

Image Processing Algorithms

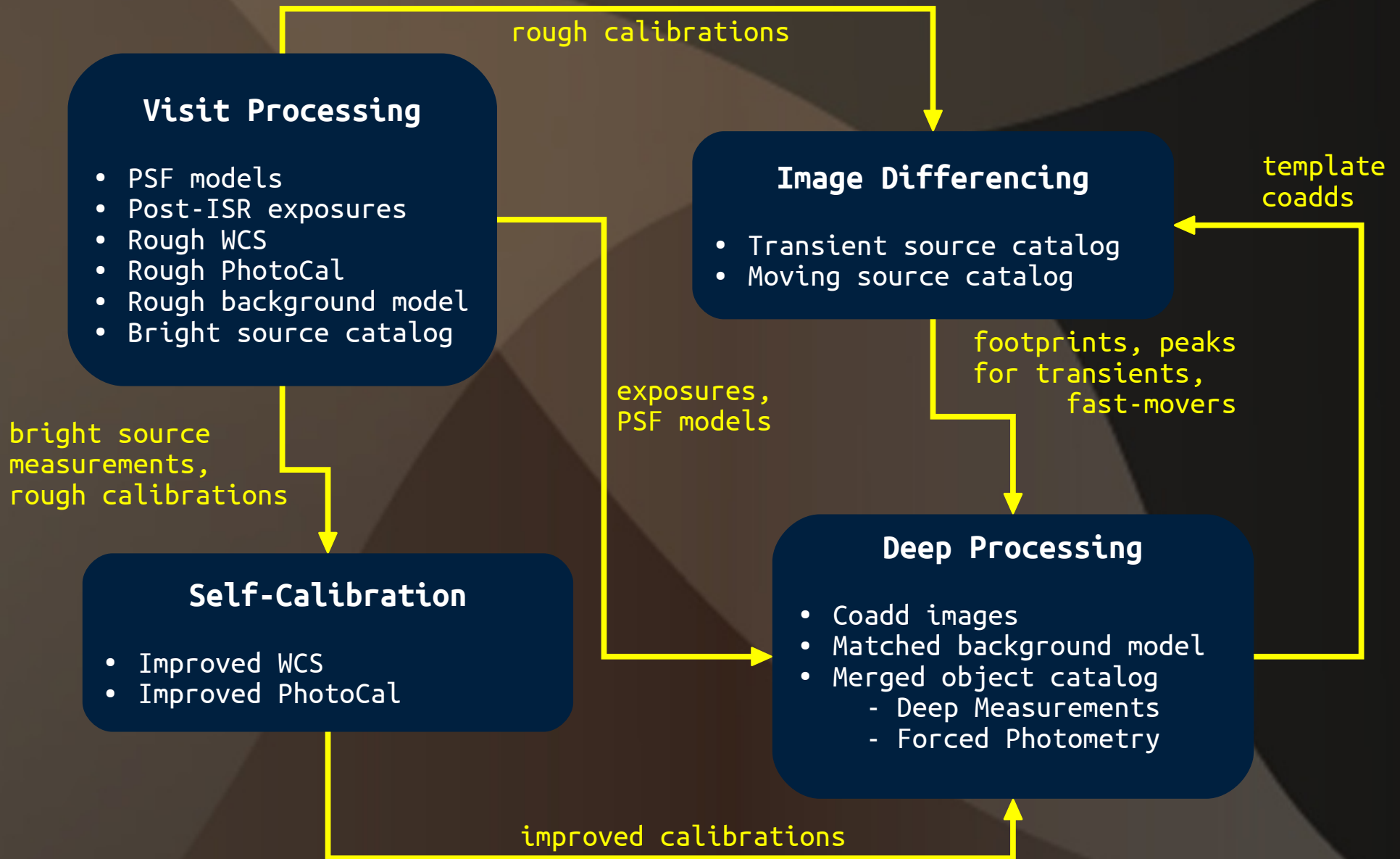
Jim Bosch

DE School / SLAC / 2015-02-02

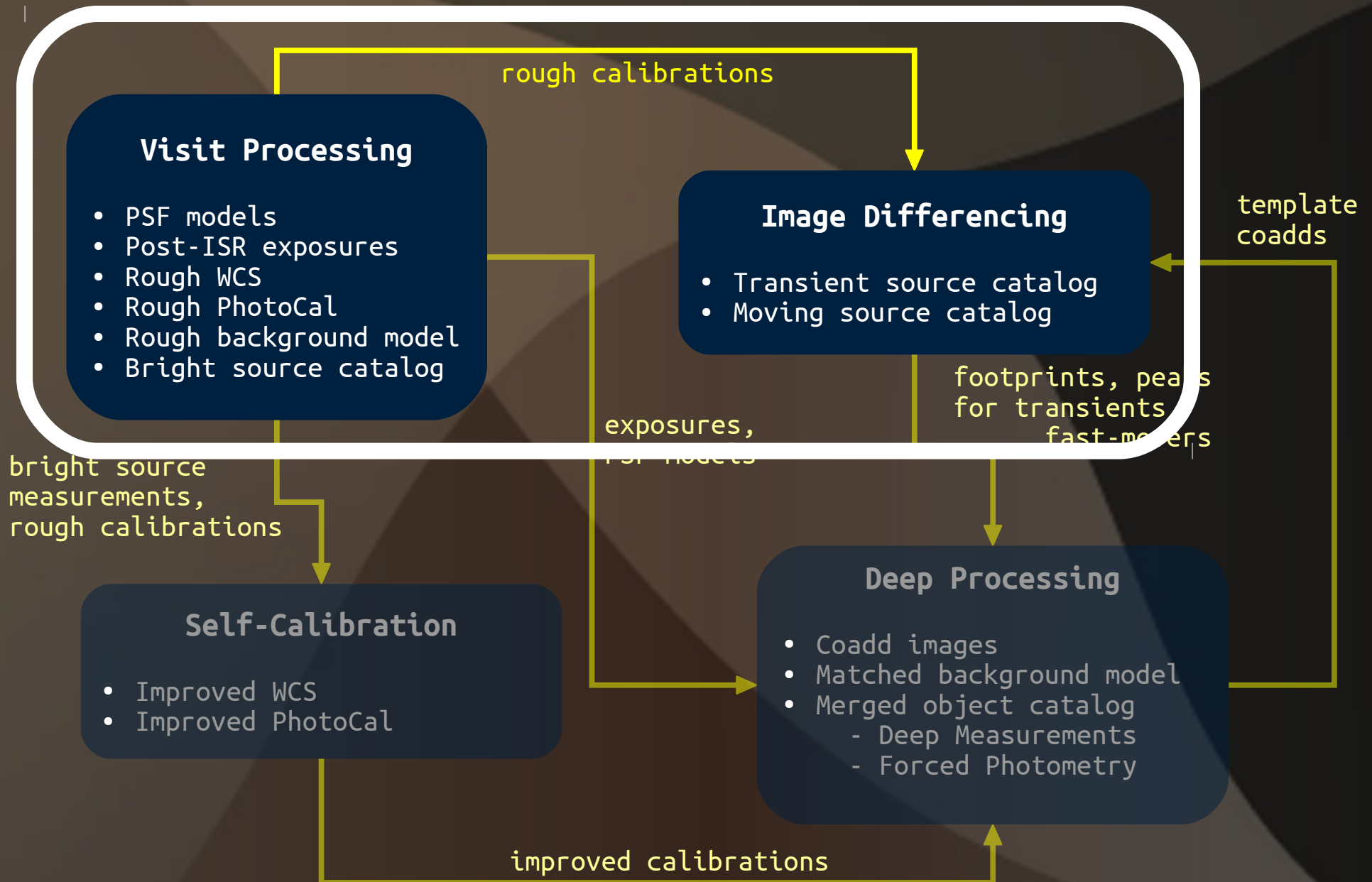
Image Processing Algorithms

- Overview of LSST Pipelines
 - Visit Processing
 - Self-Calibration
 - Image Differencing
 - Deep Processing
- Detection Algorithms
 - The Simple Case
 - Brainstorming Extensions
- Kinds of Coadds

LSST Data Release Pipeline (Yearly)



LSST Alert Pipeline (Nightly)



Visit Processing

CCD N

CCD N+1

Snap 1

Snap 2

ISR

ISR

Combine Snaps

Snap 1

Snap 2

ISR

ISR

Combine Snaps

Background Estimation

Detection

Detection

Deblending

Deblending

Measurement

Measurement

Astrometric Calibration (WCS Modeling)

