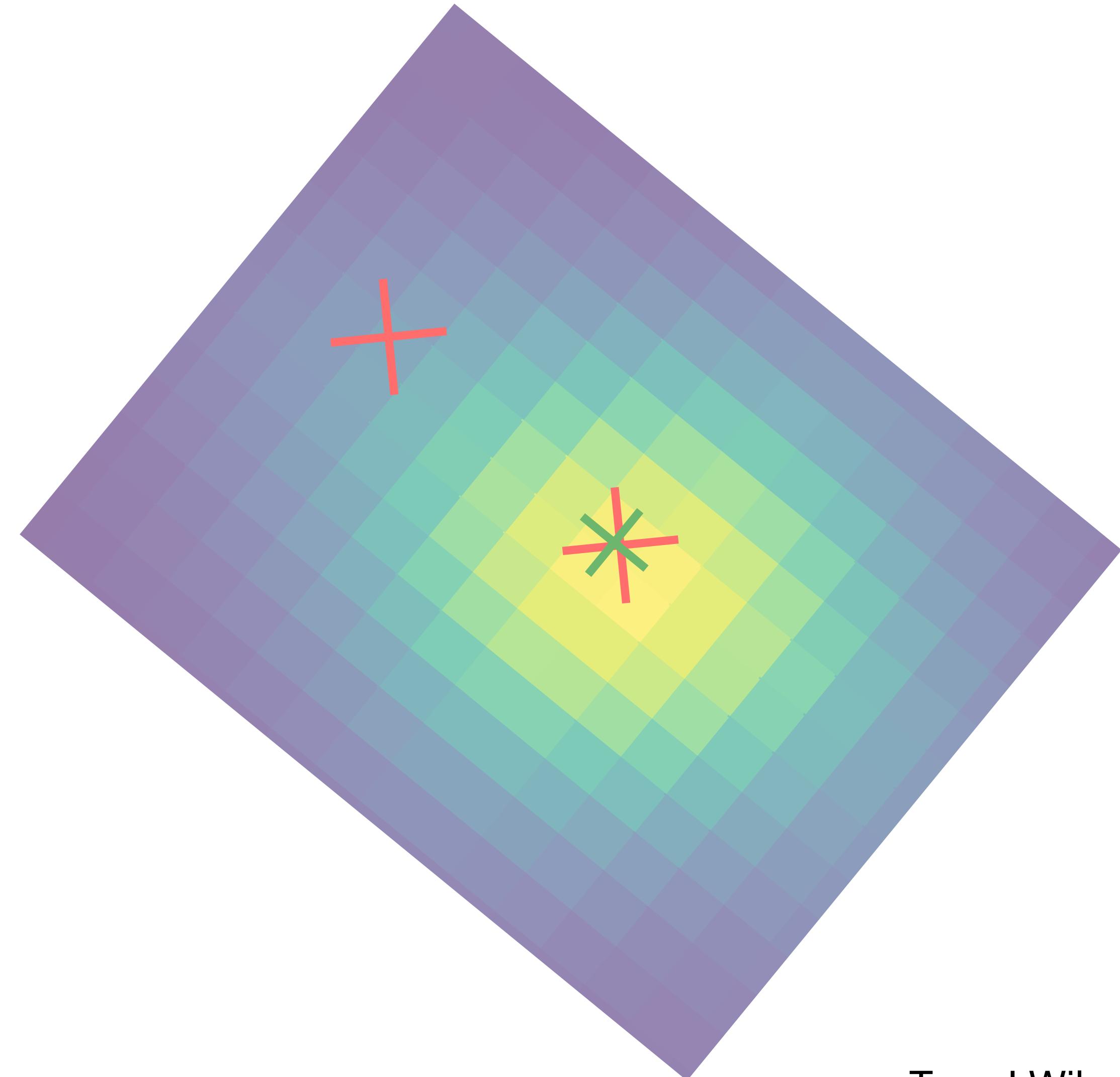


Wilson & Naylor (2017)

Wilson & Naylor (2018b)

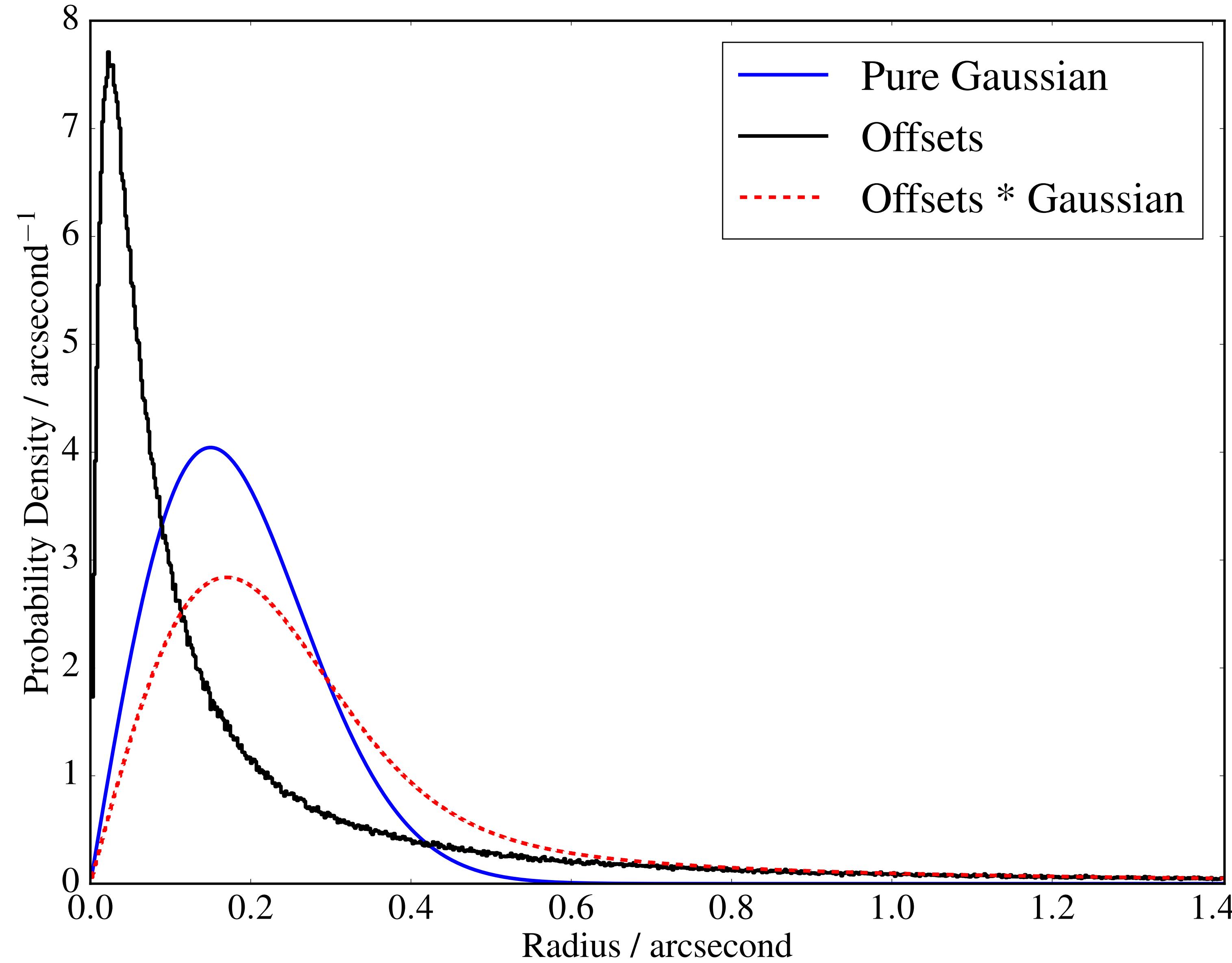
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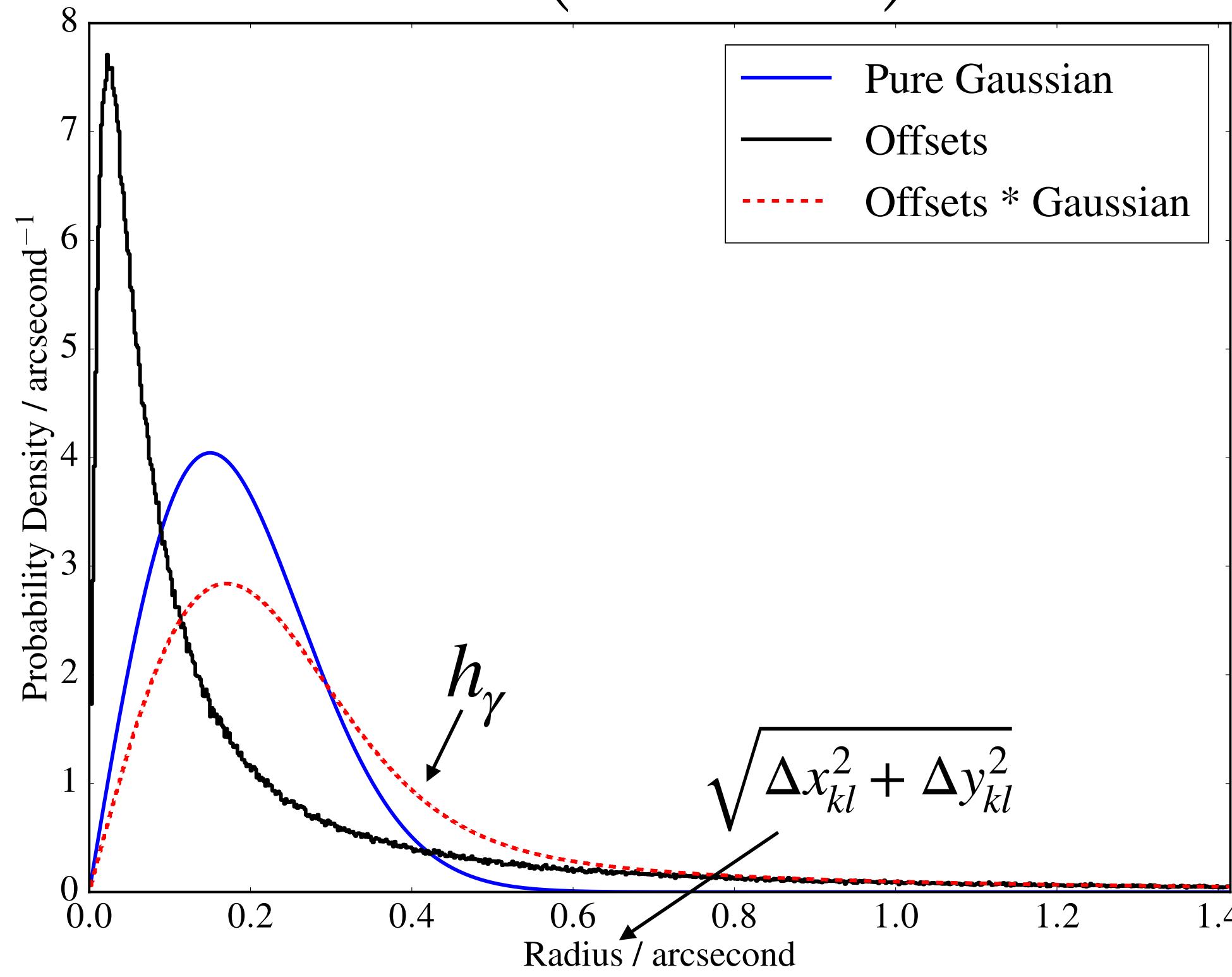


Separation Likelihood Function

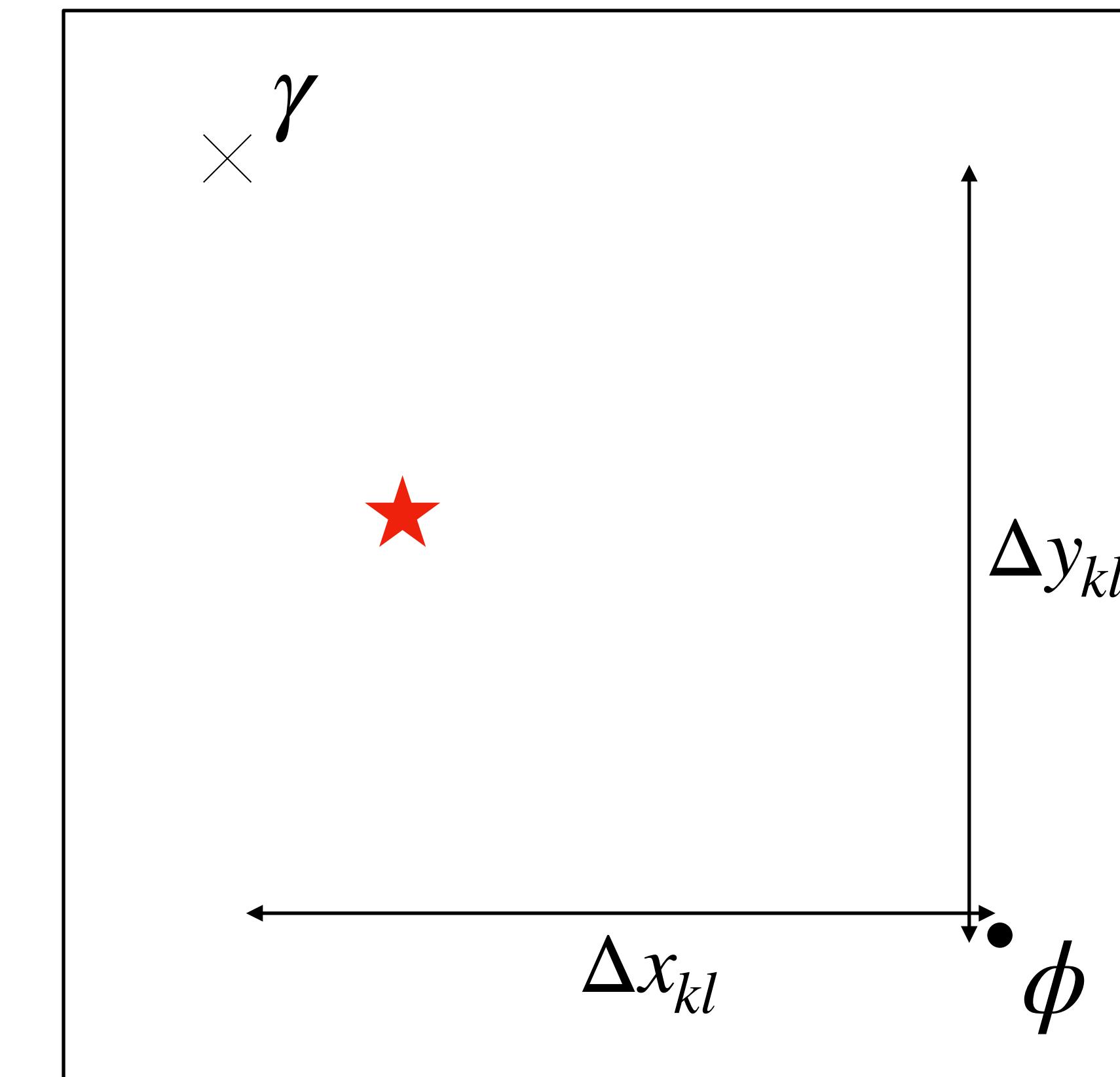
$$g(x_k, y_k, x_l, y_l) = \iint_{-\infty}^{+\infty} h_\gamma(x_0 - x_k, y_0 - y_k) h_\phi(x_l - x_0, y_l - y_0) p(x_0, y_0) dx_0 dy_0$$
$$= N_c \times (h_\gamma * h_\phi)(\Delta x_{kl}, \Delta y_{kl})$$

Wilson & Naylor (2018a)

$$g(\Delta x, \Delta y, \sigma) = (2\pi\sigma^2)^{-1} \exp\left(-\frac{1}{2}\frac{\Delta x^2 + \Delta y^2}{\sigma^2}\right) \text{ where } \sigma^2 = \sigma_1^2 + \sigma_2^2$$



Wilson & Naylor (2018b)



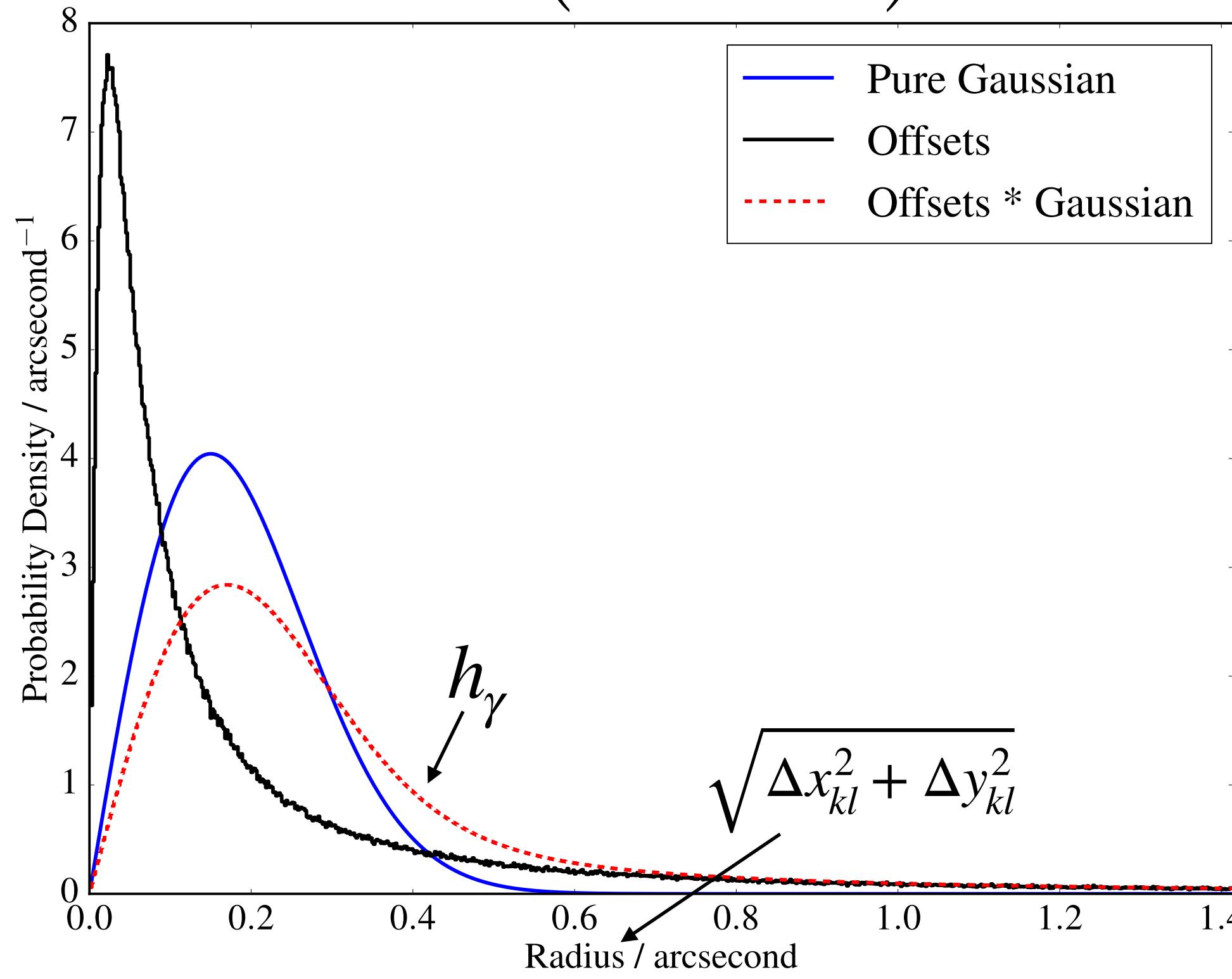
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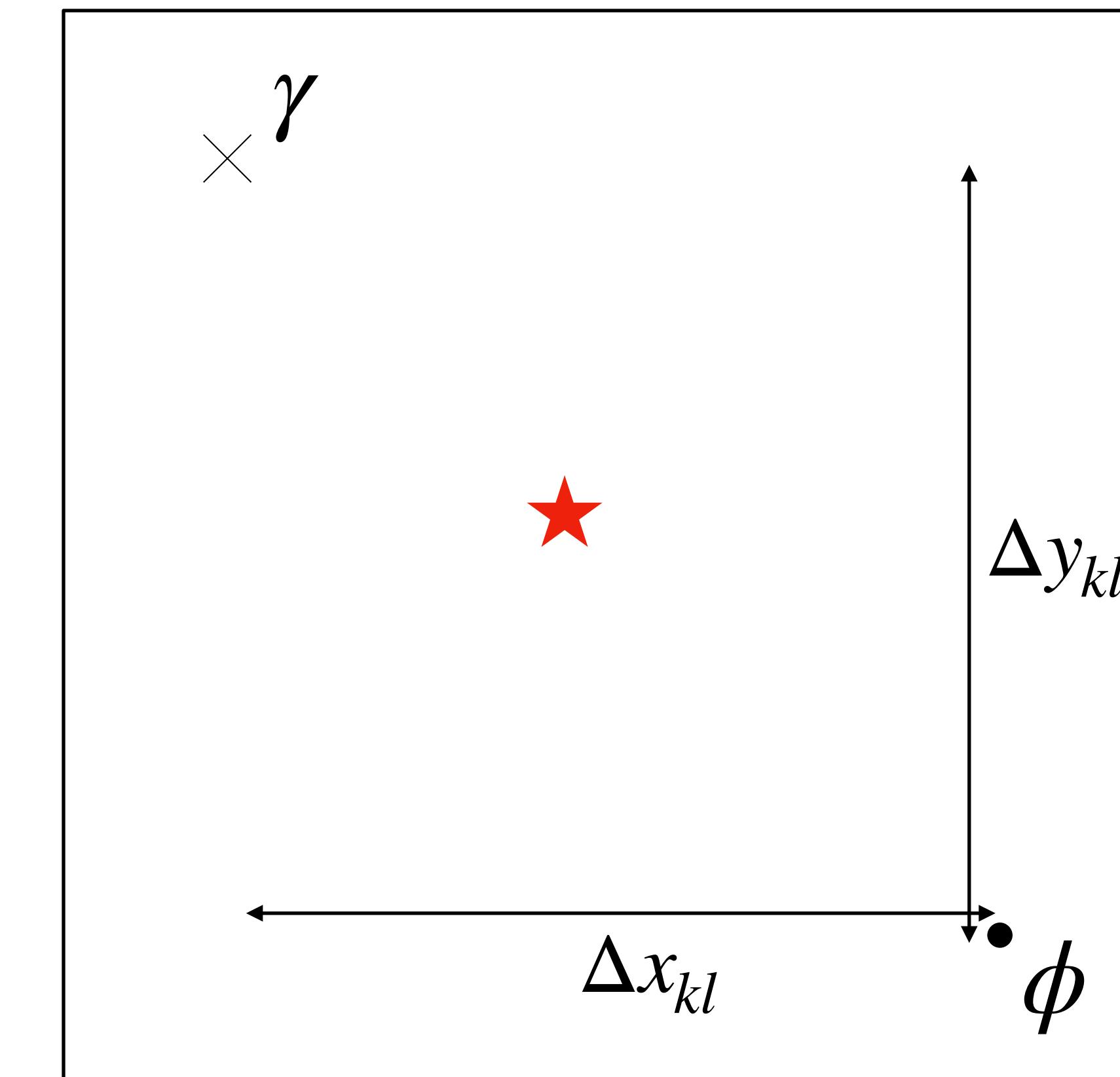
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Wilson & Naylor (2018b)



