



An improved bias correction scheme for SSMIS

Anna Booton and Bill Bell



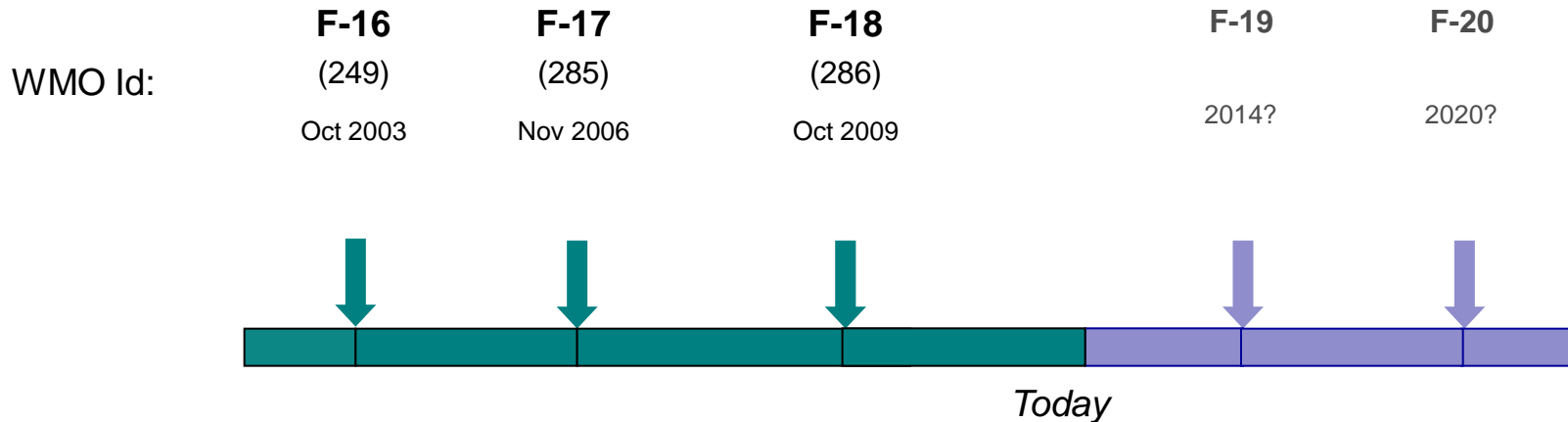
Outline

- Introduction to SSMIS
- A new orbital bias correction scheme
- Implementation within the Met Office's variational bias correction framework



The Special Sensor Microwave Imager Sounder (SSMIS) Instruments

The US Defense* have three satellites flying SSMIS instruments.



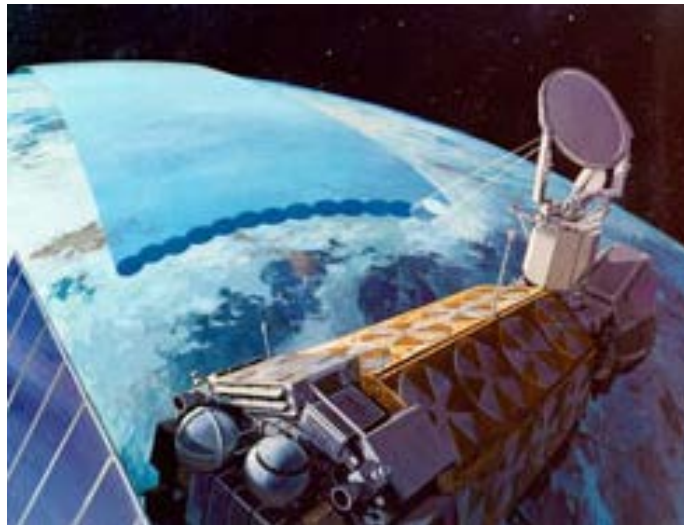
They provide both..

- temperature (50 – 63 GHz)
 - humidity (183GHz)
 - imager data (90 -150 GHz)
- } soundings

measuring radiances from the surface, into the upper atmosphere (mesosphere)

The Special Sensor Microwave Imager Sounder (SSMIS) Instruments' Design

- The instruments are conical scanning radiometers
- Operate by reflecting upwelling radiation off a large rotating mirror into a group of apertures