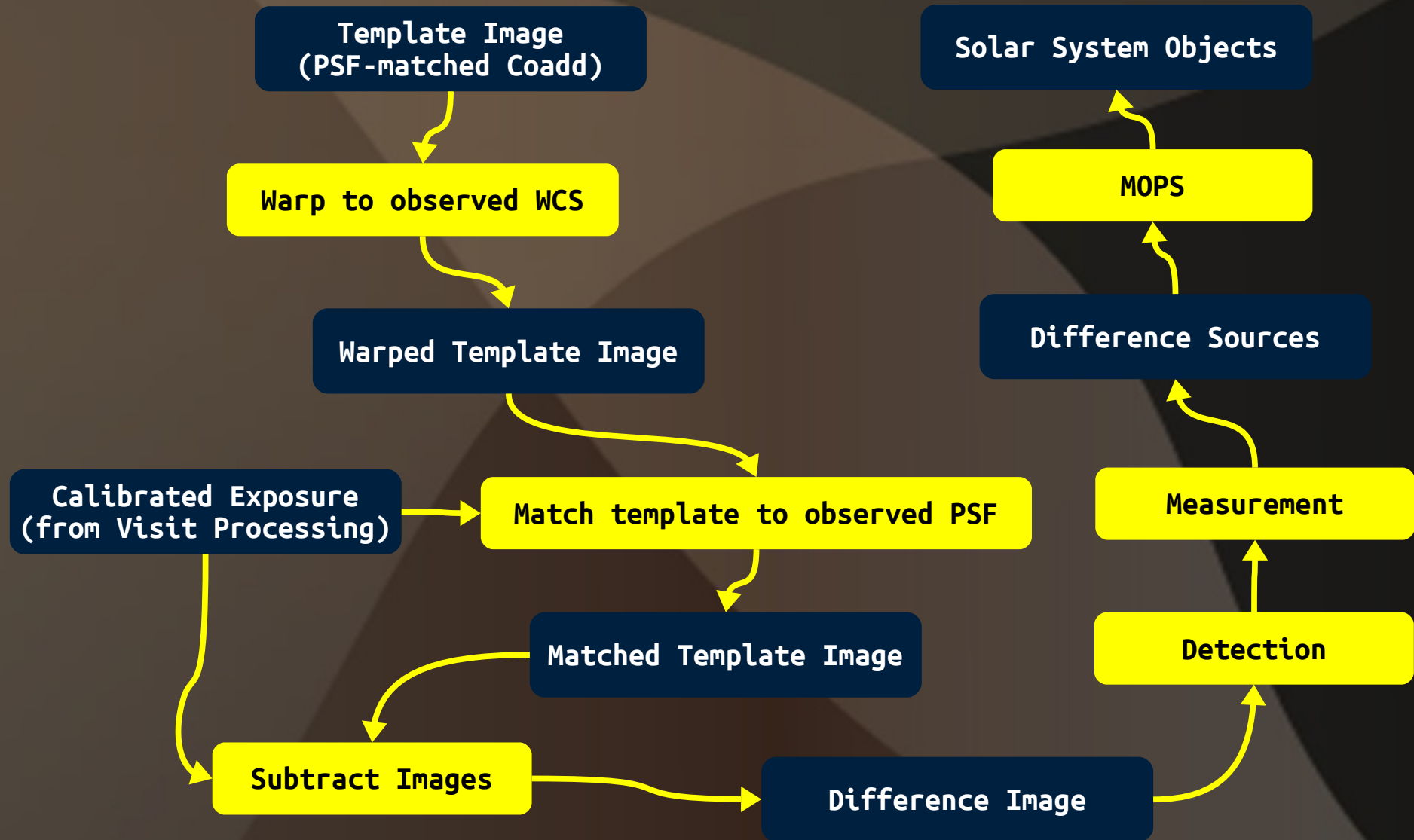
PSF Modeling

Photometric Calibration, Aperture Corrections

Iterate!