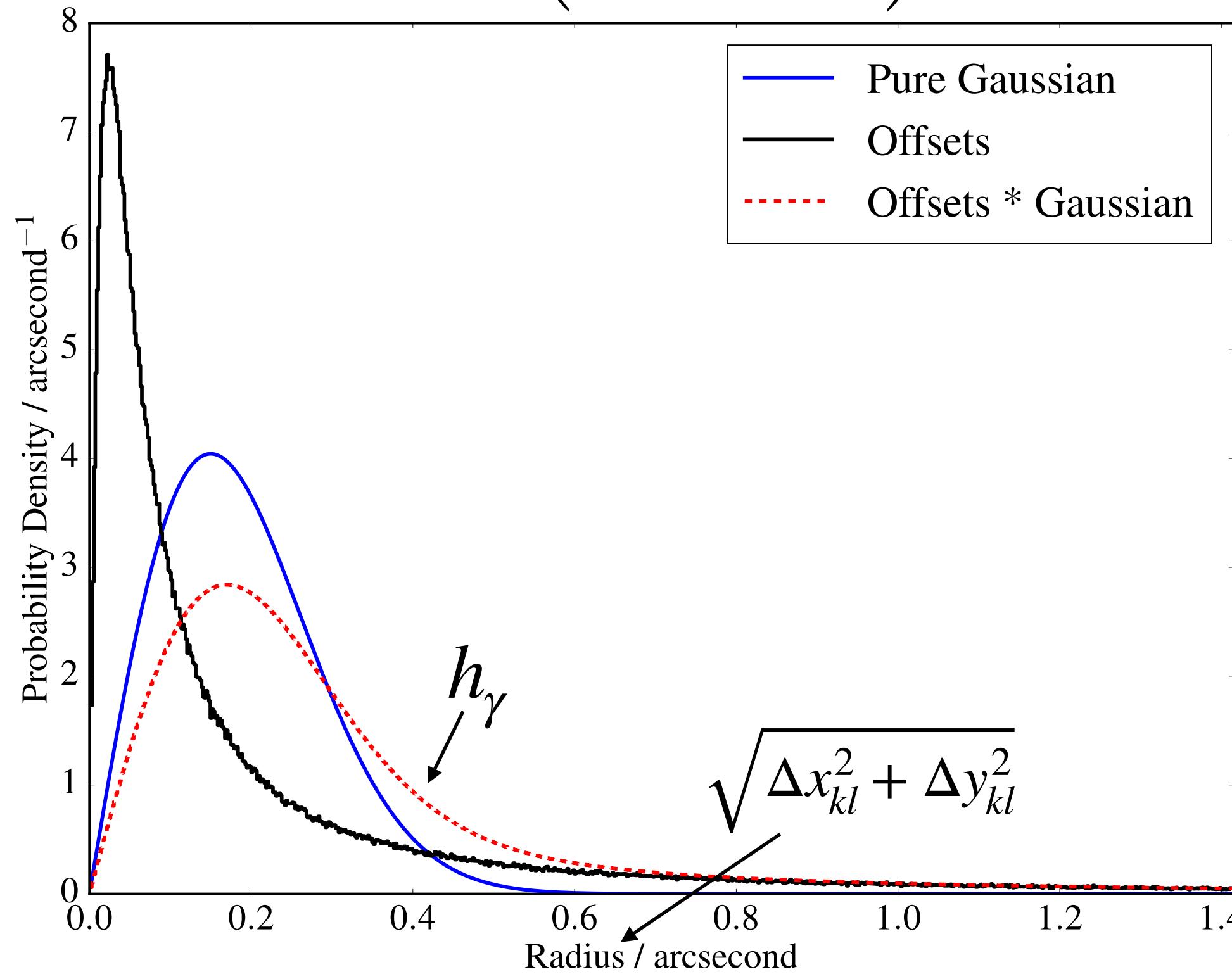
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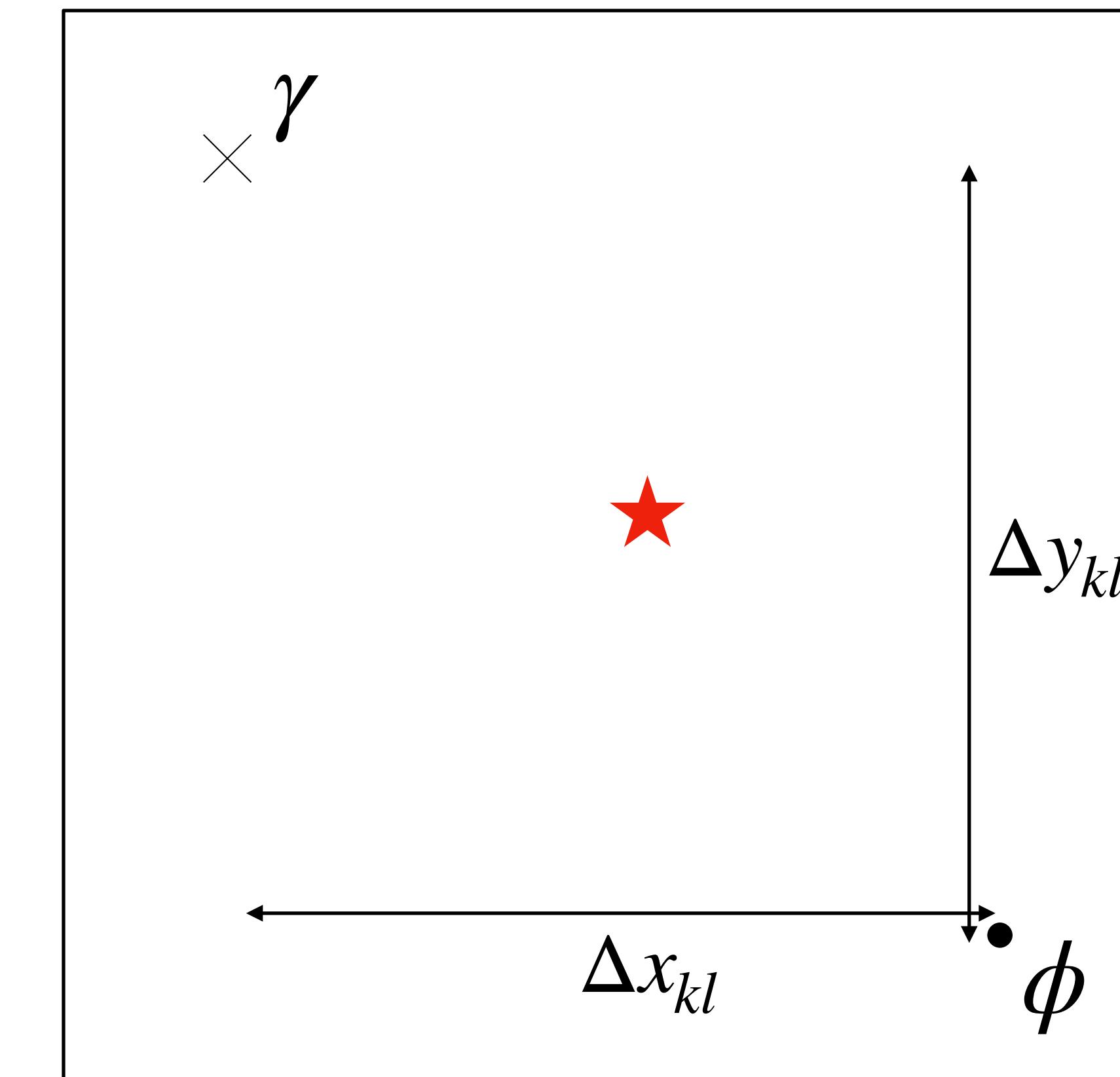
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Wilson & Naylor (2018b)



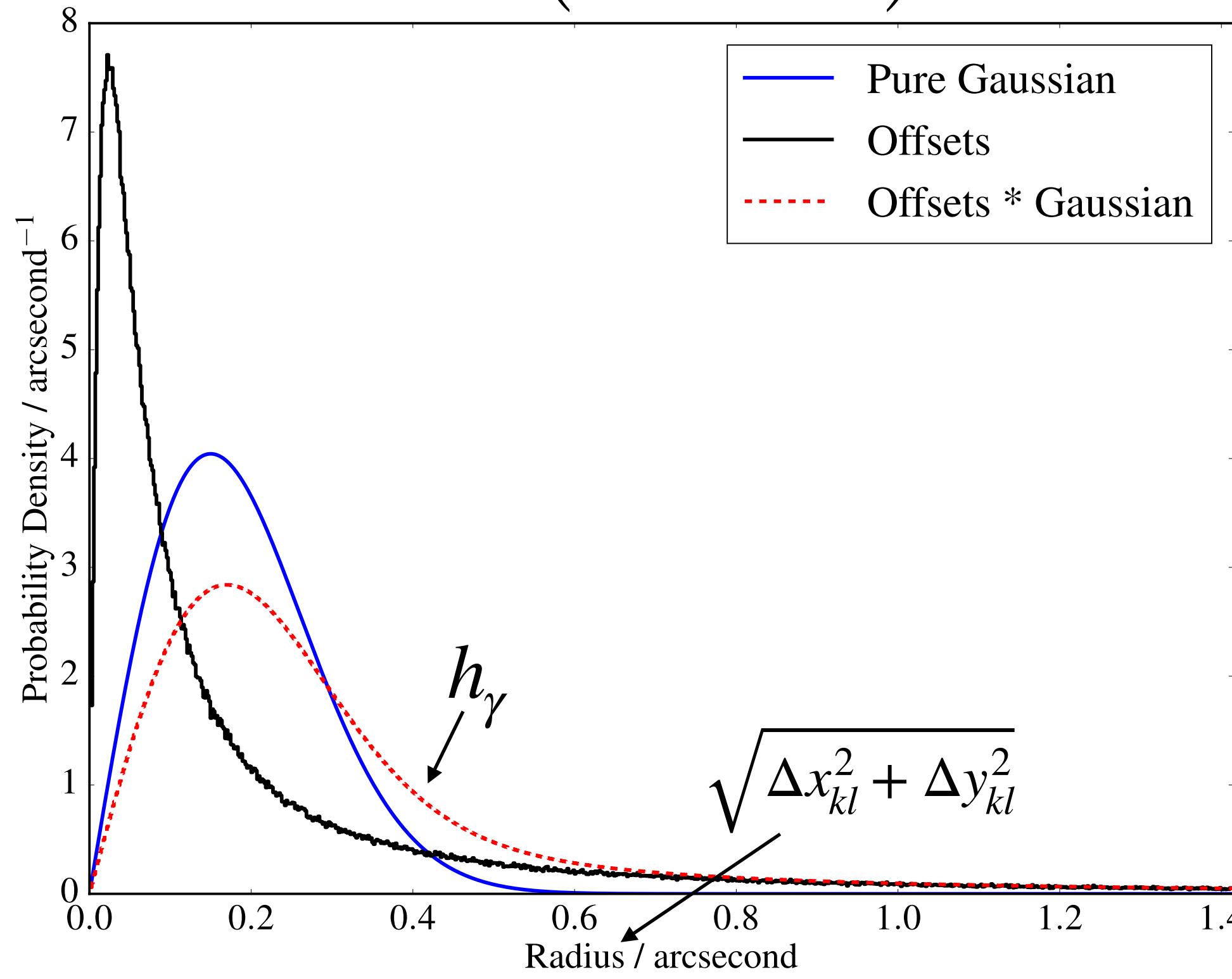
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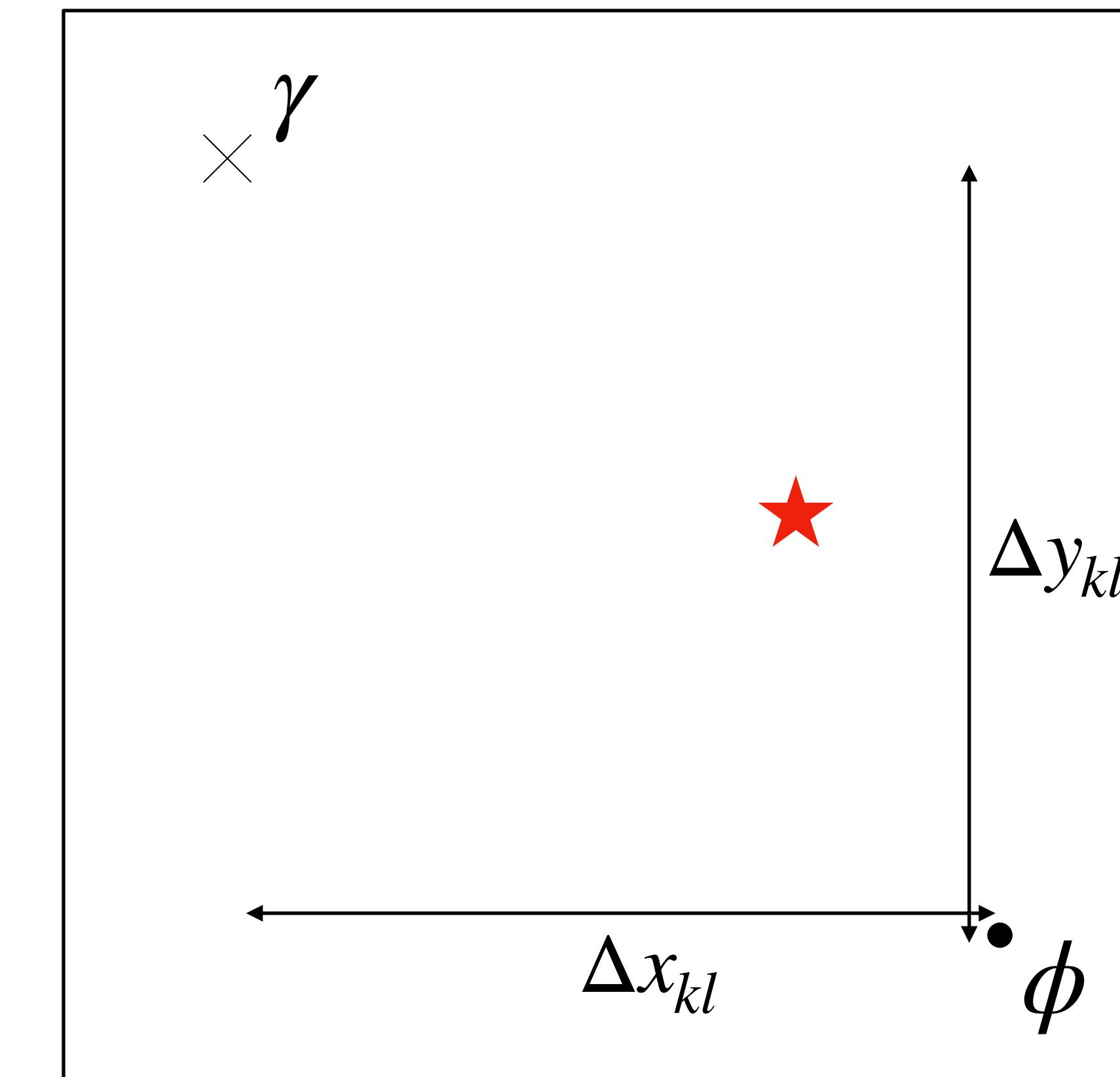
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Wilson & Naylor (2018b)



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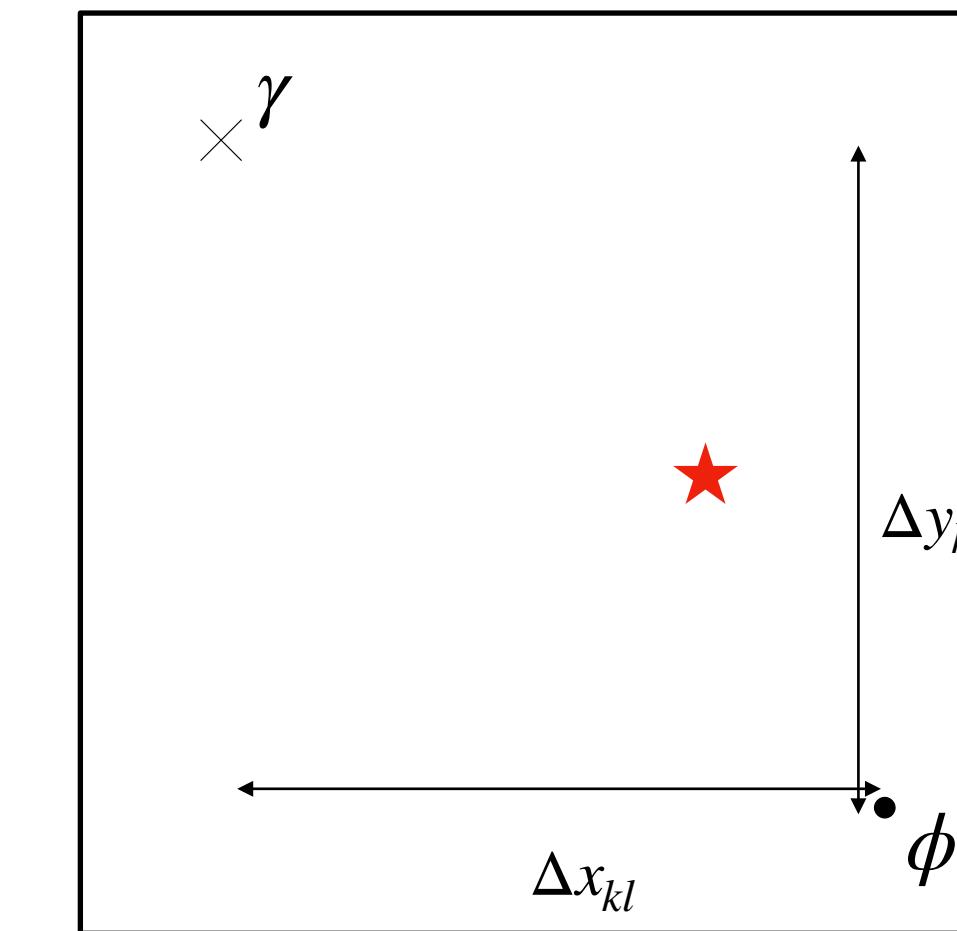
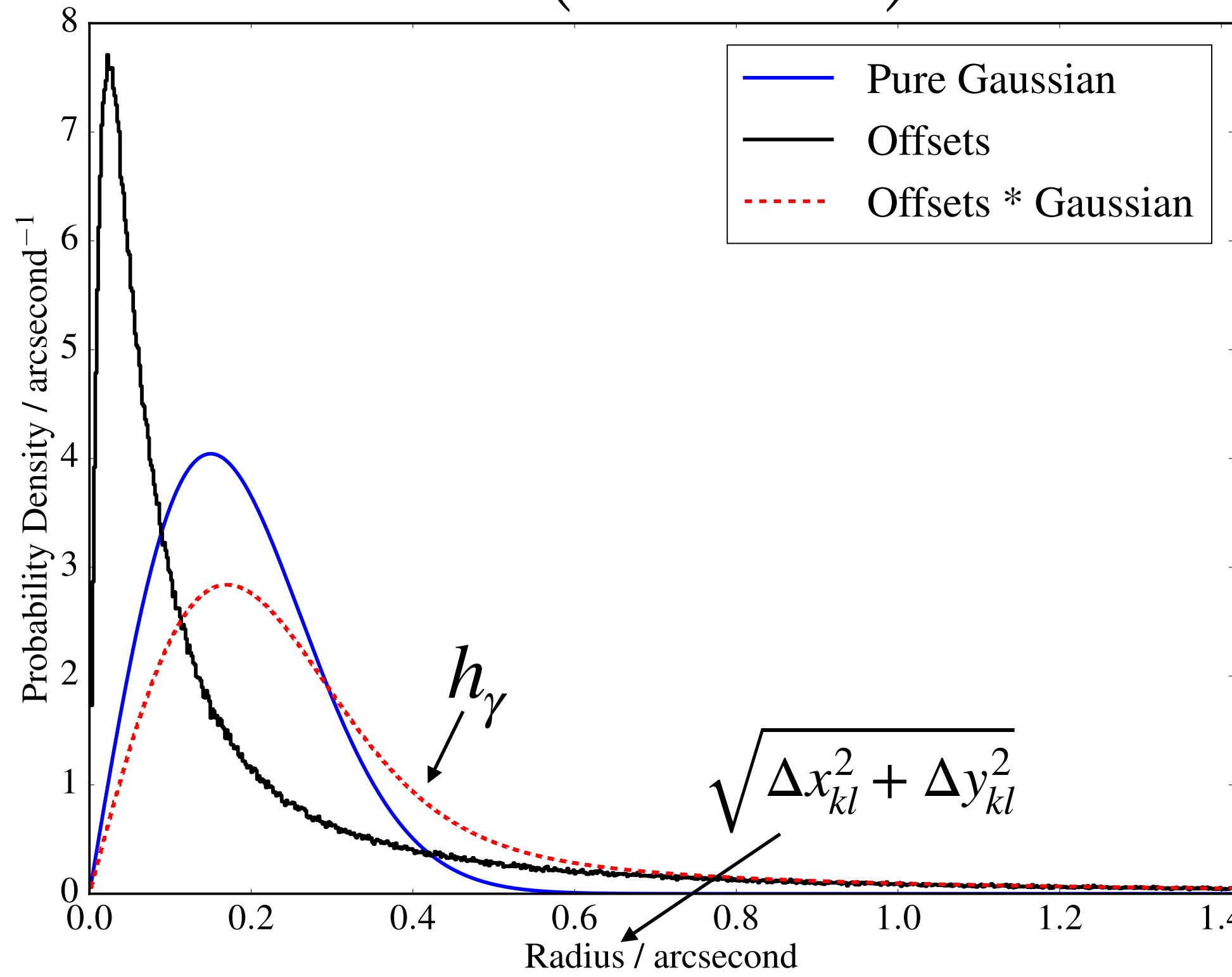
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Wilson & Naylor (2018a)

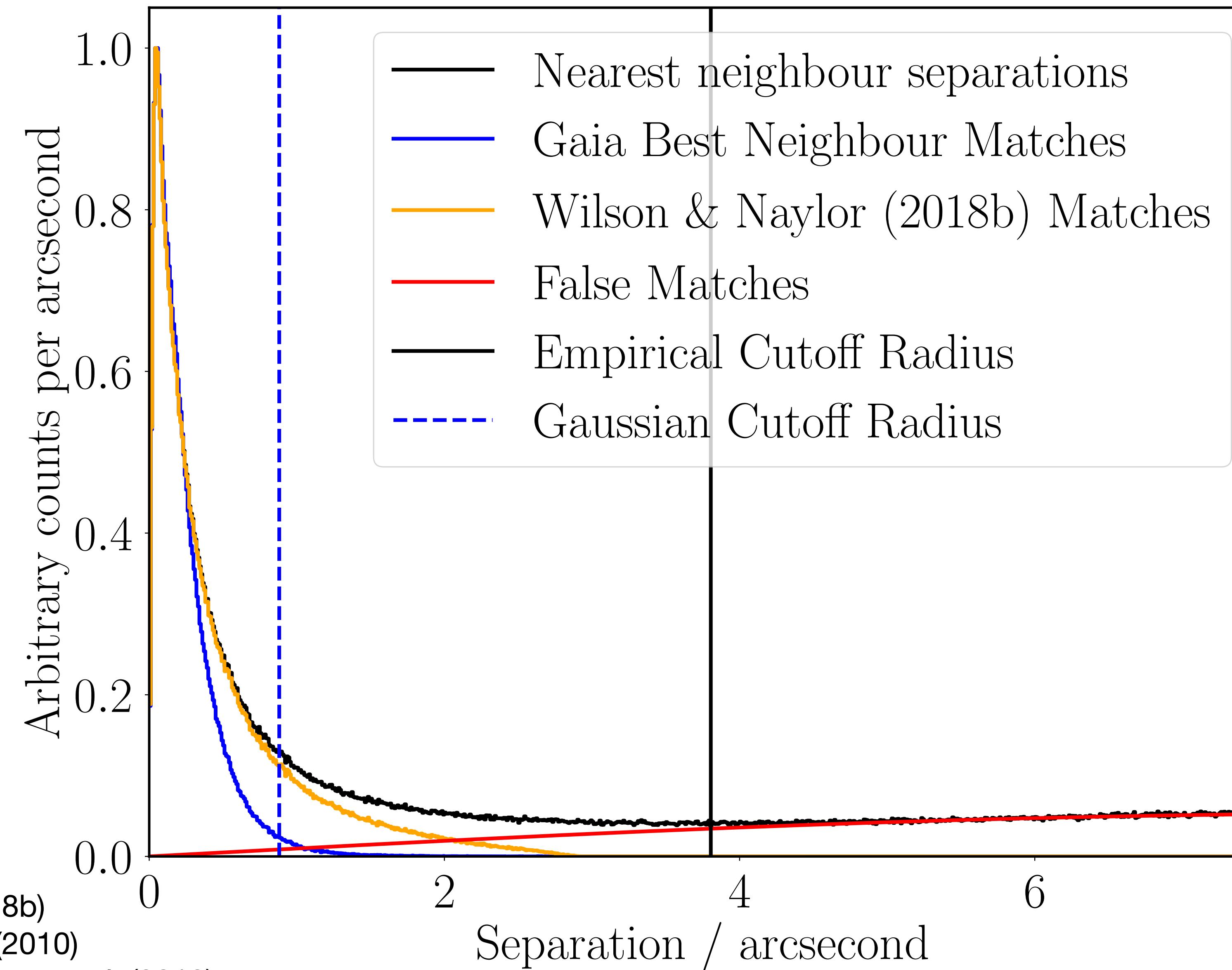
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John Herschel's result, 170 years on, appears wrong, but only because the universe got in the way.
“Were the succession of stars endless... there could be absolutely no point, in all that background, at which would not exist a star.”
— Edgar Allan Poe, Eureka (1848)

