

# Bias correction for high precision science

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# Outline

The problem

Existing bias-removal methods

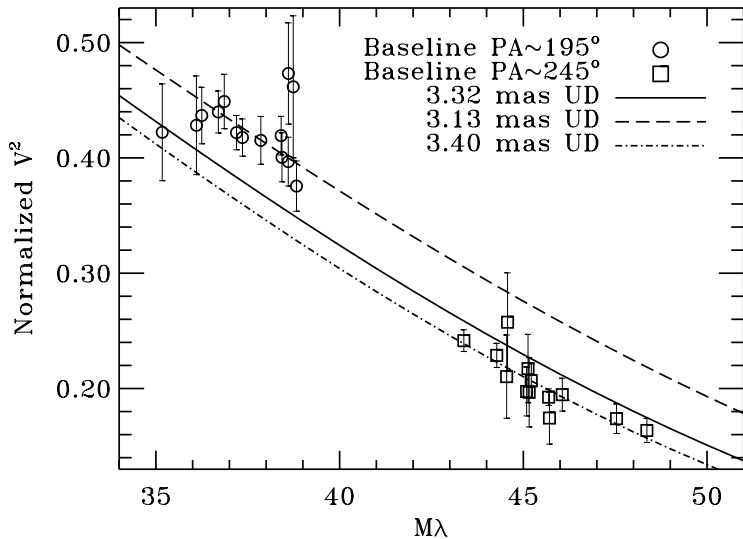
New method

Examples

# Acknowledgements

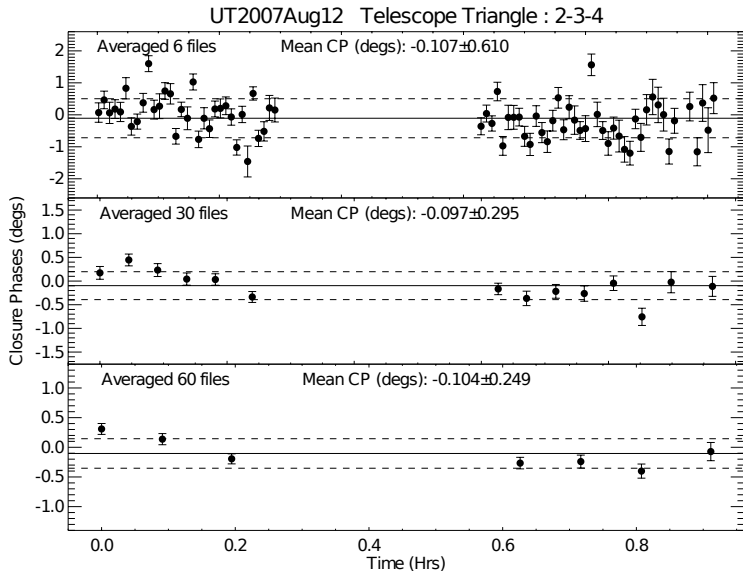
- ▶ Hrobjarthur Thorsteinsson
- ▶ James Gordon

# High-precision measurements yield new science

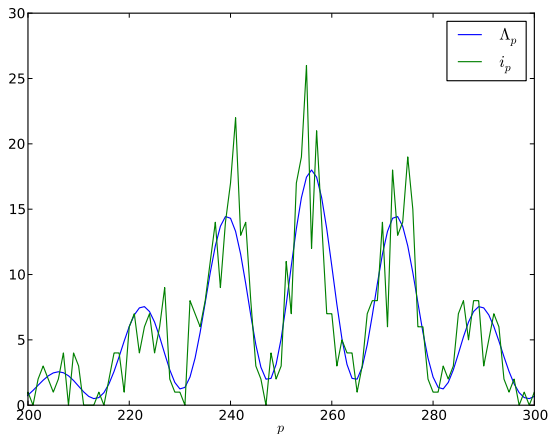


Van Belle *et al.* 2001

# Closure phases are a high-precision observable

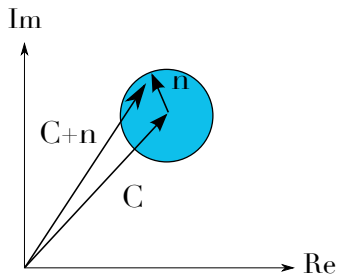


# Detection noise affects all interferometry data



- ▶ Mean photon rate  $\{\Lambda_p, \quad p = 1 \dots N_{\text{pix}}\}$
- ▶ Detected flux  $\{i_p, \quad p = 1 \dots N_{\text{pix}}\}$

# Zero-mean noise can lead to a non-zero bias



$$C = \sum_p \exp(2\pi i u x_p) \Lambda_p \quad c = \sum_p \exp(2\pi i u x_p) i_p = C + n$$

$$\begin{aligned} \langle |C + n|^2 \rangle &= \langle |C|^2 \rangle + 2 \langle \Re [C n^*] \rangle + \langle |n|^2 \rangle \\ &= \langle |C|^2 \rangle + \langle |n|^2 \rangle \end{aligned}$$

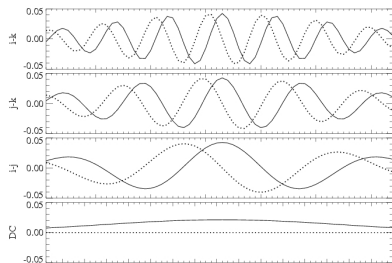
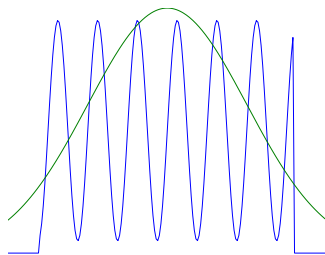
## Photon-noise bias on the bispectrum is well-known

$$\langle C^{ij} C^{jk} C^{ki} \rangle = \langle c^{ij} c^{jk} c^{ki} \rangle - \langle |c^{ij}|^2 \rangle - \langle |c^{jk}|^2 \rangle - \langle |c^{ki}|^2 \rangle + 2 \langle N_{\text{phot}} \rangle$$