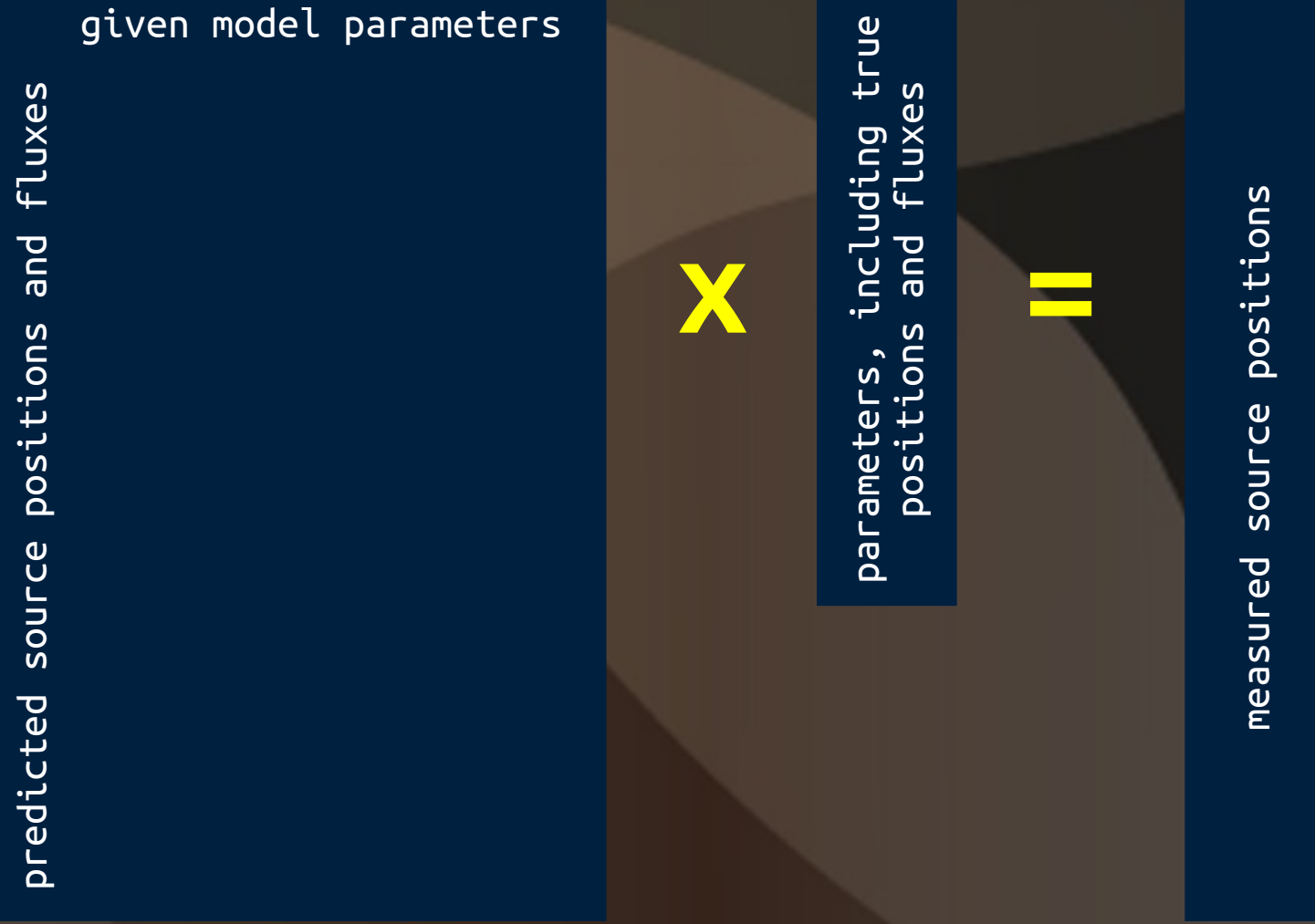


Image Differencing



All of this is one CCD at a time

Self-Calibration



Deep Processing

- Improve background models
 - demand consistency between epochs (“background matching”)
- Detect sources at full depth
 - probably use some coadds; which ones?
- Associate deep sources with transient/moving sources
- Deblend everything consistently
 - probably use some coadds; which ones?
- Measure everything consistently
 - start on coadds
 - tune up fitting using all epochs together

Detection: Point Source, Single Image

- We want to find all point sources in an image with known Gaussian noise and known PSF, above a certain *significance*.
- Here's our log-likelihood:

$$L(\alpha, x, y) = -\ln P(I|x, y, \alpha)$$
$$= \frac{1}{2} \sum_{i,j} \left[\frac{I_{i,j} - \alpha p_{i,j}(x, y)}{\sigma_{i,j}} \right]^2$$

pixel indices

Detection: Point Source, Single Image

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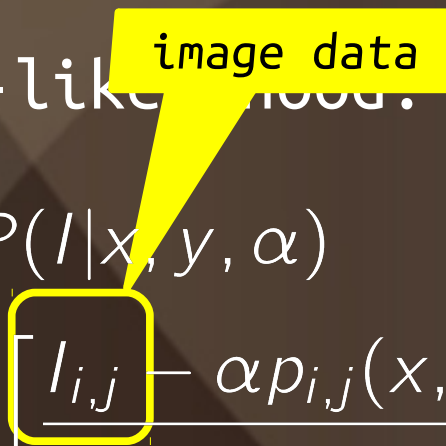
source position our log-likelihood:

$$L(\alpha, x, y) = -\ln P(I|x, y, \alpha)$$
$$= \frac{1}{2} \sum_{i,j} \left[\frac{I_{i,j} - \alpha p_{i,j}(x, y)}{\sigma_{i,j}} \right]^2$$