Match Separations



Wilson & Naylor (2018b)

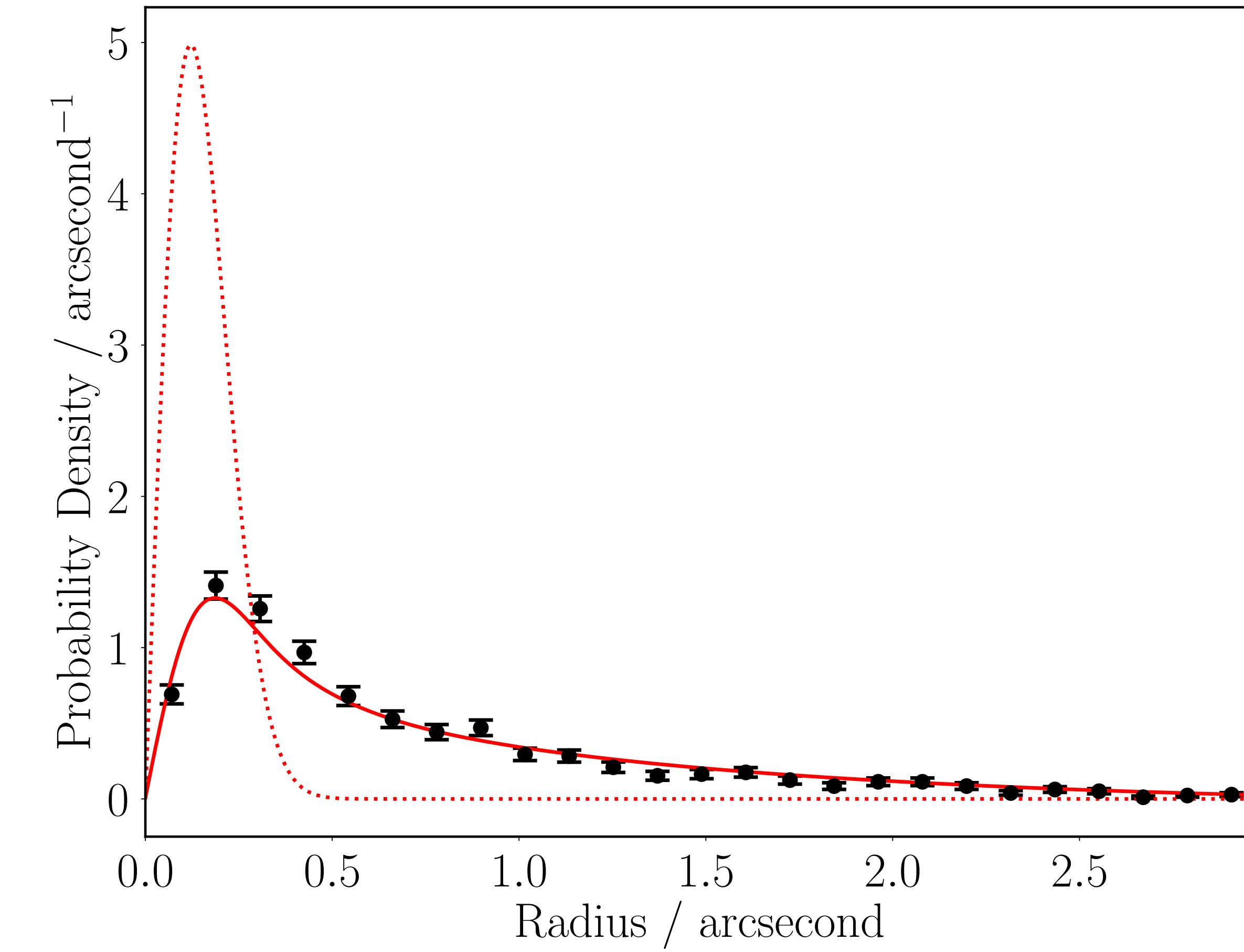
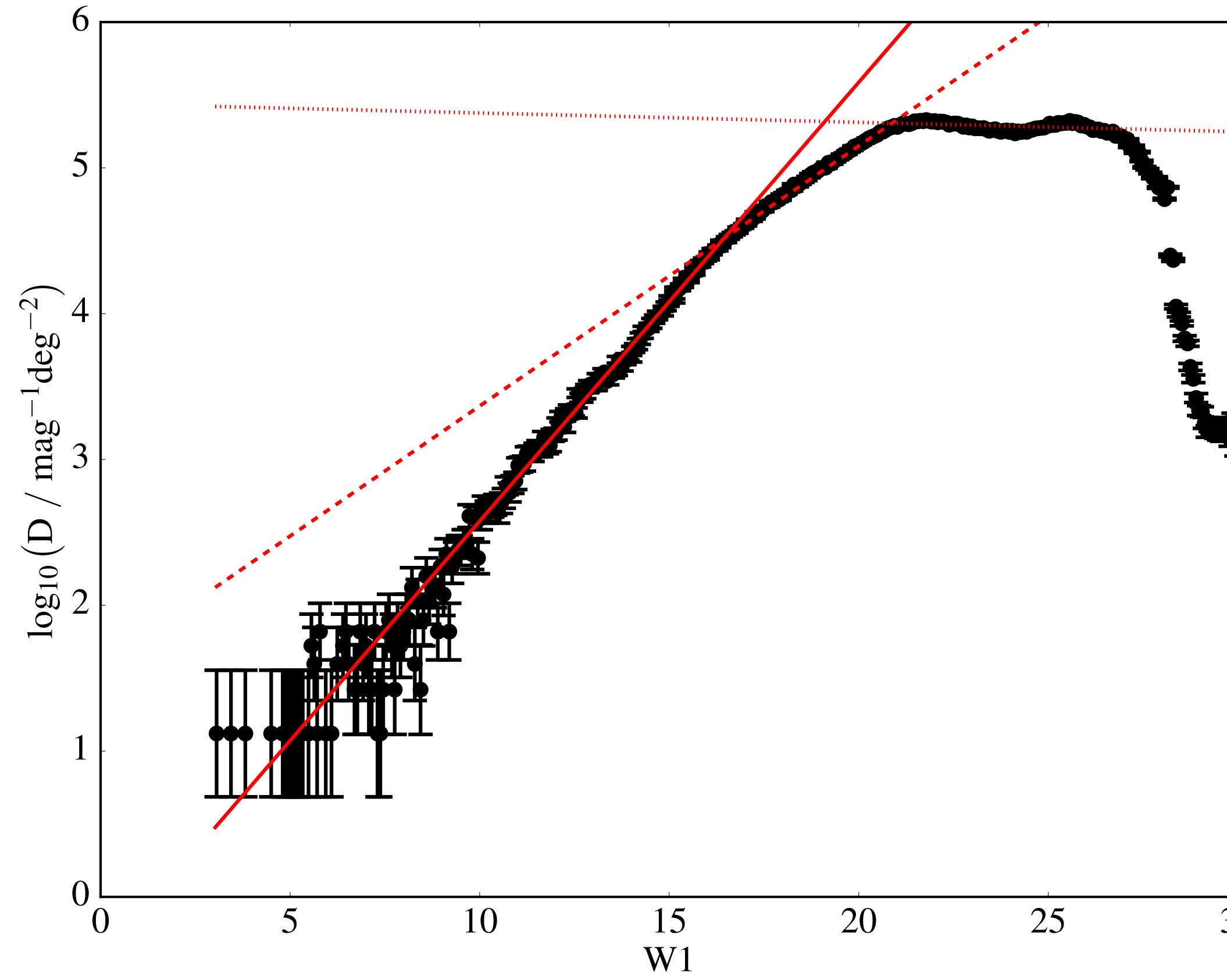
WISE - Wright et al. (2010)

Gaia matches - Marrese et al. (2019)

Gaia DR2 - Gaia Collaboration, Brown A. G. A., et al. (2018)

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Building Empirical AUFS

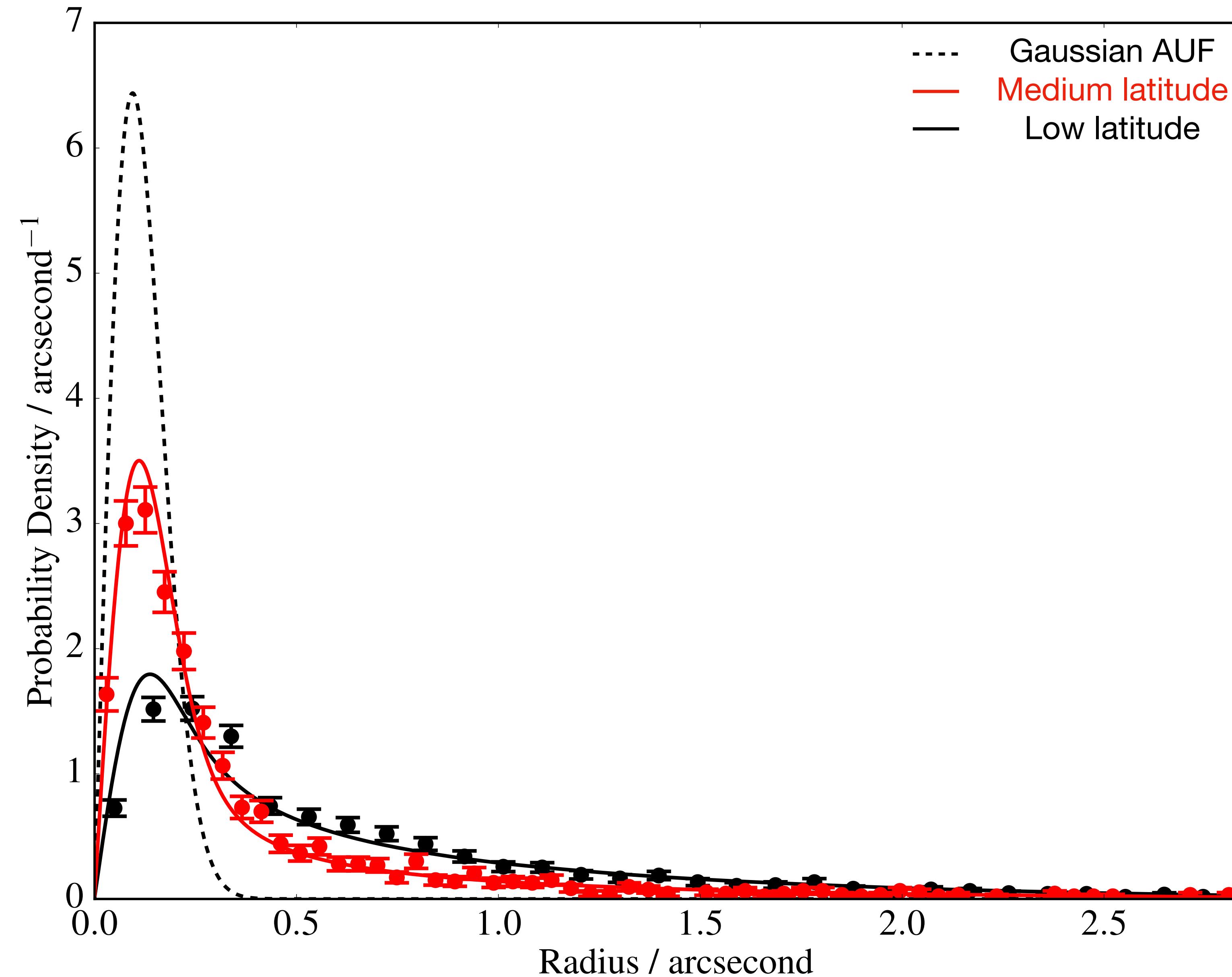


Wilson & Naylor (2018b)

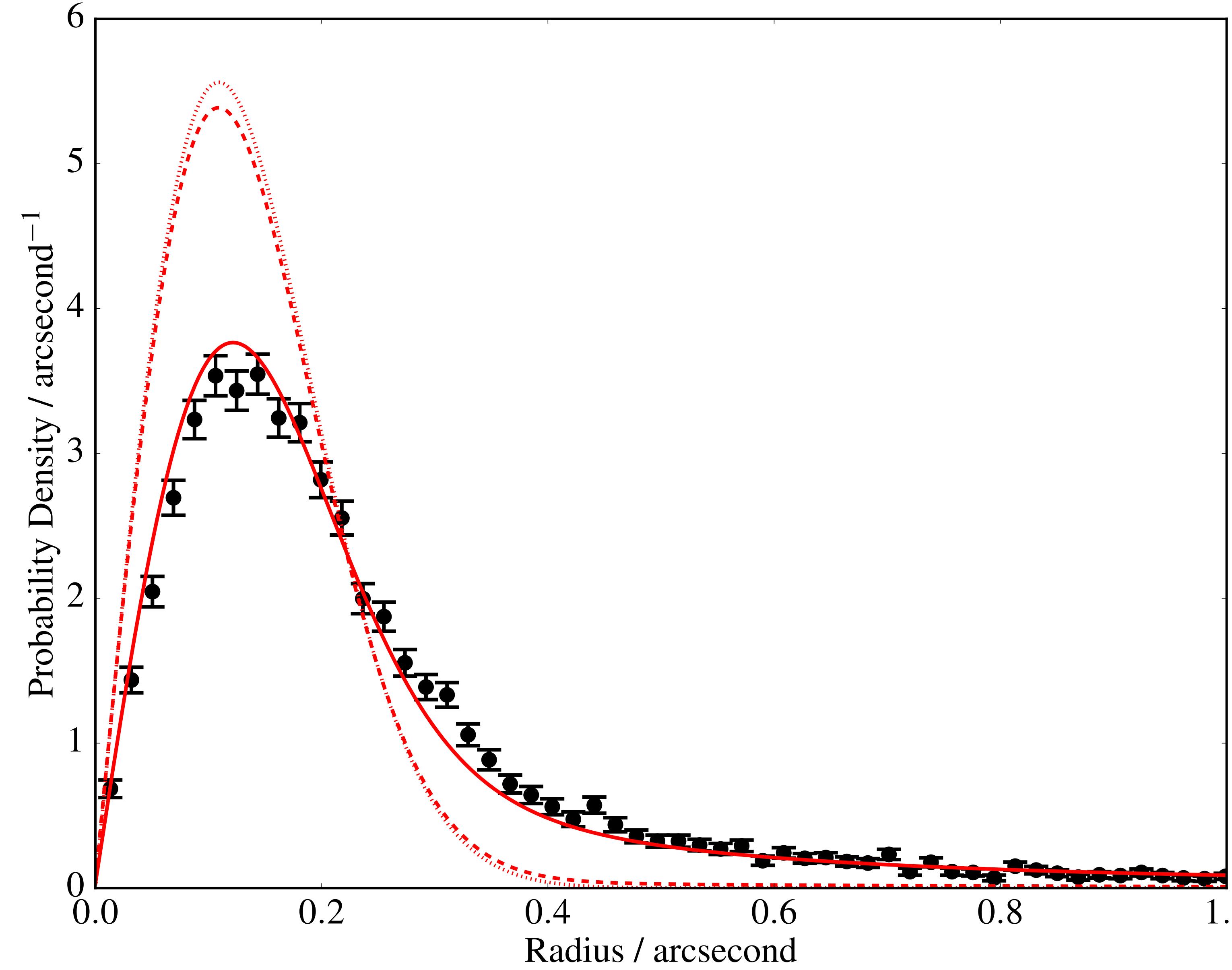
TRILEGAL - Girardi et al. (2005)

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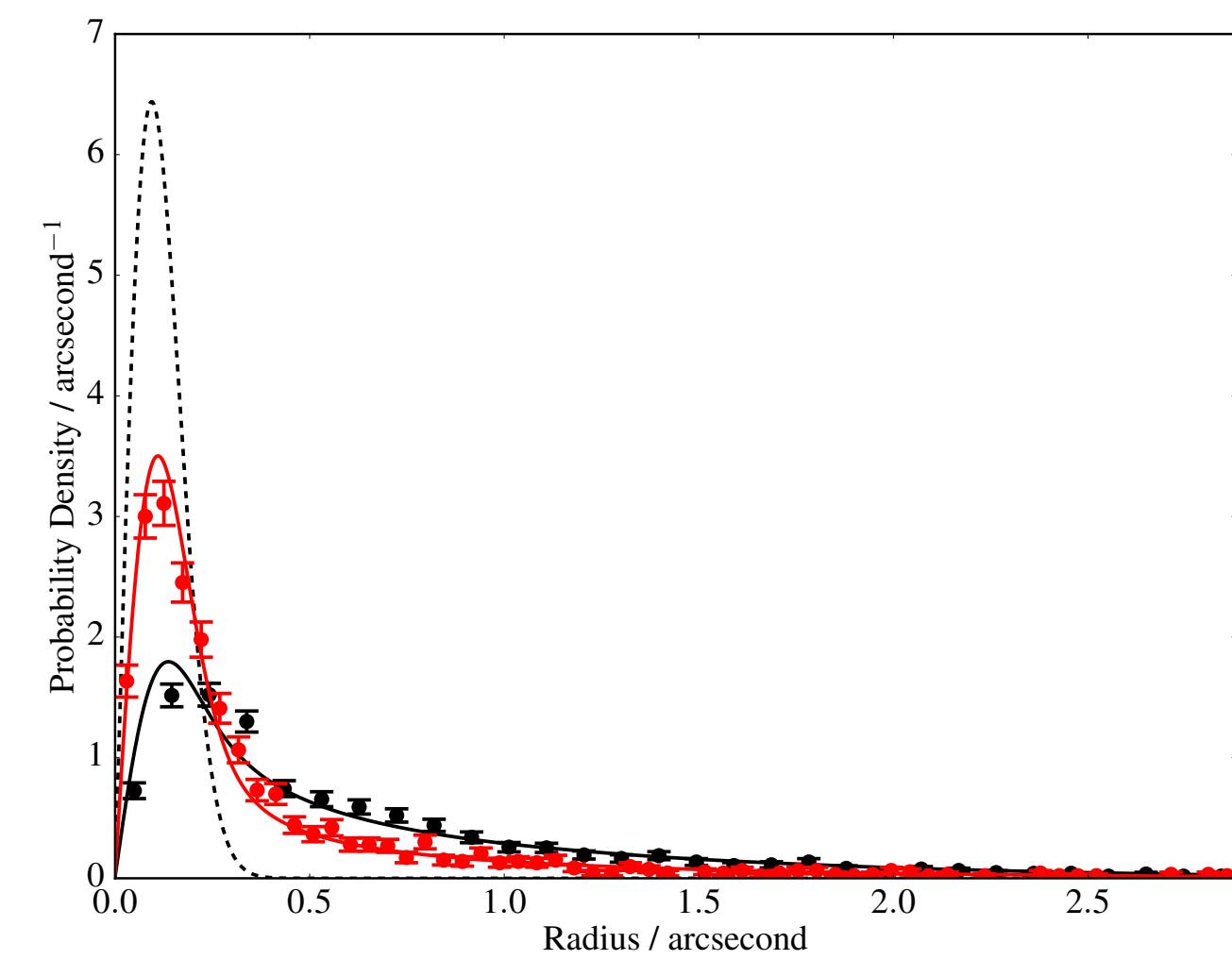
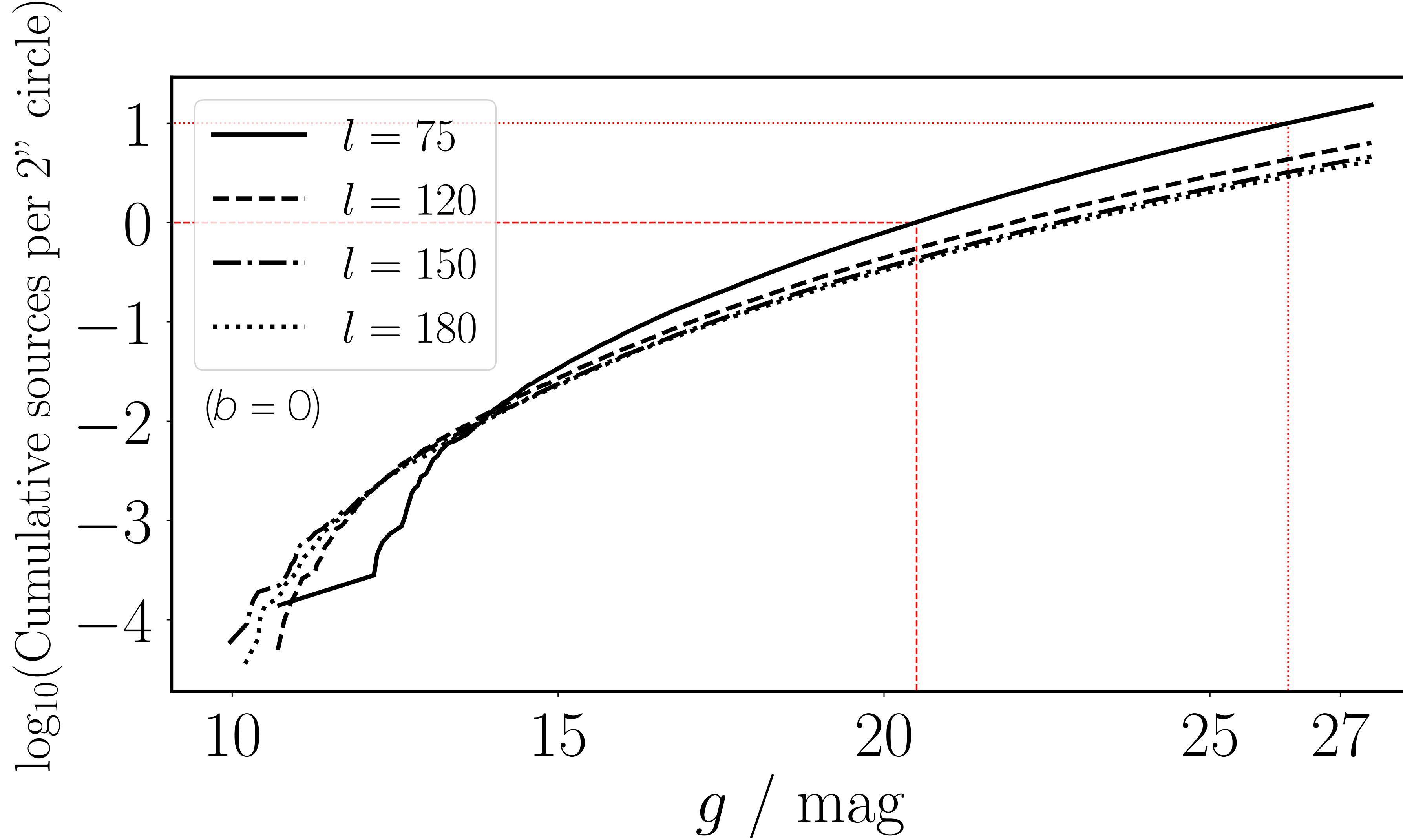
Crowding Normalisation