# Bias correction is only possible under restricted conditions

- ▶ “DFT conditions”
  - ▶ Integer number of fringe cycles in a scan
  - ▶ No “tapering” allowed
- ▶ No read noise allowed



# We generalise the estimator for the complex amplitude

DFT estimator

$$c^{ij} = \sum_p \exp(2\pi i u^{ij} x_p) i_p$$

Generalised to

$$c^{ij} = \sum_p H_p^{ij} i_p$$

For example, a tapered DFT:

$$H_p^{ij} = W(x_p) \exp(2\pi i u^{ij} x_p)$$

Can represent *any* linear estimator, e.g. P2VM, ABCD

We allow any combination of read noise and photon noise

$$P_{\text{Poisson}}(N_p | \Lambda_p) = \sum_n^{\infty} \delta(N_p - n) \frac{\Lambda_p^{N_p}}{N_p!} \exp[-\Lambda_p].$$

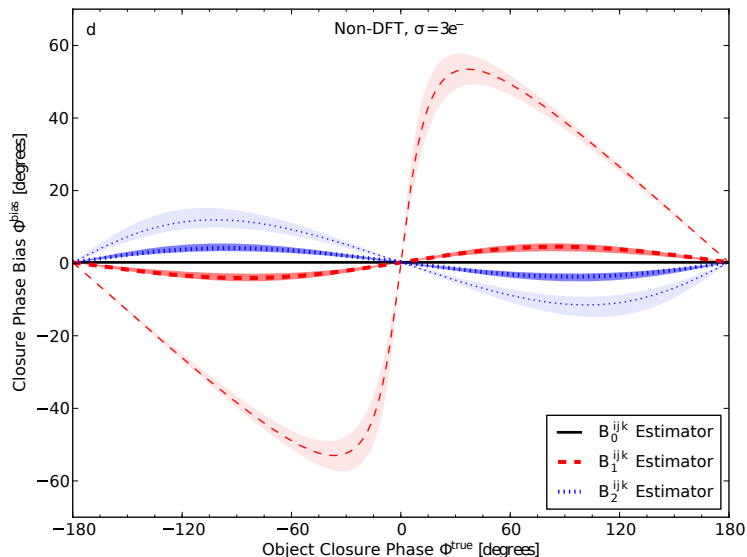
$$P_{\text{Gaussian}}(\epsilon_p | \sigma_p, N_p) = \frac{1}{\sqrt{2\pi\sigma_p^2}} \exp\left[-\frac{(\epsilon_p)^2}{2\sigma_p^2}\right].$$

$$P(i_p | \Lambda_p, \sigma_p) = \int_0^{i_p} P_{\text{Poisson}}(N_p | \Lambda_p) P_{\text{Gaussian}}(i_p - N_p | \sigma_p, N_p) dN_p.$$

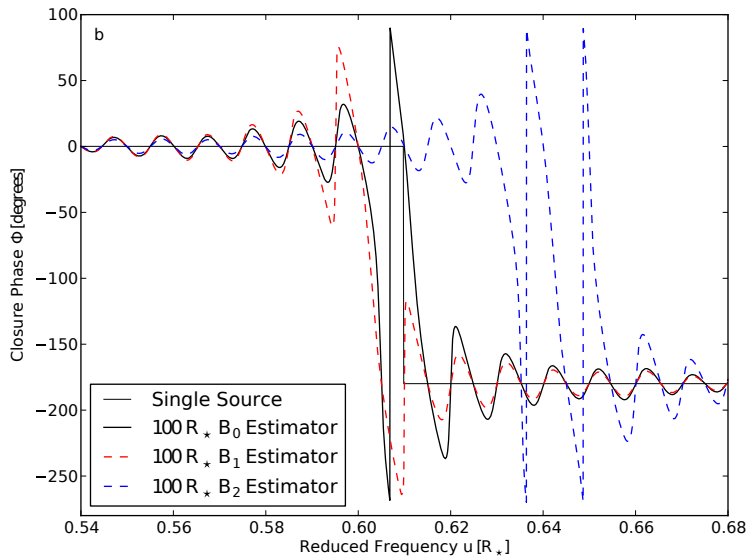
# We derive a bias-free bispectrum estimator

$$\begin{aligned} B_0^{ijk} &= c^{ij} c^{jk} c^{ki} \\ &- c^{ij} \sum_p (i_p + \sigma_p^2) H_p^{jk} H_p^{ki} \\ &- c^{jk} \sum_p (i_p + \sigma_p^2) H_p^{ij} H_p^{ki} \\ &- c^{ki} \sum_p (i_p + \sigma_p^2) H_p^{ij} H_p^{jk} \\ &+ \sum_p (2i_p + 3\sigma_p^2) H_p^{ij} H_p^{jk} H_p^{ki}. \end{aligned}$$

# The closure phase bias typically increases for non-zero closure phase



# A practical example is closure-phase nulling



# Summary

- ▶ Systematic errors can arise from detection noise
- ▶ We have presented a general bias-free bispectrum estimator
- ▶ A small step towards higher-precision science

# Spare slides



# The power spectrum variance contains coupled terms