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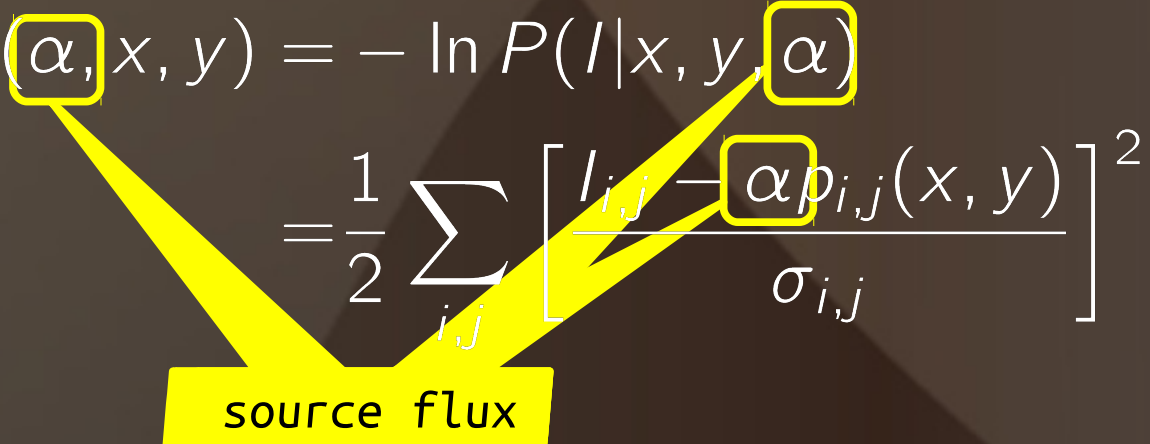
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source flux



Detection: Point Source, Single Image

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- Here's our log-likelihood:


$$L(\alpha, x, y) = -\ln P(I|x, y, \alpha)$$

$$= \frac{1}{2} \sum_{i,j} \left[\frac{l_{i,j} - \alpha p_{i,j}(x, y)}{\sigma_{i,j}} \right]^2$$

PSF model

Detection: Point Source, Single Image

- We want to find all point sources in an image with known Gaussian noise and known PSF, above a certain *significance*.
- Here's our log-likelihood:

$$L(\alpha, x, y) = -\ln P(I|x, y, \alpha)$$
$$= \frac{1}{2} \sum_{i,j} \left[\frac{l_{i,j} - \alpha p_{i,j}(x, y)}{\sigma_{i,j}} \right]^2$$


pixel noise

Detection: Point Source, Single Image

$$L(\alpha, x, y) = \frac{1}{2} \sum_{i,j} \left[\frac{l_{i,j} - \alpha p_{i,j}(x, y)}{\sigma_{i,j}} \right]^2$$

Detection: Point Source, Single Image

$$L(\alpha, x, y) = \frac{\alpha^2}{2} \Phi(x, y) - \alpha \Psi(x, y) + K$$

$$\Phi(x, y) = \sum_{i,j} \frac{[p_{i,j}(x, y)]^2}{\sigma_{i,j}^2}$$

$$\Psi(x, y) = \sum_{i,j} \frac{l_{i,j} p_{i,j}(x, y)}{\sigma_{i,j}^2}$$

Detection: Point Source, Single Image

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$$\frac{\partial L(\alpha, x, y)}{\partial \alpha} = \alpha \Phi(x, y) - \Psi(x, y) = 0$$

$$\alpha_{ML} = \frac{\Psi(x, y)}{\Phi(x, y)}$$

Detection: Point Source, Single Image

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$$\alpha_{ML} = \frac{\Psi(x, y)}{\Phi(x, y)}$$

$$L_{ML} = \frac{\Psi^2(x, y)}{2\Phi(x, y)}$$

Detection: Point Source, Single Image

Log-Likelihood for an object with flux α_{ML} :