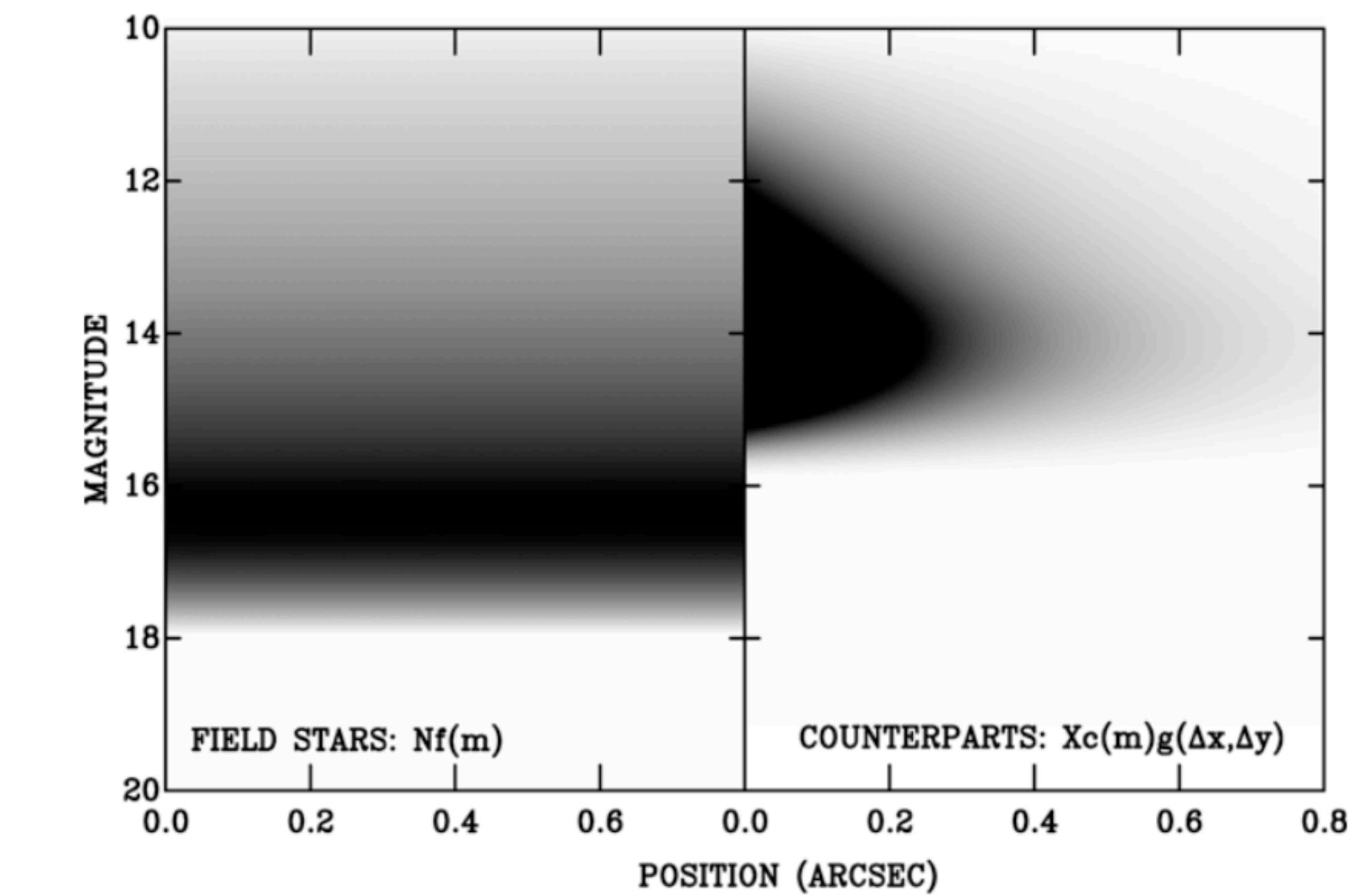
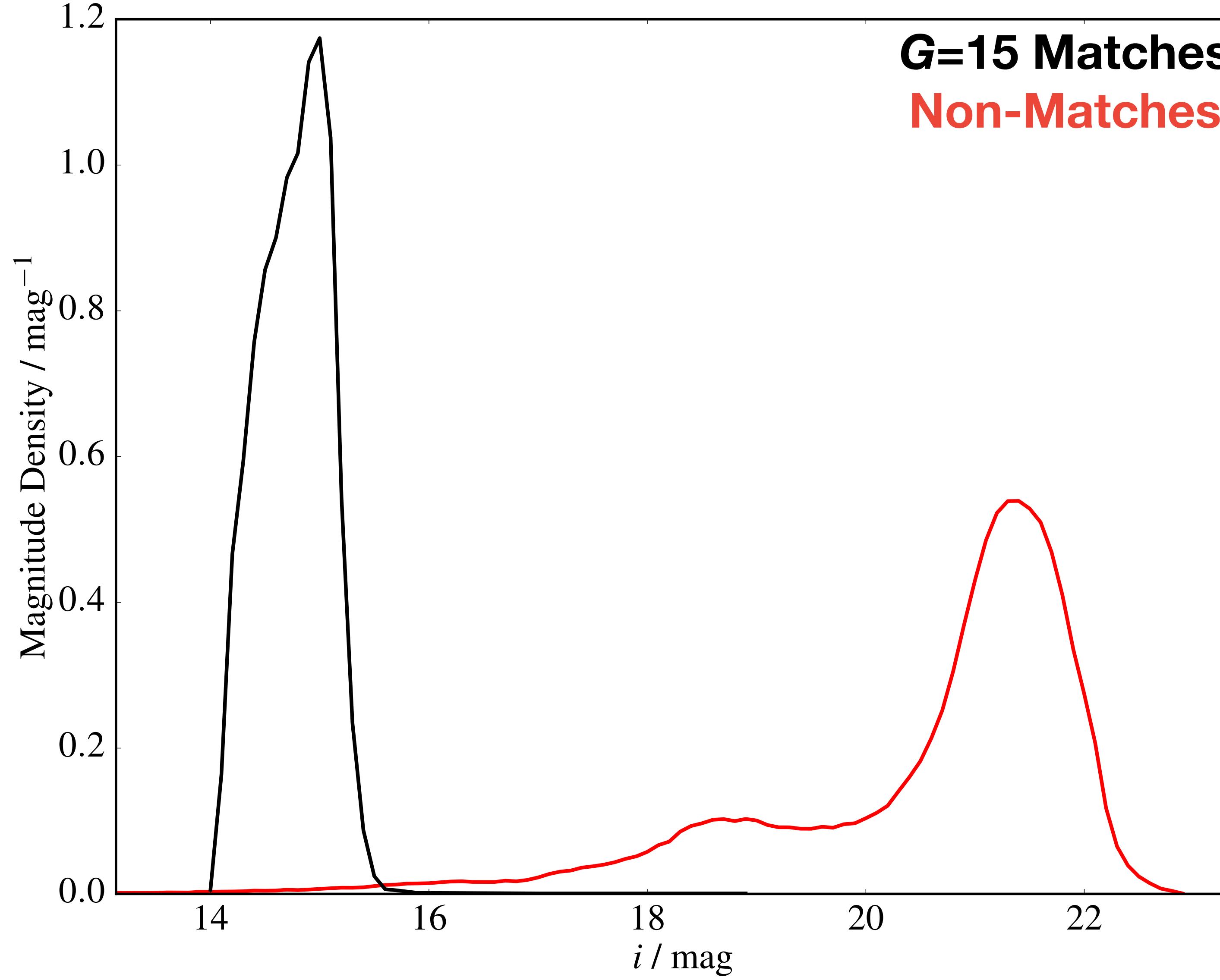
Extra-galactic Effects of Crowding



Vera C. Rubin Observatory's LSST



Including the Magnitude Information



Naylor, Broos, & Feigelson (2013)

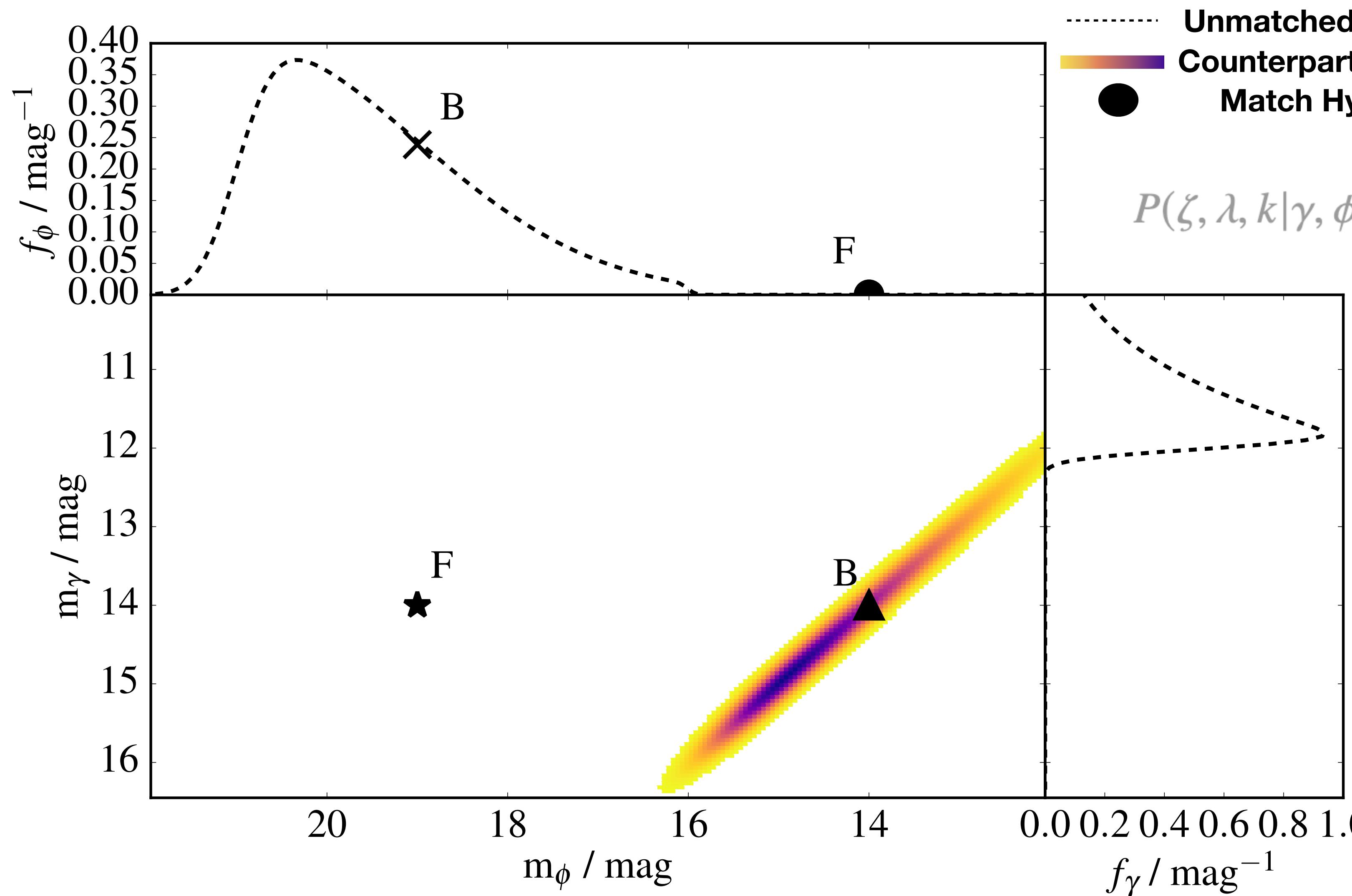
Wilson & Naylor (2018a)

IPHAS - Barentsen et al. (2014)

Gaia DR2 - Gaia Collaboration, Brown A. G. A., et al. (2018)

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Including the Magnitude Information

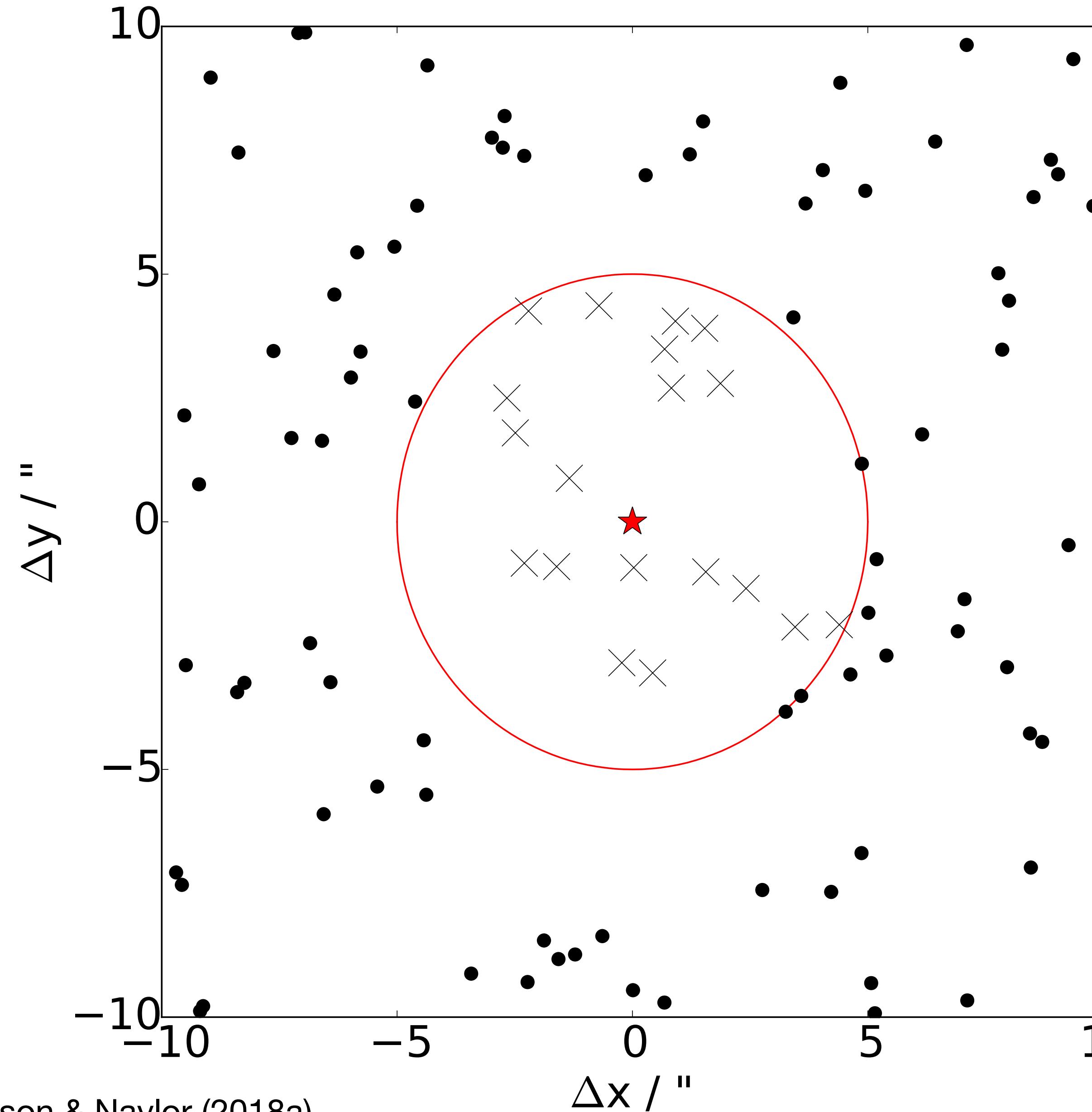


Unmatched Distribution
Counterpart Distribution
Match Hypotheses

$$P(\zeta, \lambda, k | \gamma, \phi) = K^{-1} \times \prod_{\delta \notin \zeta \cap \delta \in \gamma} N_\gamma f_\gamma^\delta \prod_{\omega \notin \lambda \cap \omega \in \phi} N_\phi f_\phi^\omega \prod_{i=1}^k N_c G_{\gamma\phi}^{\zeta_i \lambda_i} c_{\gamma\phi}^{\zeta_i \lambda_i}$$

Unmatched Matched

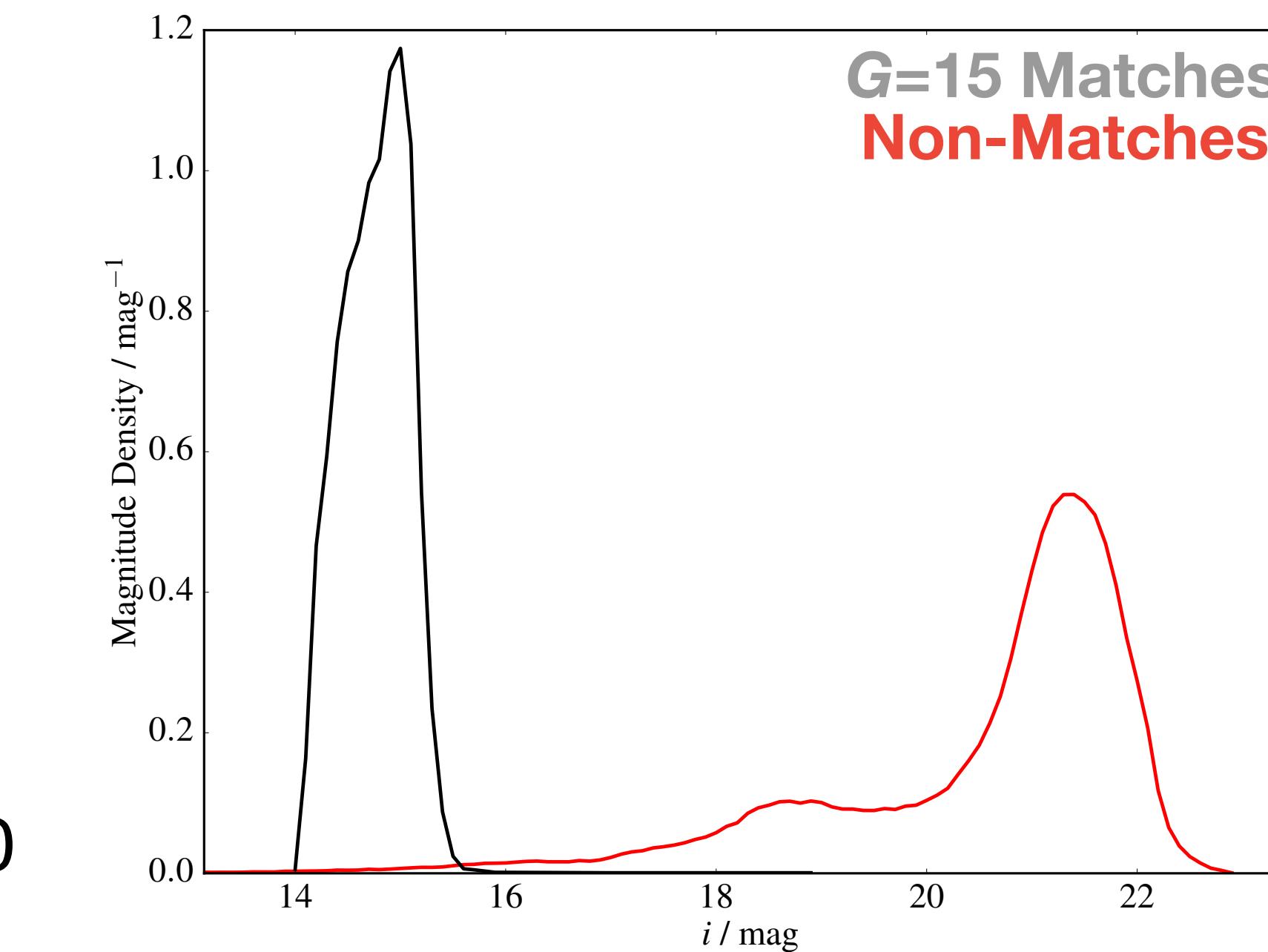
The “Field Star” Distribution



Wilson & Naylor (2018a)

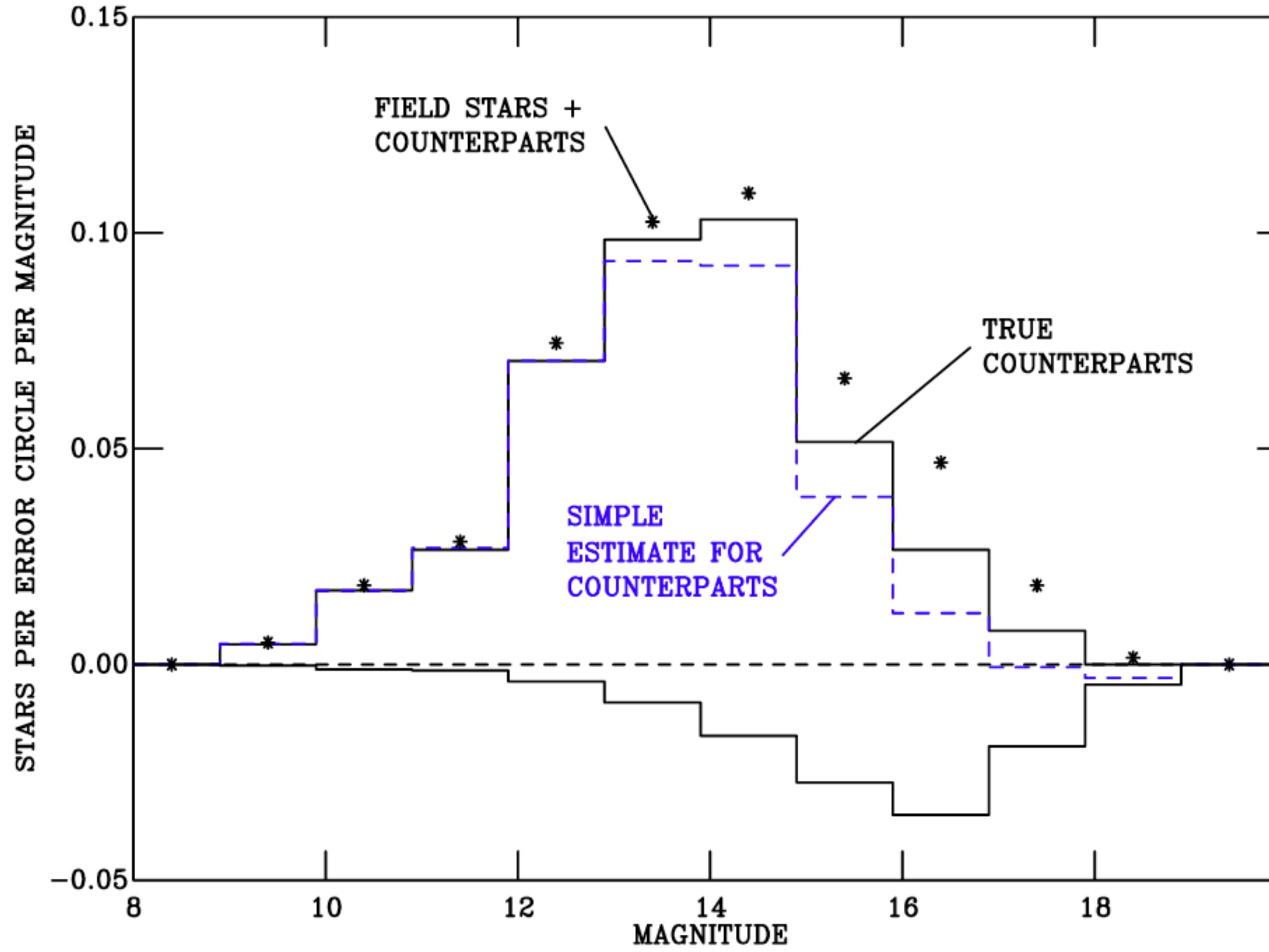
$$P(\zeta, \lambda, k | \gamma, \phi) = K^{-1} \prod_{\delta \notin \zeta \cap \delta \in \gamma} N_\gamma f_\gamma^\delta \prod_{\omega \notin \lambda \cap \omega \in \phi} N_\phi f_\phi^\omega \prod_{i=1}^k N_c G_{\gamma\phi}^{\zeta_i \lambda_i} c_{\gamma\phi}^{\zeta_i \lambda_i}$$

Unmatched Matched

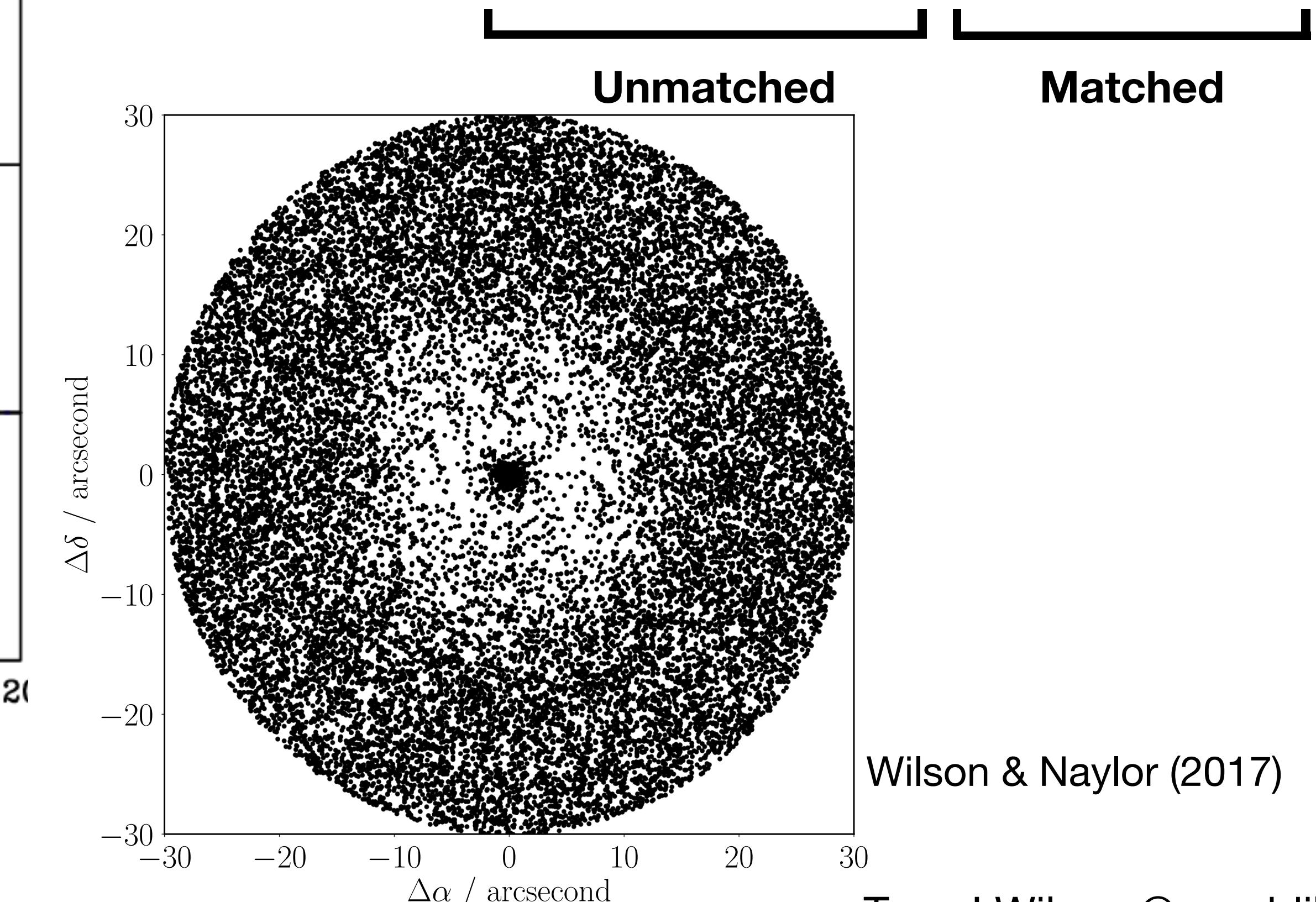


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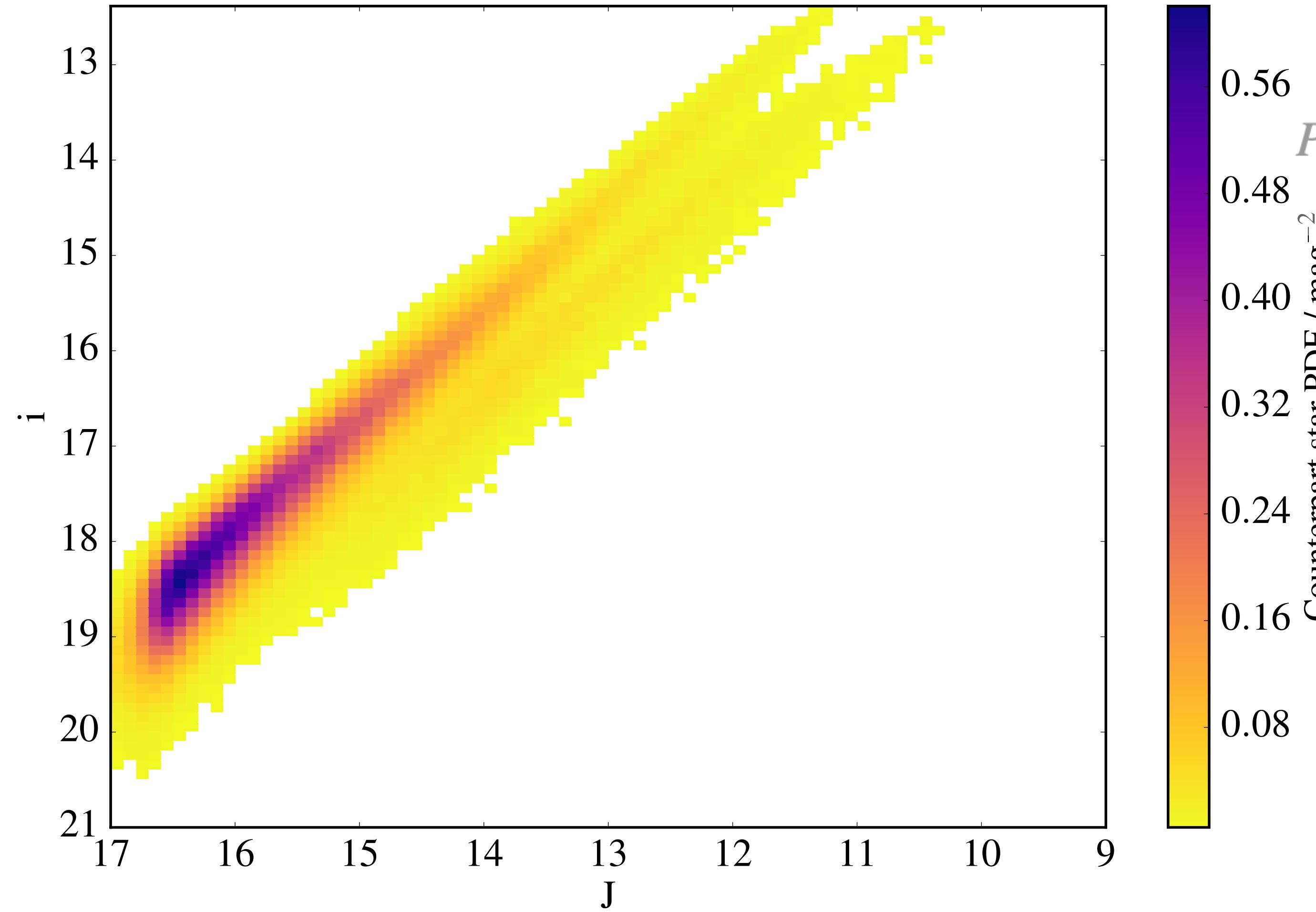
The Counterpart Source Distribution



$$P(\zeta, \lambda, k | \gamma, \phi) = K^{-1} \times \prod_{\delta \notin \zeta \cap \delta \in \gamma} N_\gamma f_\gamma^\delta \prod_{\omega \notin \lambda \cap \omega \in \phi} N_\phi f_\phi^\omega \prod_{i=1}^k N_c G_{\gamma\phi}^{\zeta_i \lambda_i} c_{\gamma\phi}^{\zeta_i \lambda_i}$$



The Counterpart Source Distribution

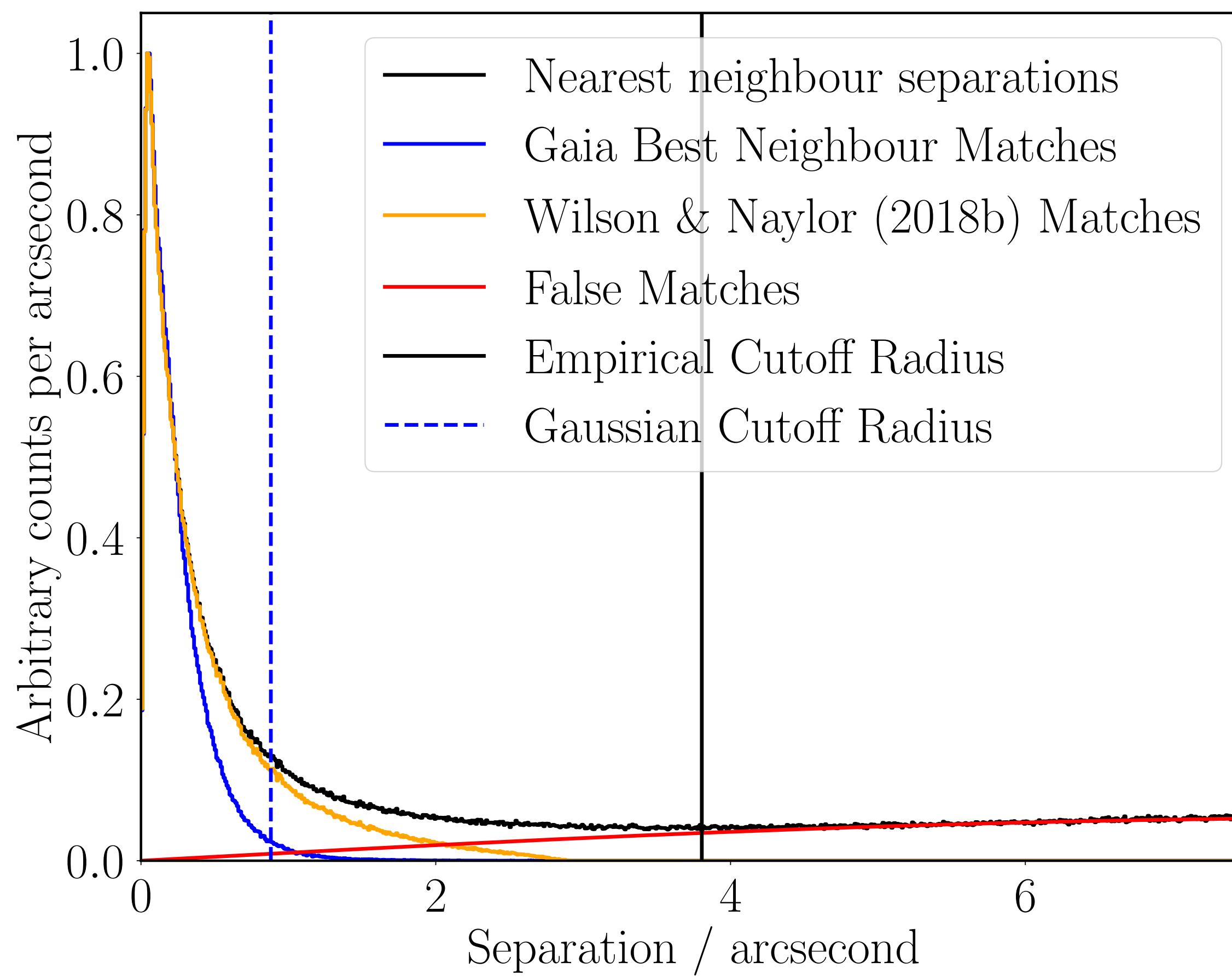


Counterpart star PDF / mag^{-2}

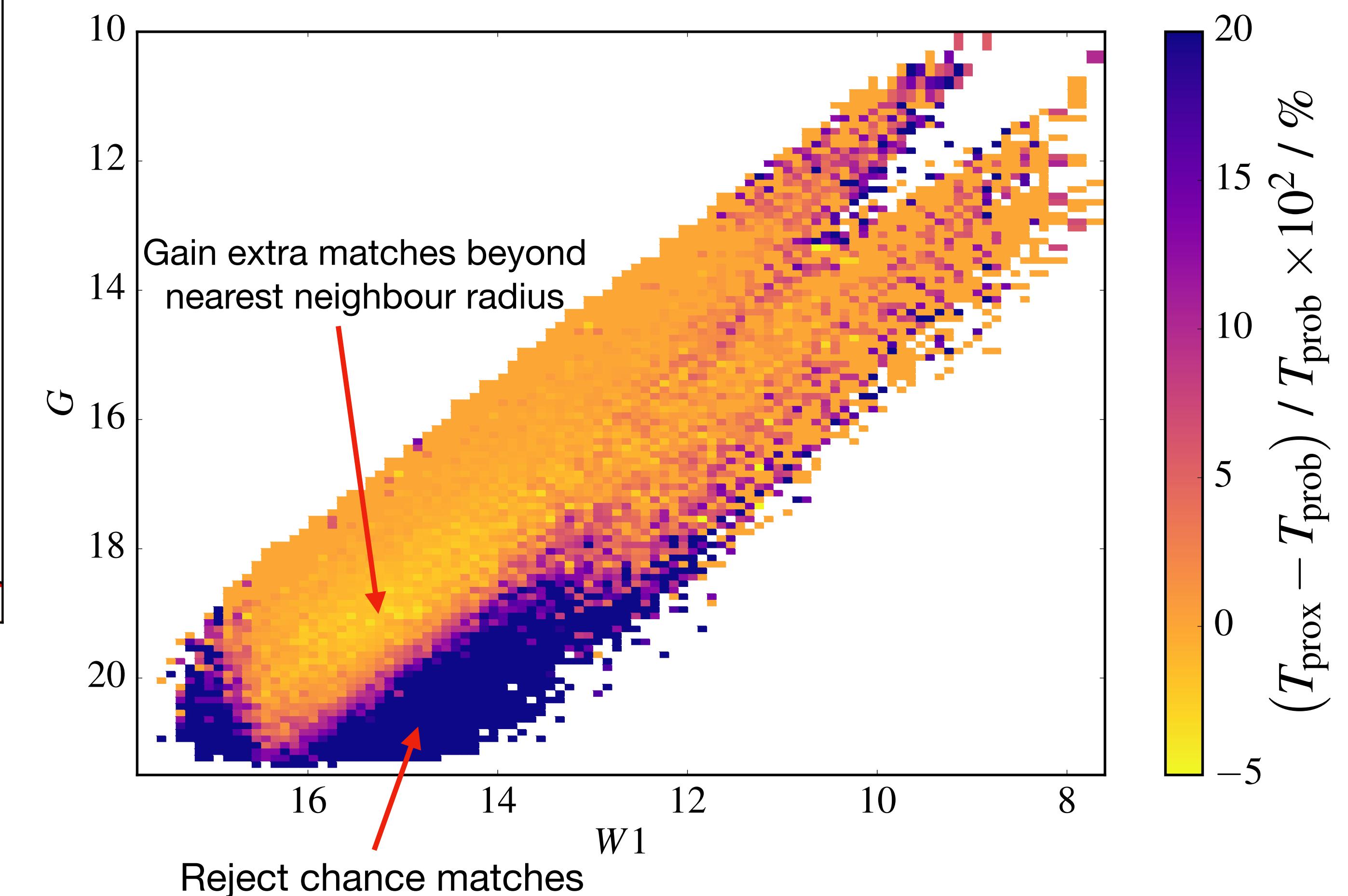
$$P(\zeta, \lambda, k | \gamma, \phi) = K^{-1} \times \prod_{\delta \notin \zeta \cap \delta \in \gamma} N_\gamma f_\gamma^\delta \prod_{\omega \notin \lambda \cap \omega \in \phi} N_\phi f_\phi^\omega \prod_{i=1}^k N_c G_{\gamma\phi}^{\zeta_i \lambda_i} c_{\gamma\phi}^{\zeta_i \lambda_i}$$

Unmatched **Matched**

Comparing Match Distributions

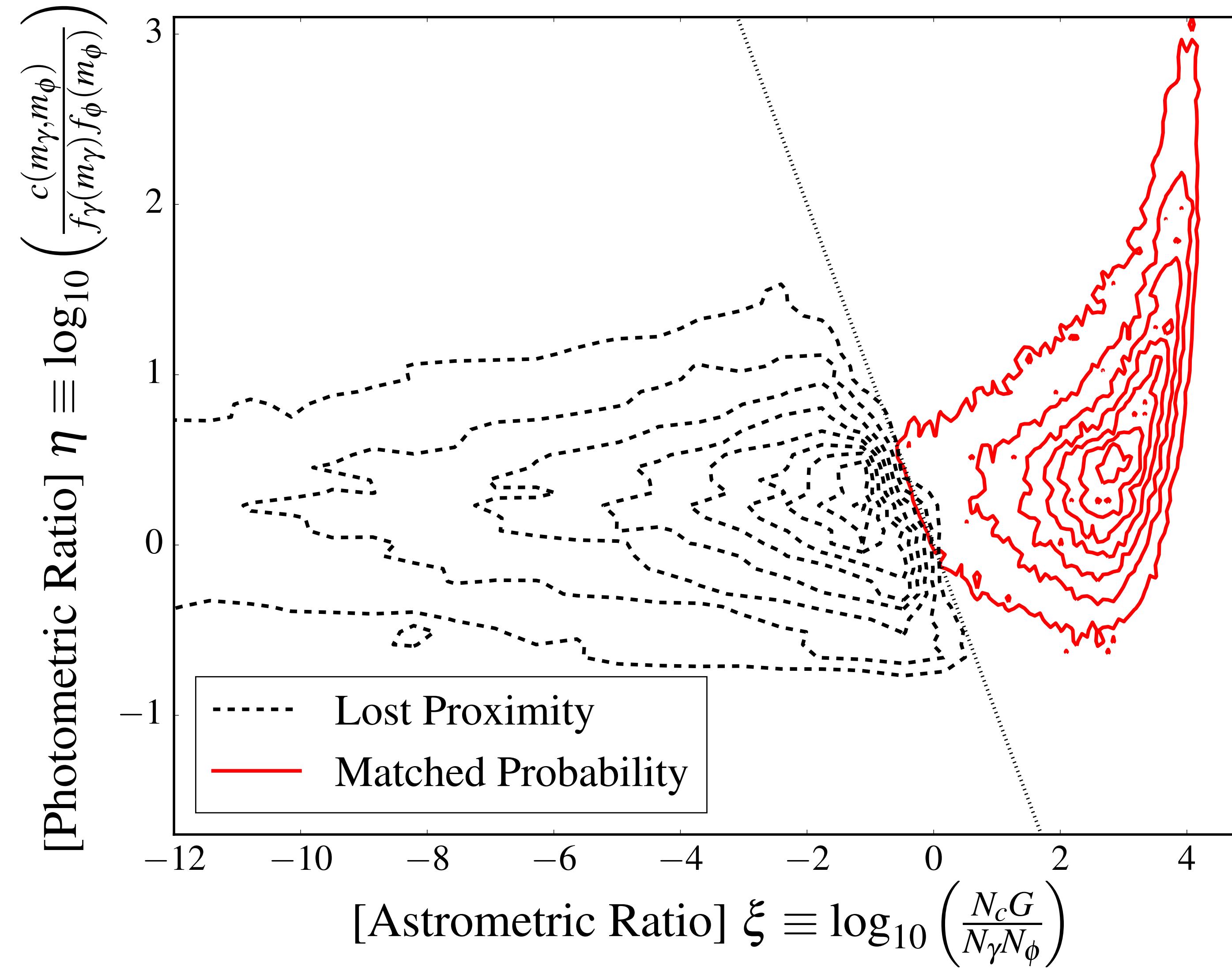


Wilson & Naylor (2018b)
WISE - Wright et al. (2010)
Gaia matches - Marrese et al. (2019)
Gaia DR2 - Gaia Collaboration, Brown A. G. A., et al. (2018)

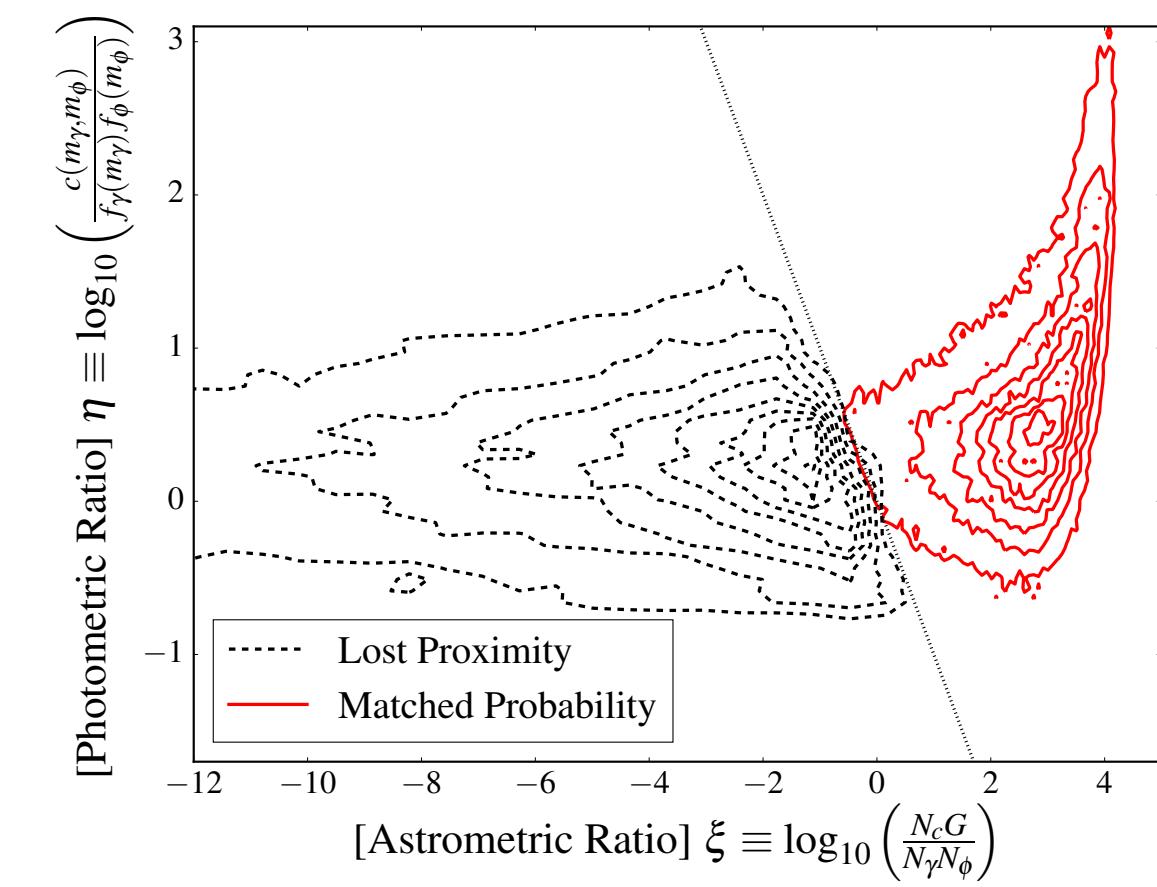
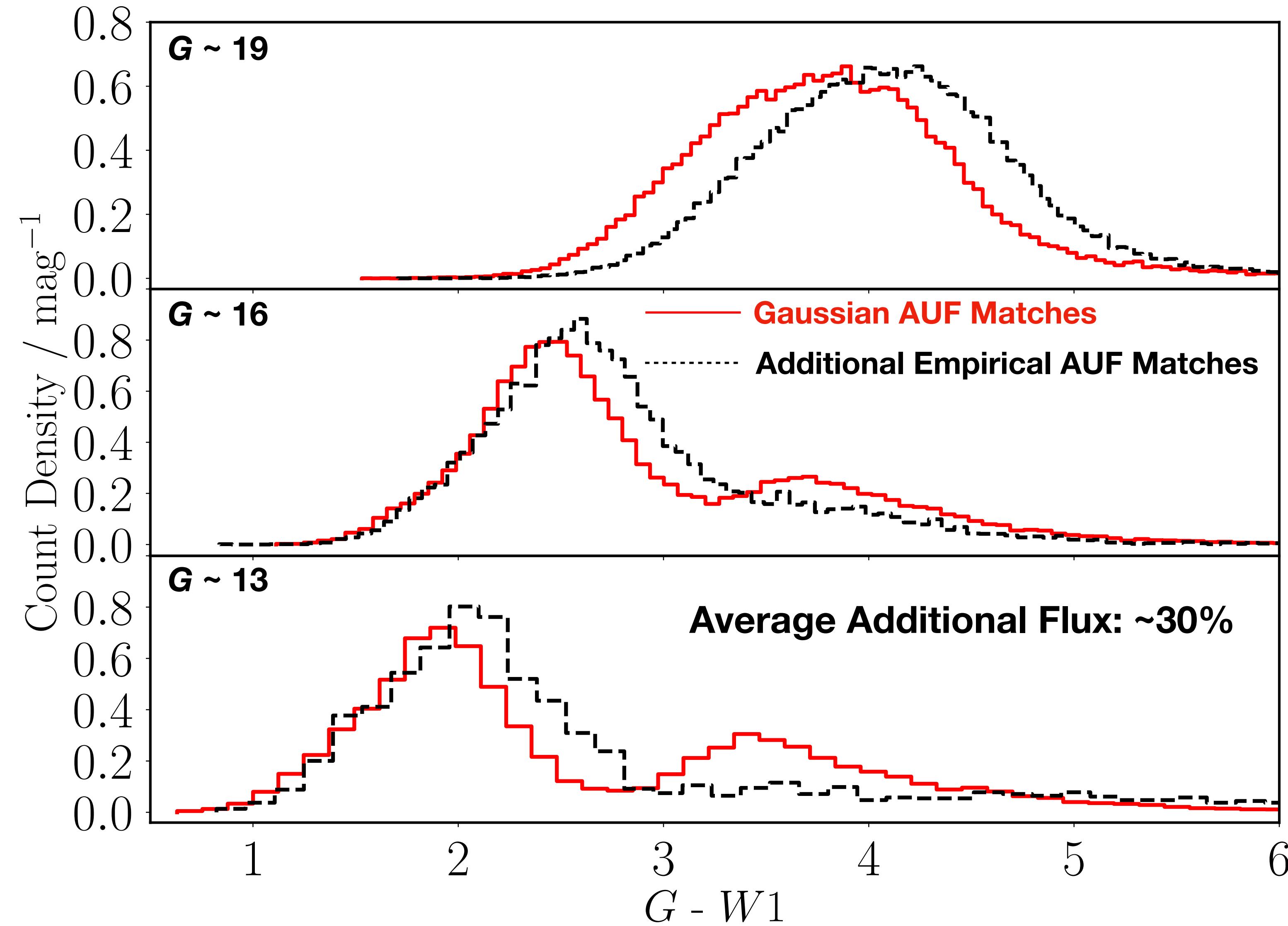