$$L_{ML} = \frac{[\psi(x, y)]^2}{2\Phi(x, y)} + K$$

Log-likelihood for zero flux: $L_0 = K$

Likelihood ratio:
$$\frac{e^{-L_{ML}(x,y)}}{e^{-L_0}} = e^{-[\nu(x,y)]^2/2}$$

Significance:
$$\nu(x, y) = \frac{\psi(x, y)}{\sqrt{\Phi(x, y)}}$$

Detection: Point Source, Single Image

$$L(\alpha, x, y) = \frac{\alpha^2}{2} \Phi(x, y) - \alpha \Psi(x, y) + K$$

$$\Phi(x, y) = \sum_{i,j} \frac{[p_{i,j}(x, y)]^2}{\sigma_{i,j}^2}$$

$$\Psi(x, y) = \sum_{i,j} \frac{l_{i,j} p_{i,j}(x, y)}{\sigma_{i,j}^2}$$

correlation of
image with PSF

$$\frac{\partial L(\alpha, x, y)}{\partial \alpha} = \alpha \Phi(x, y) - \Psi(x, y) = 0$$

$$\alpha_{ML} = \frac{\Psi(x, y)}{\Phi(x, y)}$$

$$L_{ML} = \frac{\Psi^2(x, y)}{2\Phi(x, y)}$$

Detection: Point Source, Single Image

$$L(\alpha, x, y) = \frac{\alpha^2}{2} \Phi(x, y) - \alpha \Psi(x, y) + K$$

$$\Phi(x, y) = \sum_{i,j} \frac{[p_{i,j}(x, y)]^2}{\sigma_{i,j}^2}$$

depends only on
PSF model & noise

$$\Psi(x, y) = \sum_{i,j} \frac{l_{i,j} p_{i,j}(x, y)}{\sigma_{i,j}^2}$$

$$\frac{\partial L(\alpha, x, y)}{\partial \alpha} = \alpha \Phi(x, y) - \Psi(x, y) = 0$$

$$\alpha_{ML} = \frac{\Psi(x, y)}{\Phi(x, y)}$$

$$L_{ML} = \frac{[\Psi(x, y)]^2}{2\Phi(x, y)} + K$$

Detection: Point Source, Single Image

Algorithm:

- Correlate image with PSF to get $\Psi(x, y)$
- Create $\Phi(x, y)$ as an image (or pretend it's constant)
- Find regions where $\nu(x, y)$ is greater than the target threshold (“footprints”)
- Within each footprint, find peaks

Detection: Point Source, ~~Single Image~~ Multiple Epochs

Algorithm:

- Correlate image with PSF to get $\Psi(x, y)$
- Create $\Phi(x, y)$ as an image (or pretend it's constant)
- Find regions where $\nu(x, y)$ is greater than the target threshold (“footprints”)
- Within each footprint, find peaks

Detection: Point Source, ~~Single Image~~ Multiple Epochs

Algorithm:

- Correlate image with PSF to get $\Psi(x, y)$
- Create $\Phi(x, y)$ as an image (or pretend it's constant)
- Find regions where $\nu(x, y)$ is greater than the target threshold (“footprints”)
- Within each footprint, find peaks

Detection: Point Source, ~~Single Image~~ Multiple Epochs

Algorithm:

- Correlate image with PSF to get $\Psi(x, y)$
- Create $\Phi(x, y)$ as an image (or pretend it's constant)
- Warp and Coadd $\Psi(x, y)$
- Warp and Coadd $\Phi(x, y)$
- Find regions where $\nu(x, y)$ is greater than the target threshold (“footprints”)
- Within each footprint, find peaks

Detection: Brainstorming Extensions

- Point sources, multiple epochs, multiple bands, unknown SED
- Galaxies, multiple epochs, single band
- Faint, high proper motion stars, multiple epochs, single band
- Faint supernovae, multiple epochs, multiple bands, known SED

Kinds of Coadds

- Why didn't we just coadd the pixels, instead of that $\psi(x, y)$ thing?
- Ok, why don't we always just coadd $\psi(x, y)$?
- What kinds of coadds might we want to create?

More Information

The official baseline pipeline outputs:

- Data Products Definition Document: <http://ls.st/dpdd>

DM's internal design documents (WIP; more to come) and long-term to-do list:

- <https://confluence.lsstcorp.org/display/DM/Visit+Processing>
- <https://confluence.lsstcorp.org/display/DM/Deep+Processing>

DM's short-term (~6-month) to-do list:

- <https://jira.lsstcorp.org/secure/RapidBoard.jspa?rapidView=58&view=planning&epics=visible>