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Lost Gaussian-Only Matches

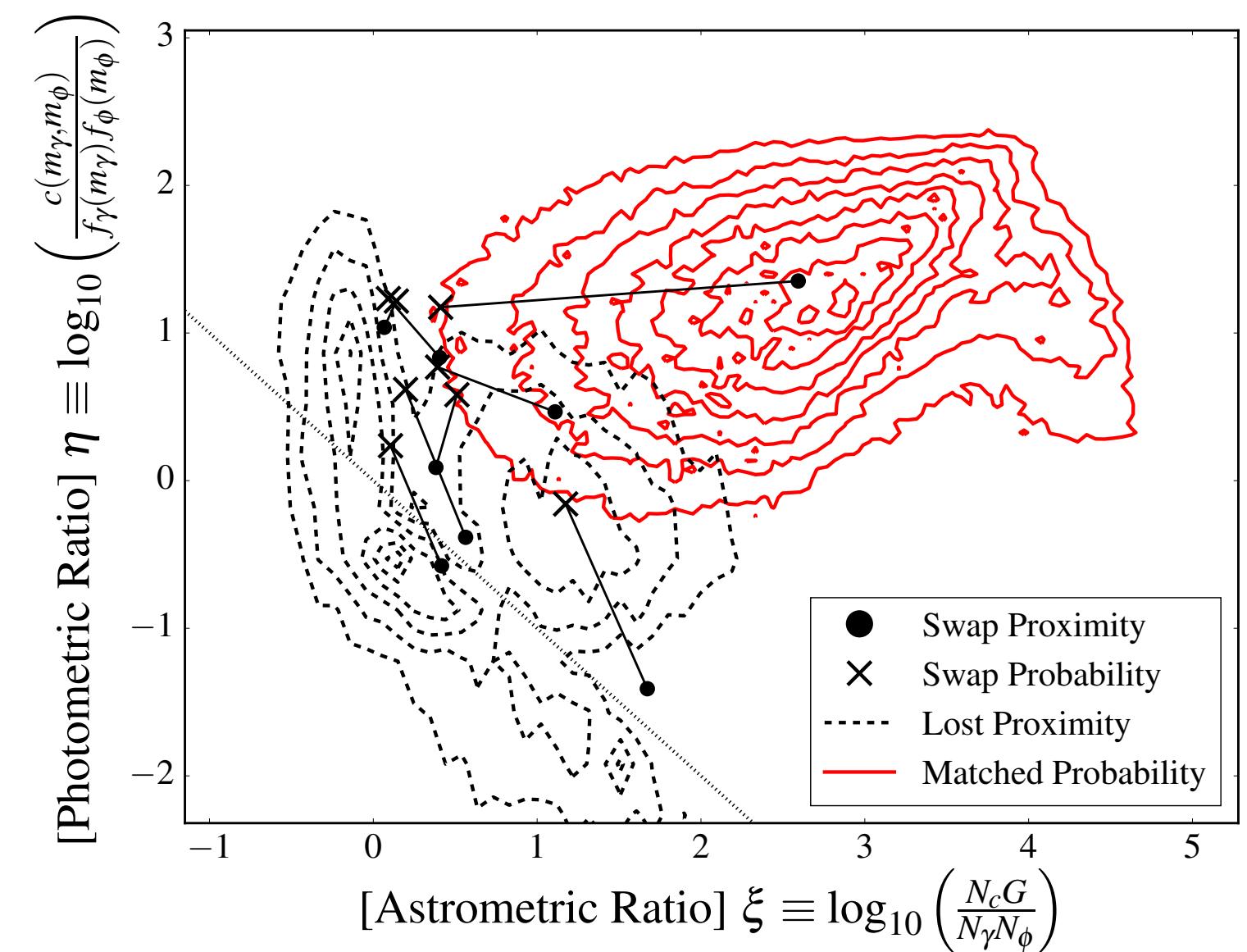
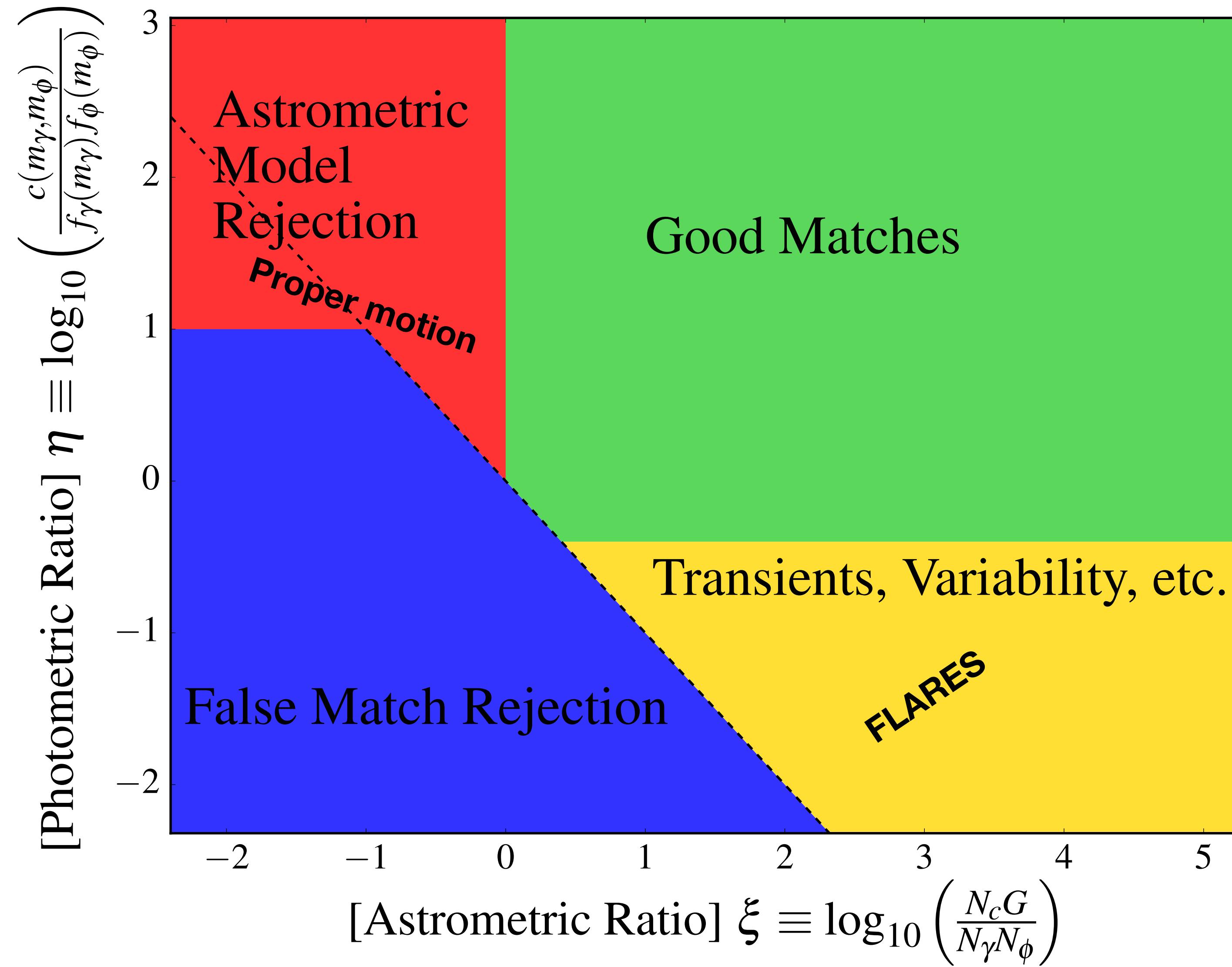


Perturbation-Colour Correlation

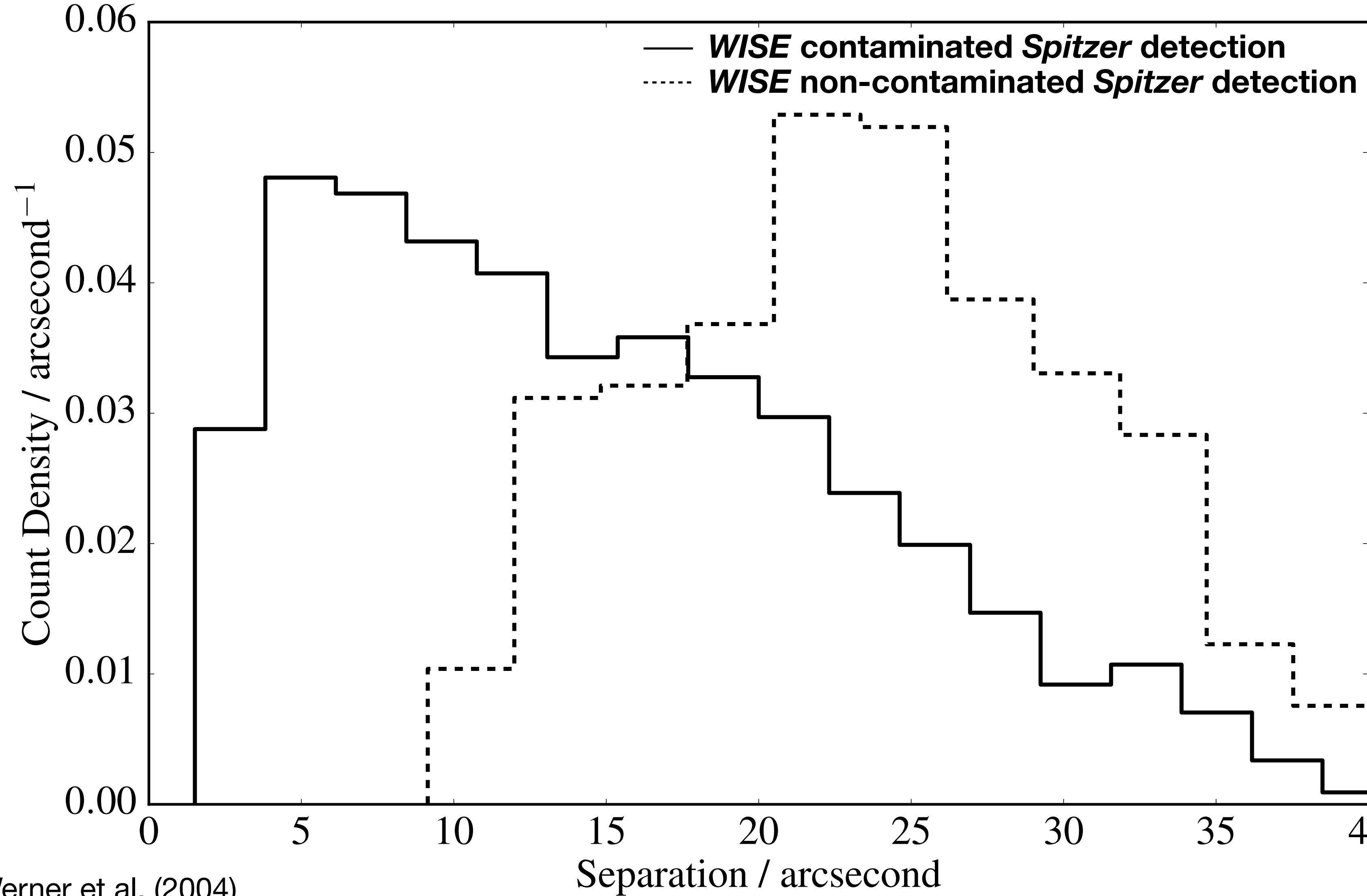


"Extra flux" has an impact on derived proper motions and parallaxes, and IR excesses!

The Likelihood Ratio Space



Resolving Contaminants



Spitzer - Werner et al. (2004)

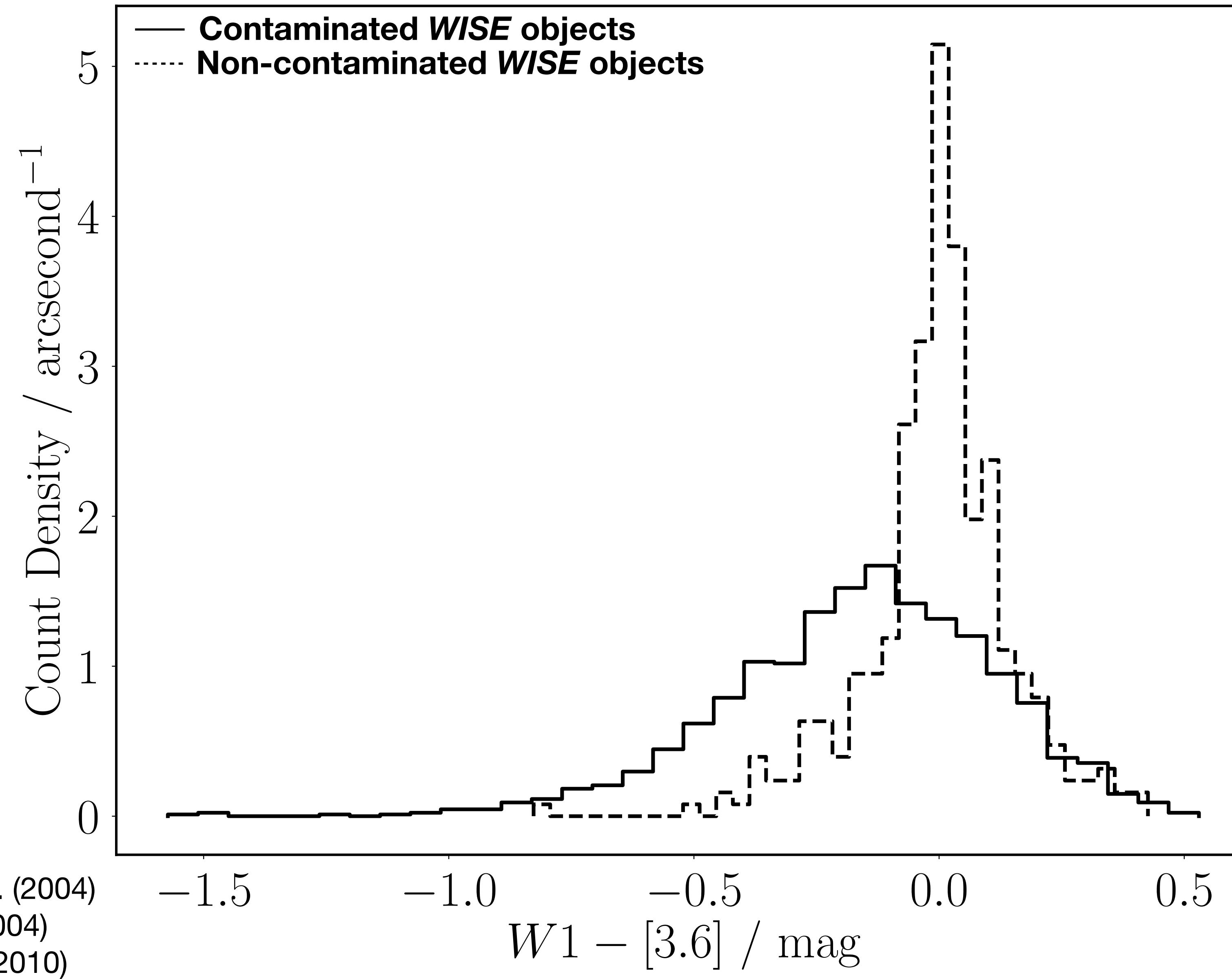
IRAC - Fazio et al. (2004)

WISE - Wright et al. (2010)

Wilson & Naylor (2018b)

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Resolving Contaminant Flux

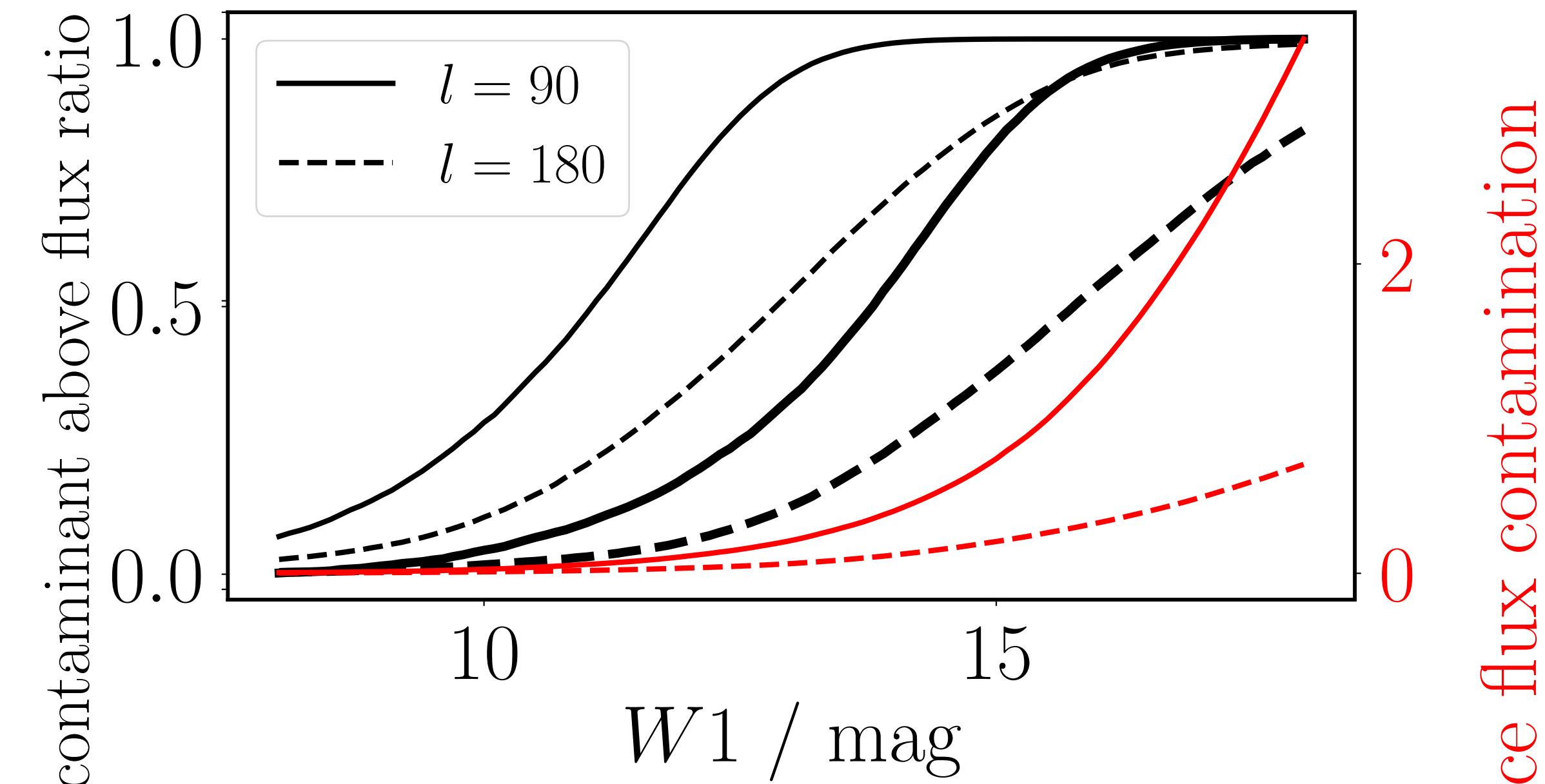


Spitzer - Werner et al. (2004)
IRAC - Fazio et al. (2004)
WISE - Wright et al. (2010)
Wilson & Naylor (2018b)

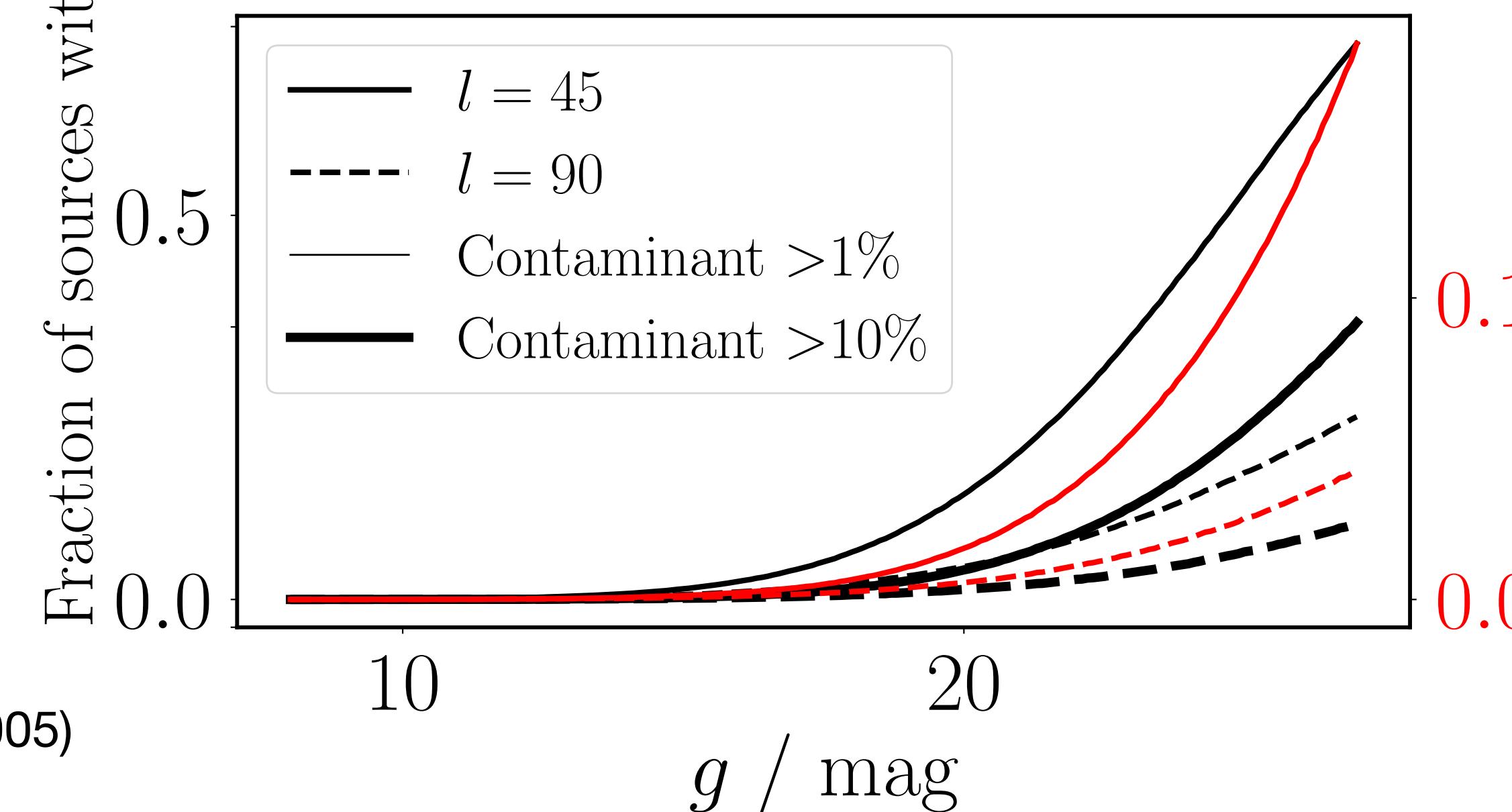
$W1 - [3.6]$ / mag

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Contamination Rates and Amounts



Total relative source flux contamination



Open Source Code: Macauff

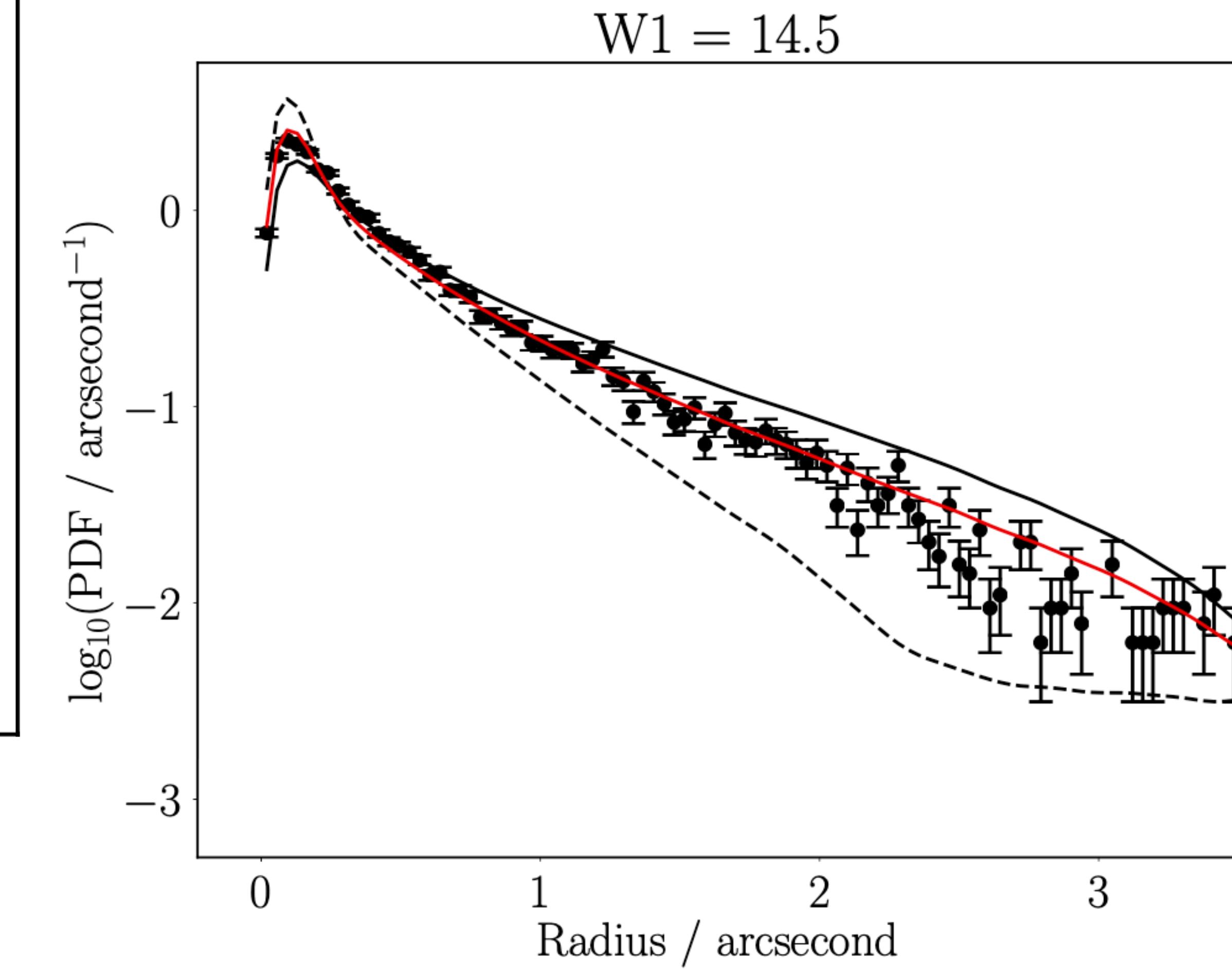
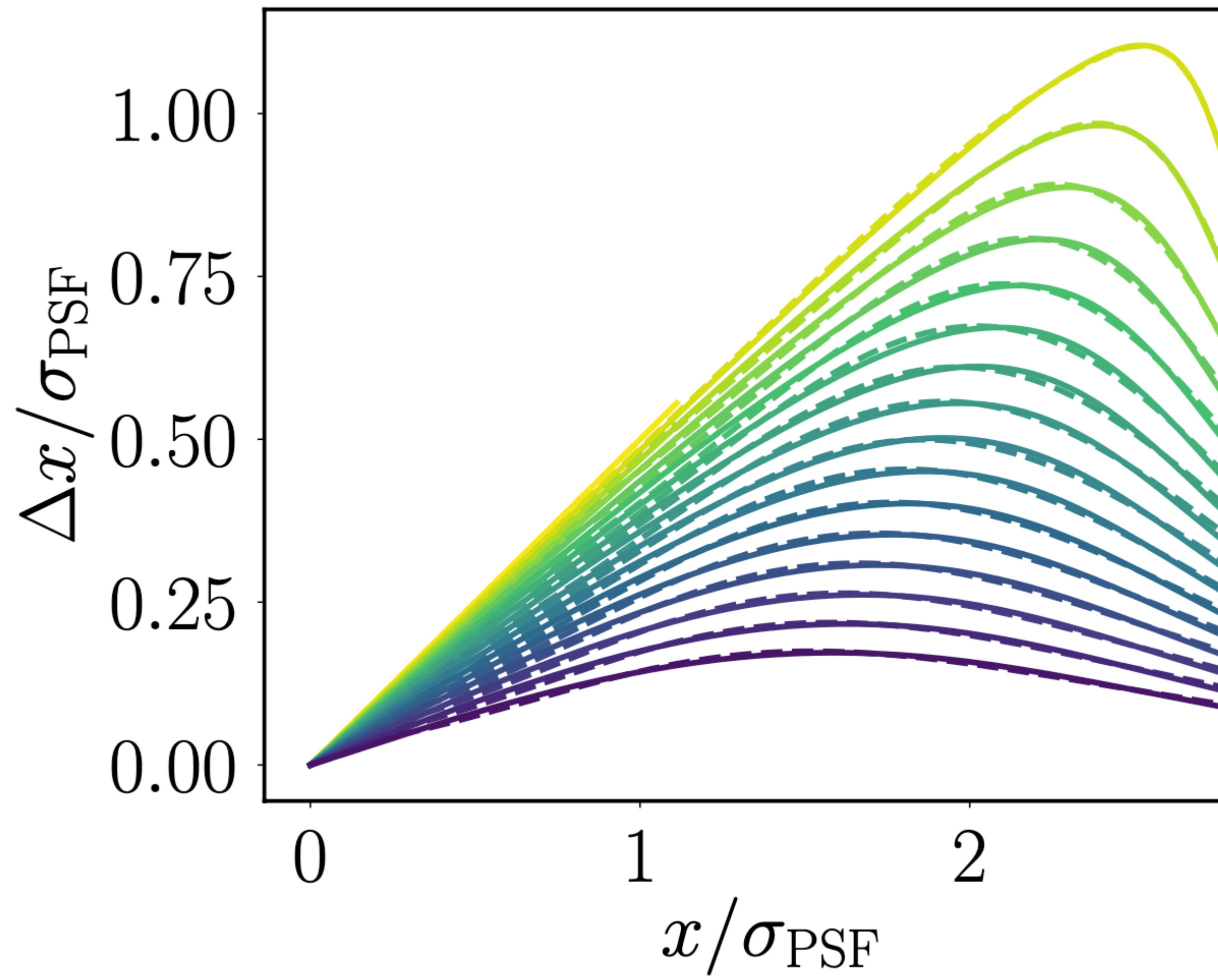
Matching Across Catalogues using the Astrometric Uncertainty Function and Flux



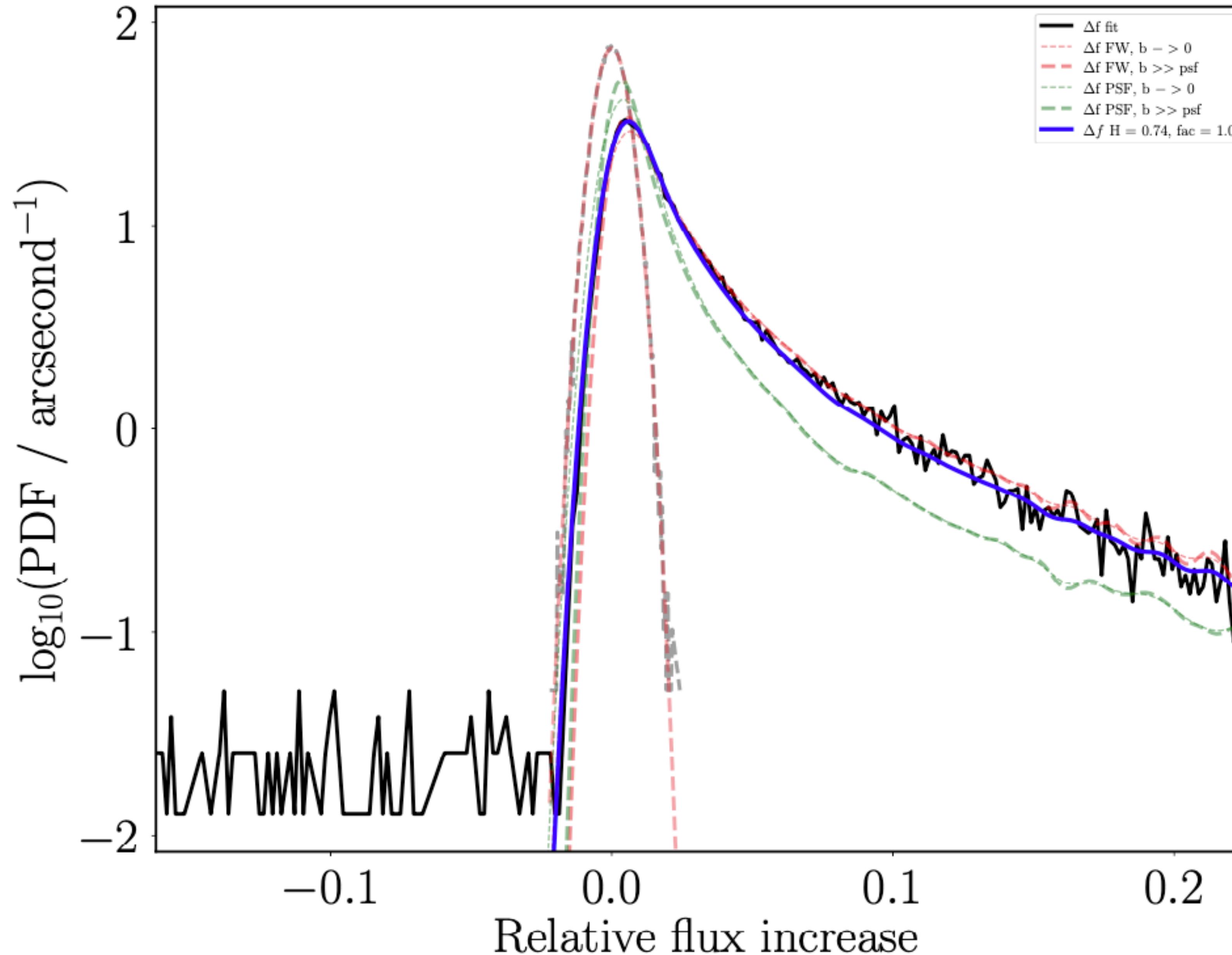
(Points if you know your tartans!)

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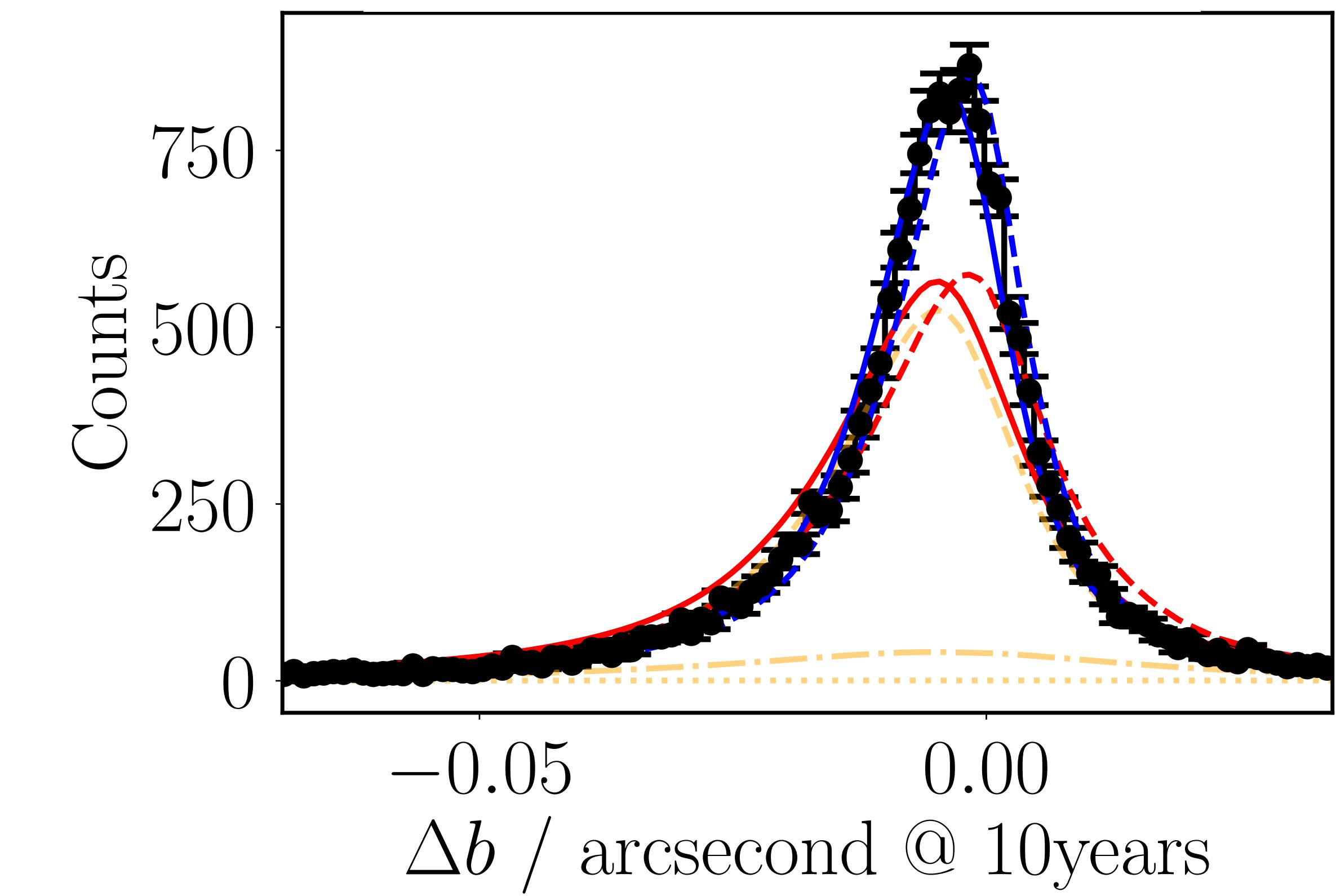
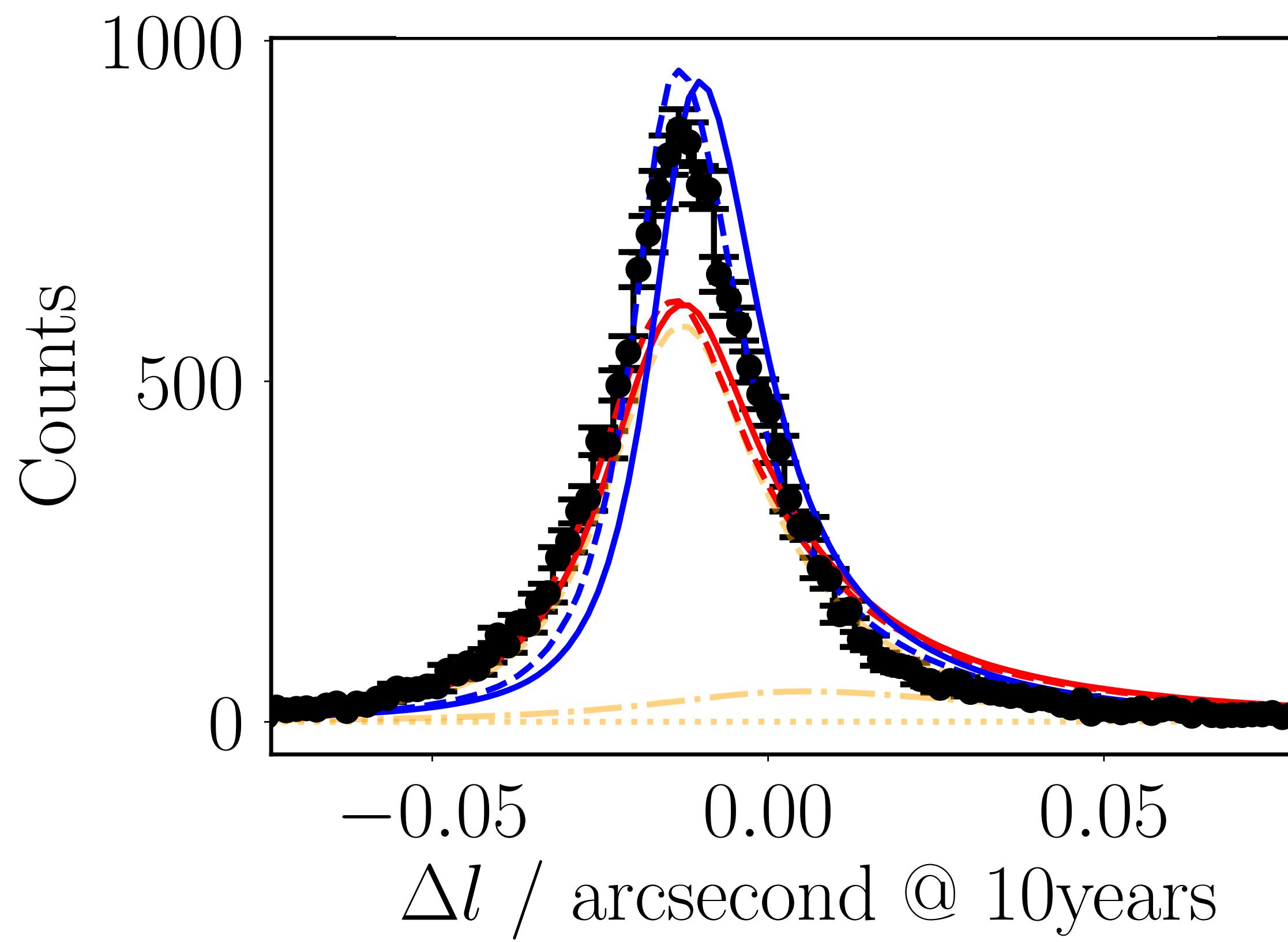
Probing the Faintest Sources



Photometric Contamination Function



Including Unknown Proper Motions



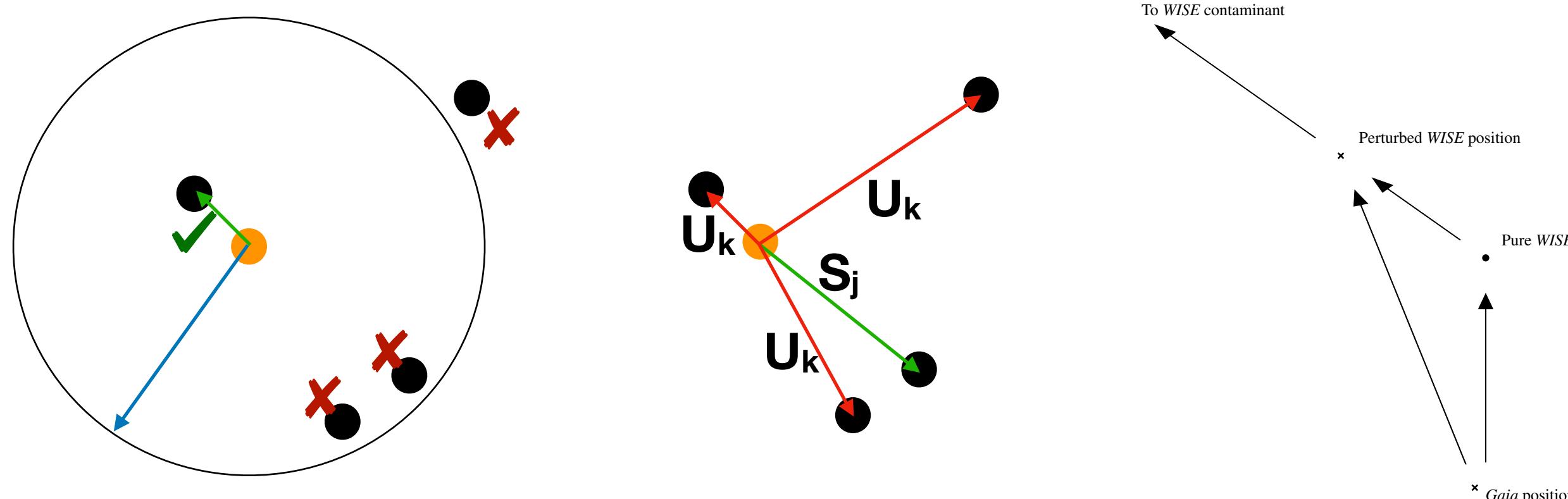
What does this mean for you?

The “busy” astronomer: uses a quick and simple 2” match -> Too many matches

The “Bayesian” astronomer: uses astrometric centroid uncertainty to reduce match radius -> Too few matches

The “careful” astronomer: includes perturbation from blended objects in the AUF -> Correct number of matches

**The “smart” astronomer: uses our cross-matches to get the correct number of matches
and information on how much flux contamination is affecting their object!**

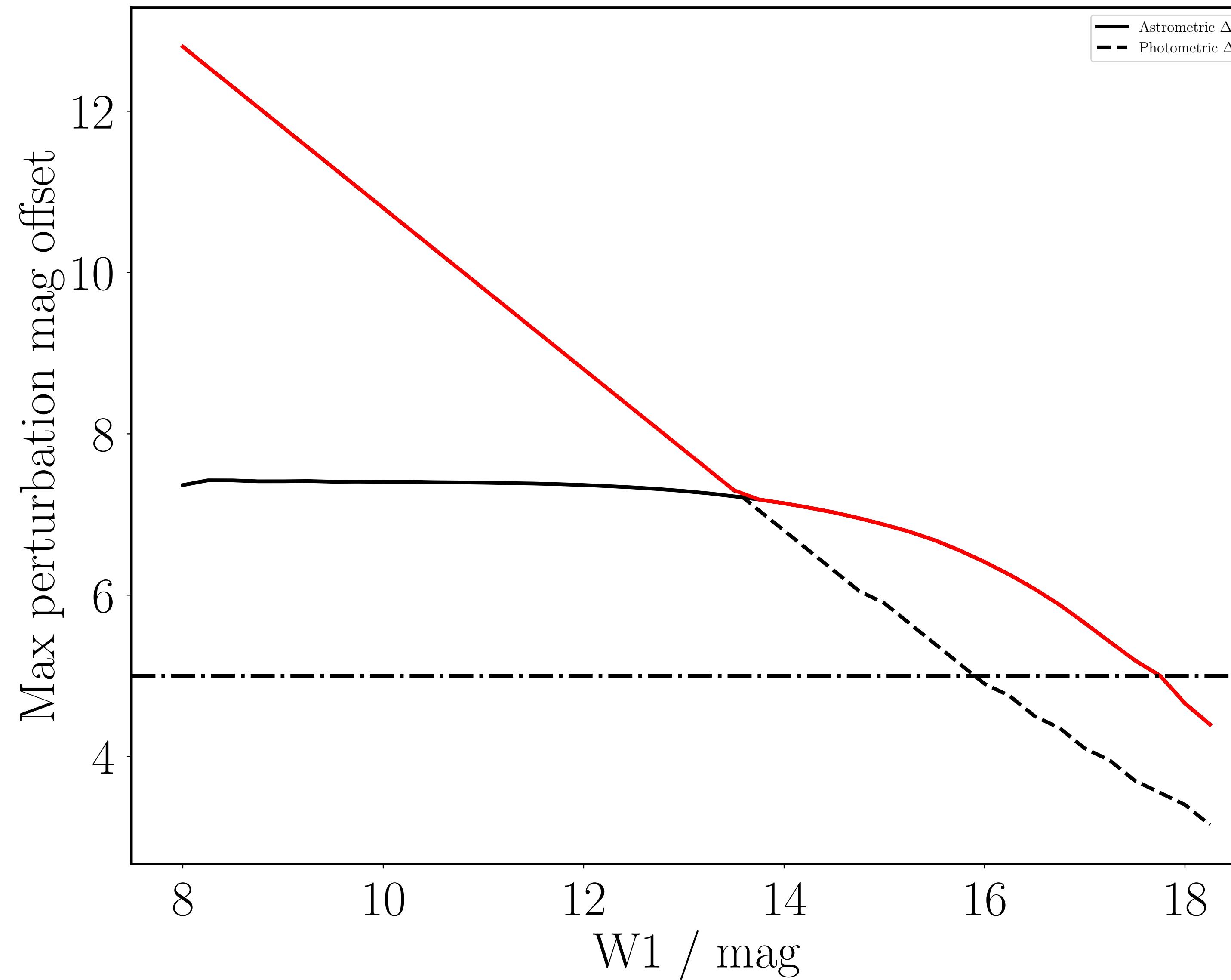


you downloading your favourite
cross-matches, probably

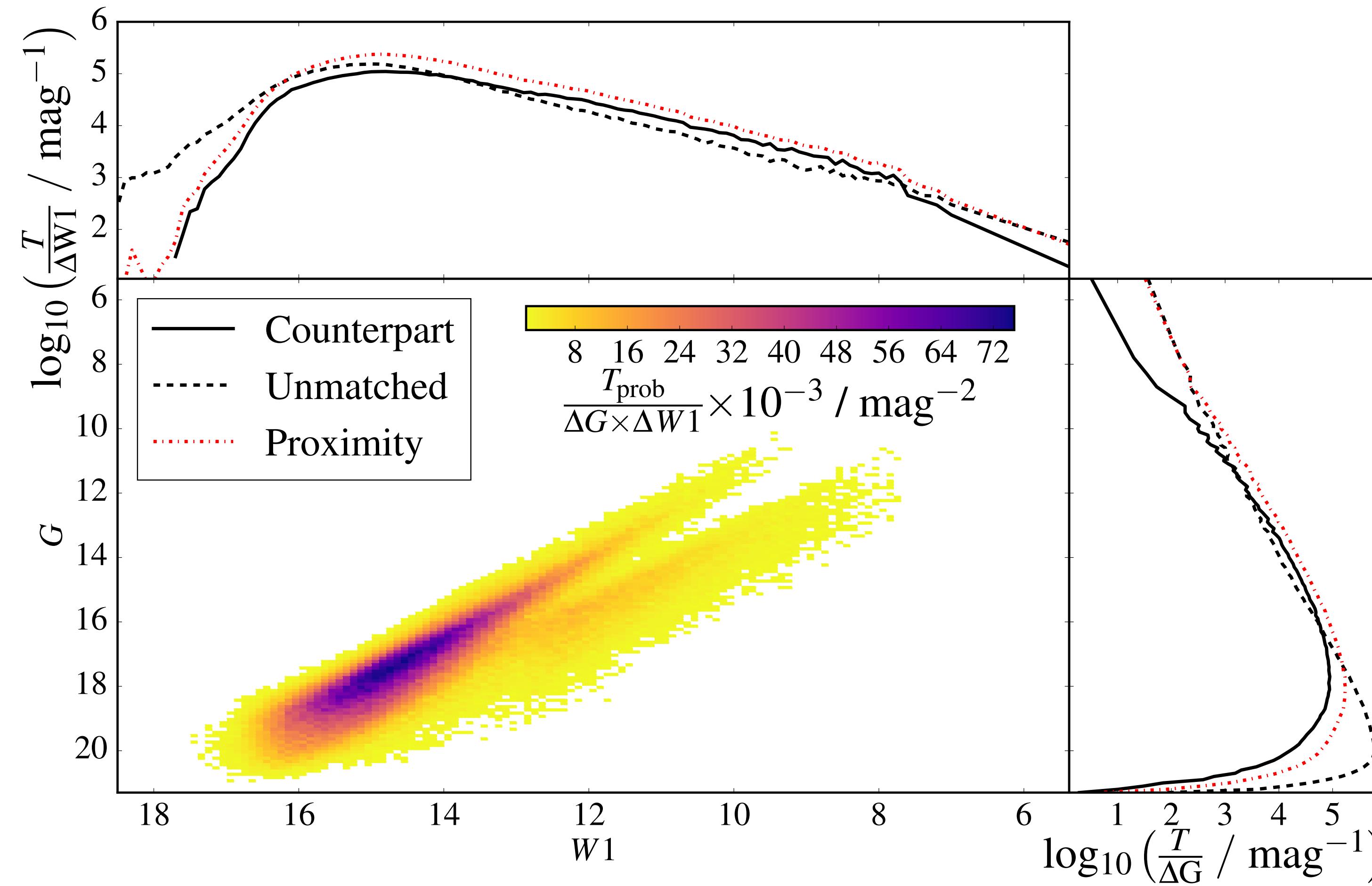
Conclusions

- Blended star contamination causes positional shifts, now modelled robustly for the first time in the AUF
- Symmetric data-driven photometric likelihood now possible
- *WISE* objects are up to 30% flux contaminated
- LSST will suffer of order 10% flux contamination in the future
 - Important for extinction/distance; “1% photometry”?
- Can include unknown proper motions easily in AUF
- Modelling of statistical flux contamination allows for the recovery of “true” fluxes
- High dynamic range matches must account for differential crowding matching to ancillary or historic data
- Upcoming LSST:UK cross-match service macauff – let me know your thoughts/needs/hopes/dreams

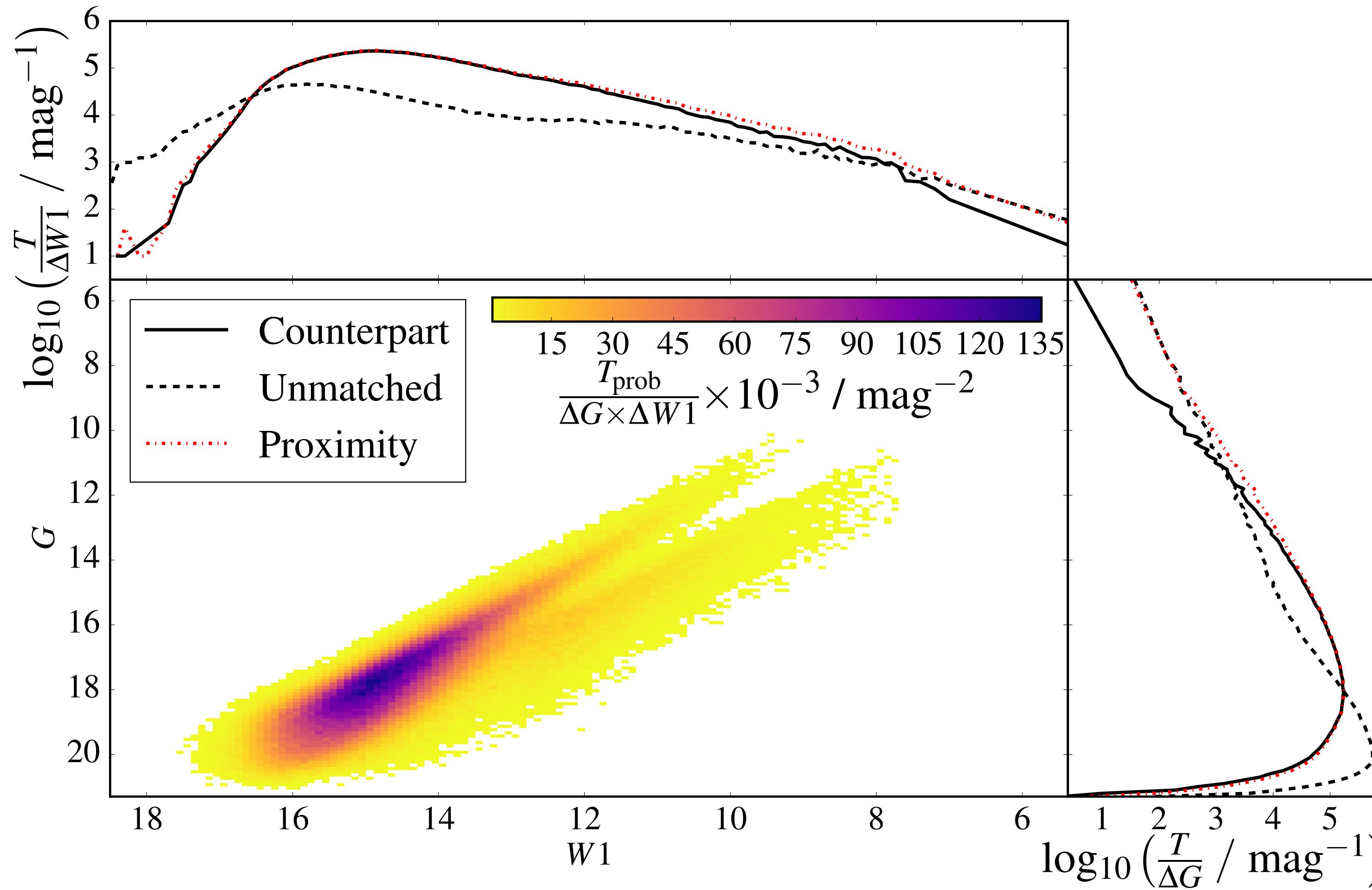
The Astrometric Uncertainty Function and LSST: A Crisis of Completeness Limit



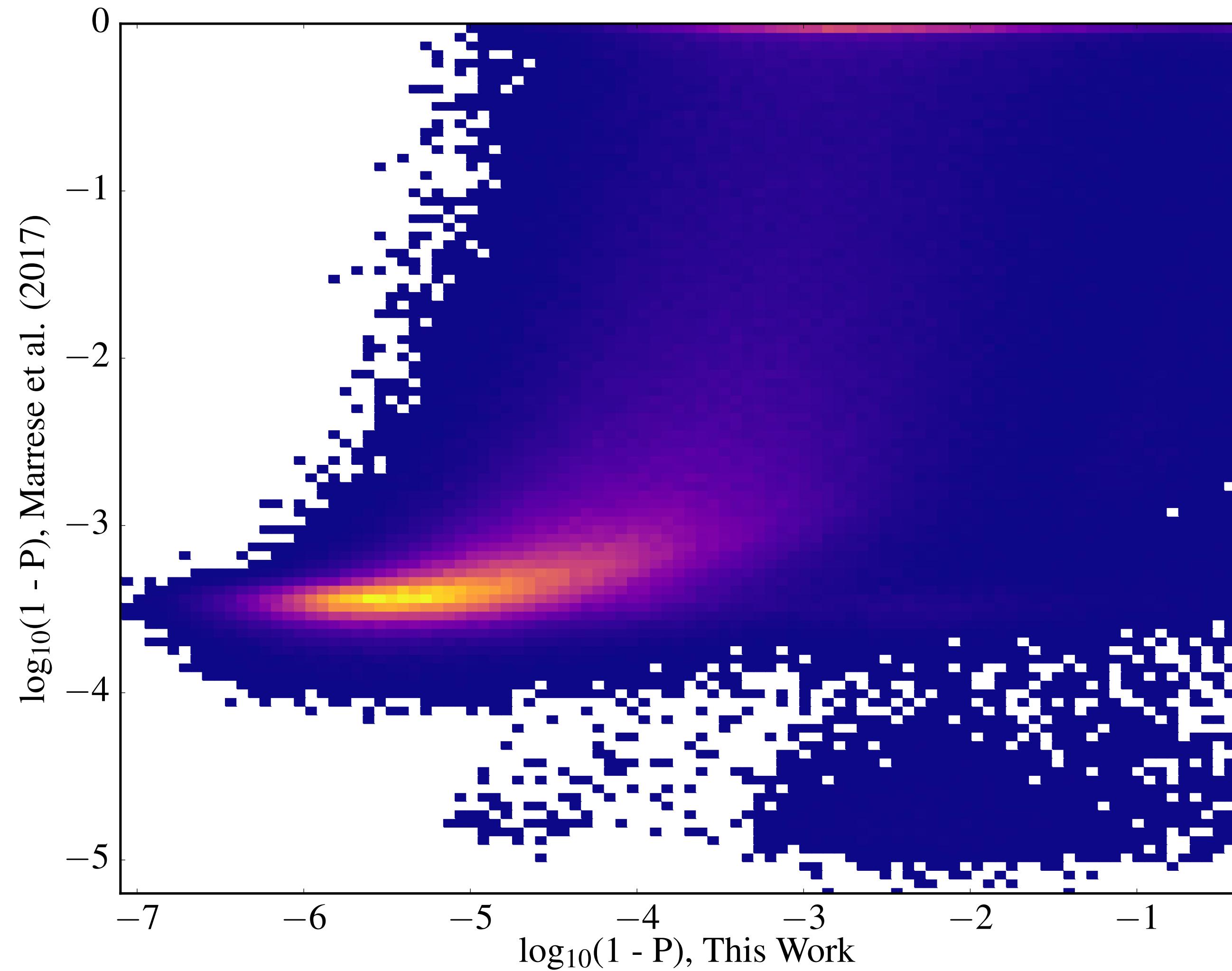
Contamination Effects: Gaia-WISE Gaussian Matches



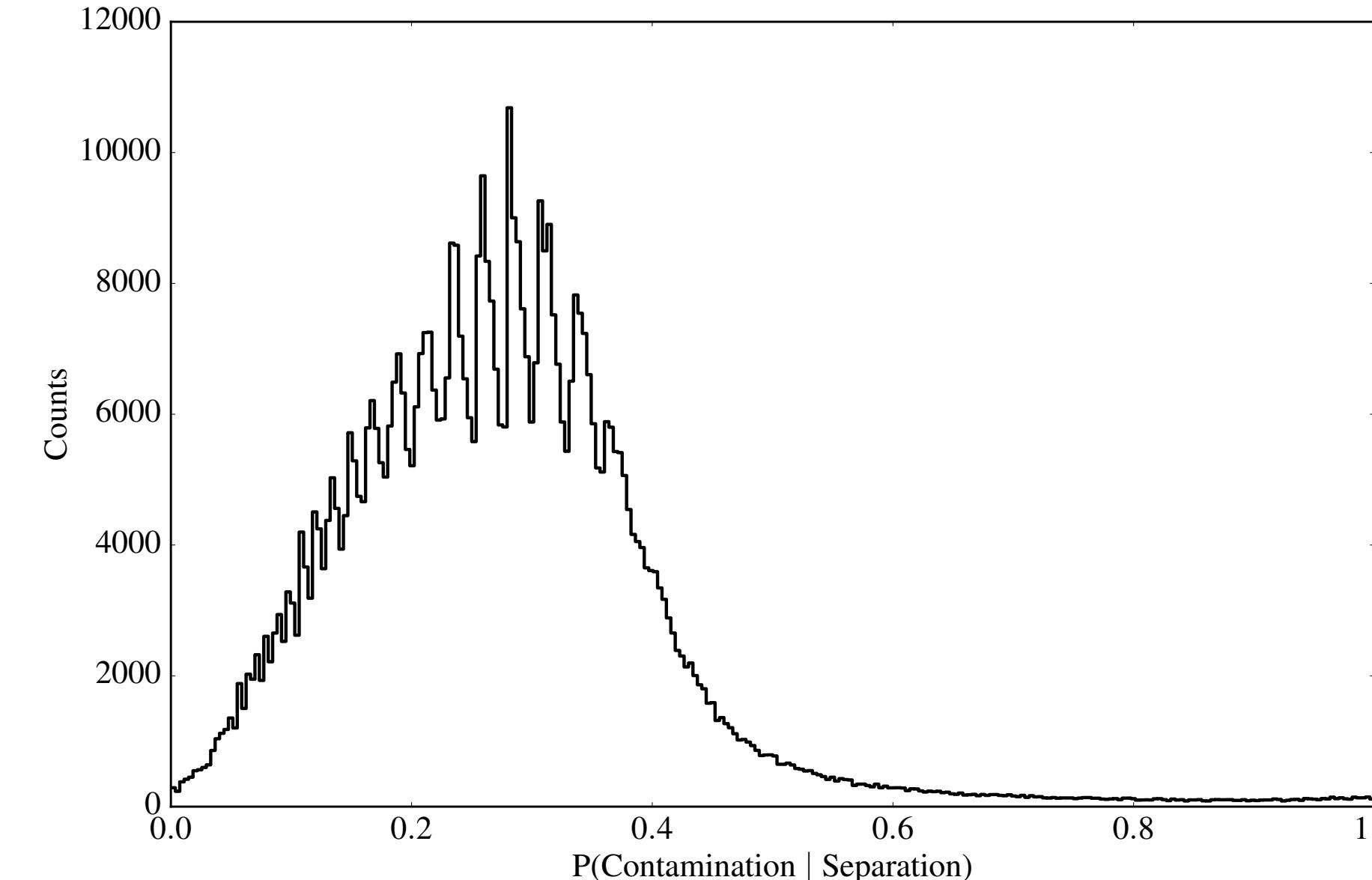
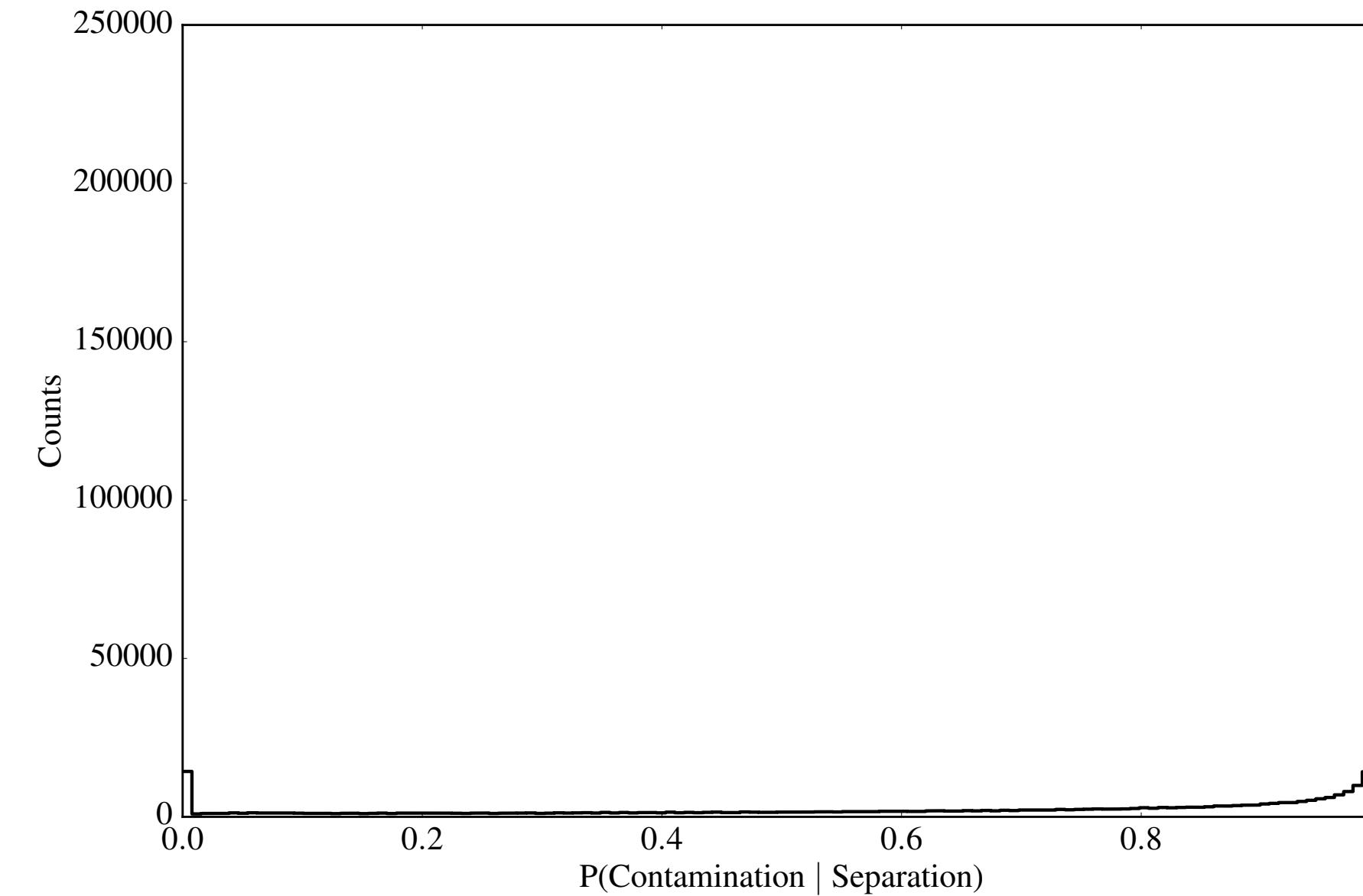
Contamination Effects: Gaia-WISE Empirical Matches



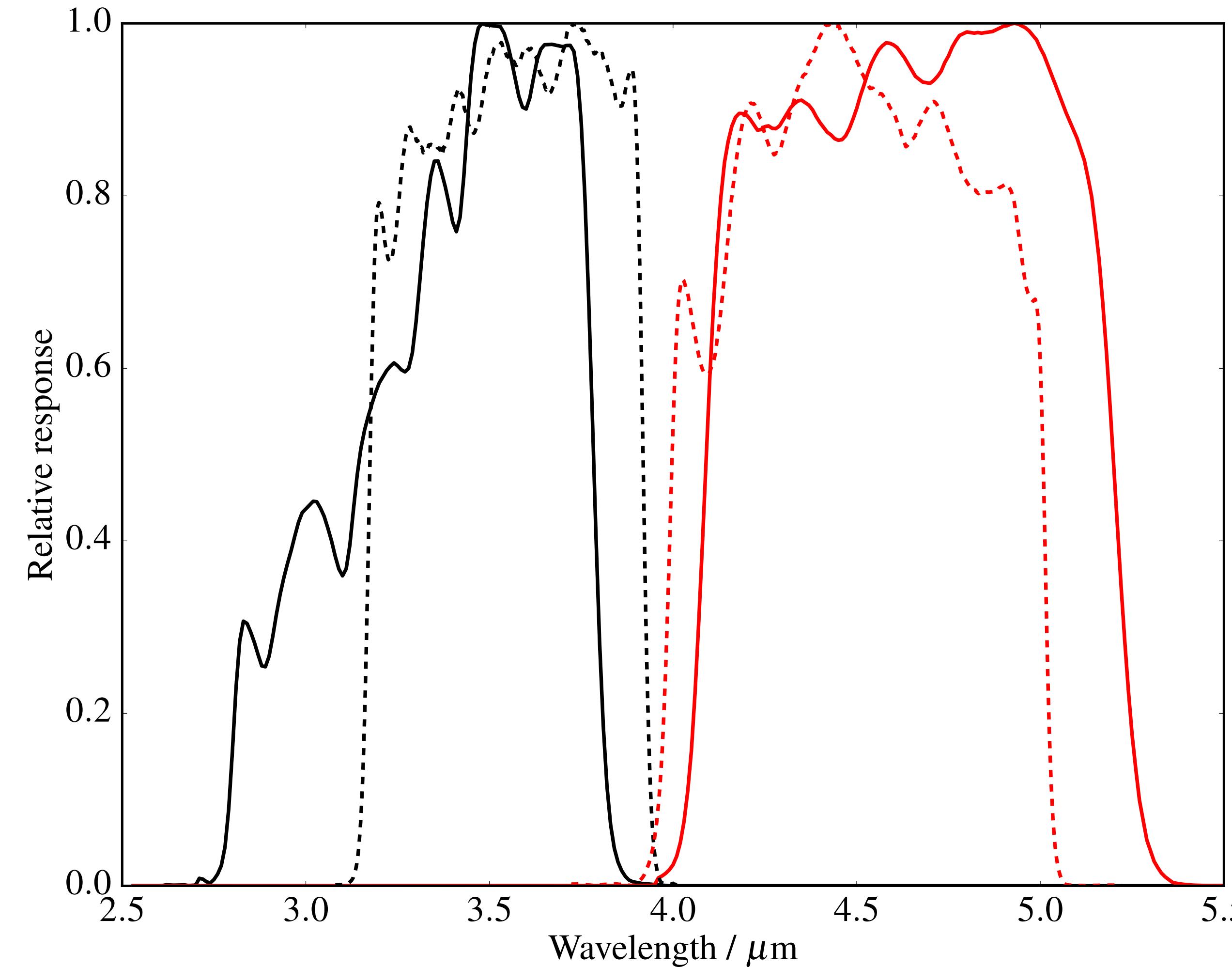
Contamination Effects: Gaia Lost Matches



Photometric Contamination: WISE/Spitzer Contamination %



Contamination Effects: Wavelength Coverage



The Astrometric Uncertainty Function: Analytical perturbations