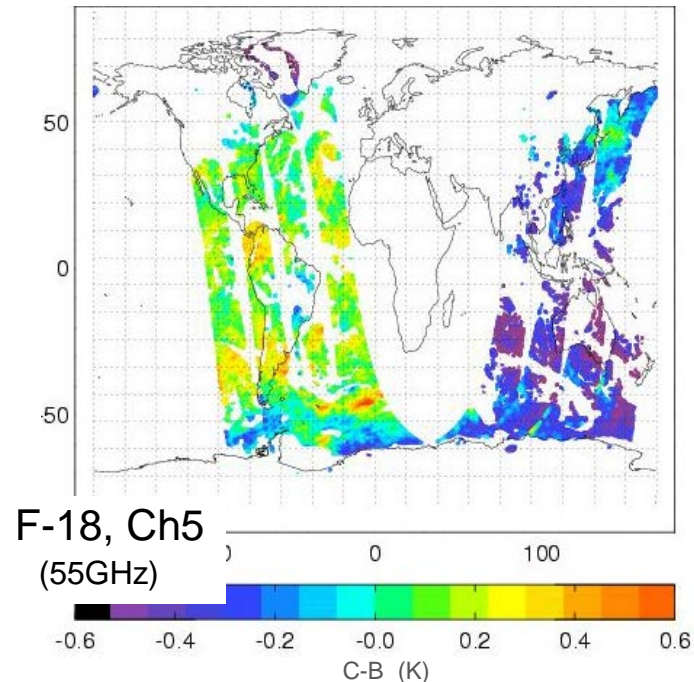
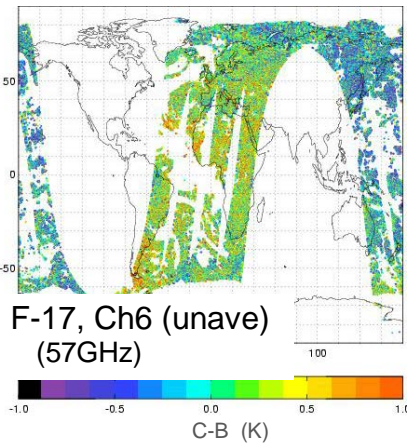
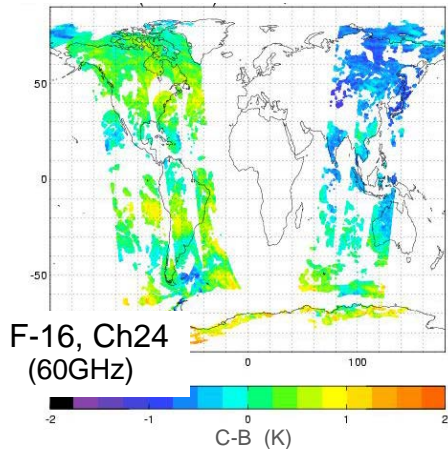
- Unfortunately this design suffers several calibration anomalies, such as:
 - Reflector emissions - *the mirror isn't perfect*
 - Solar intrusions - *into the warm calibration target*

Calibration anomalies

- The calibration anomalies manifest as complex, systematic biases
- Elucidated through comparisons of observed brightness temperatures (O) with those from NWP model backgrounds (B)

For example...

all three instruments exhibit “ascending/descending” biases

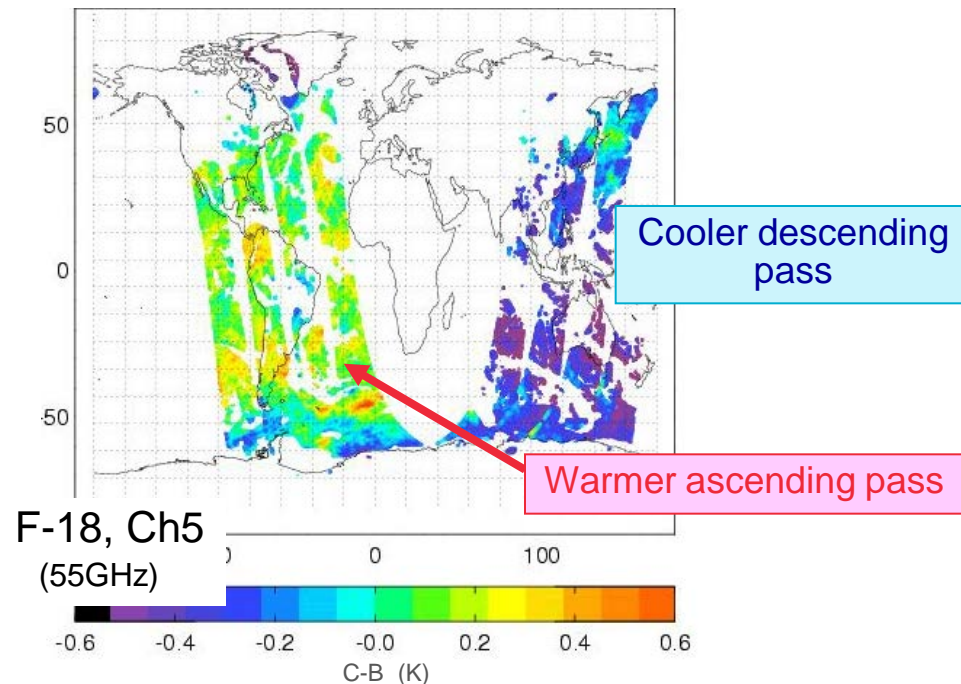
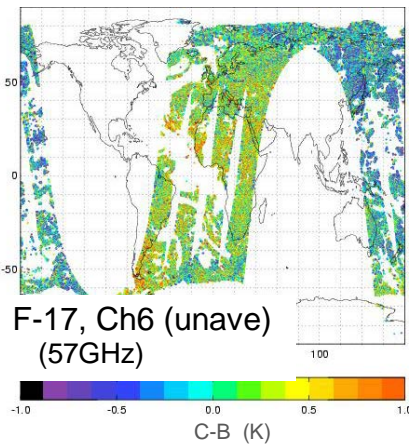
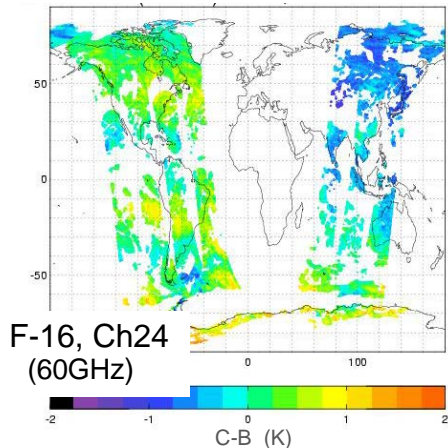


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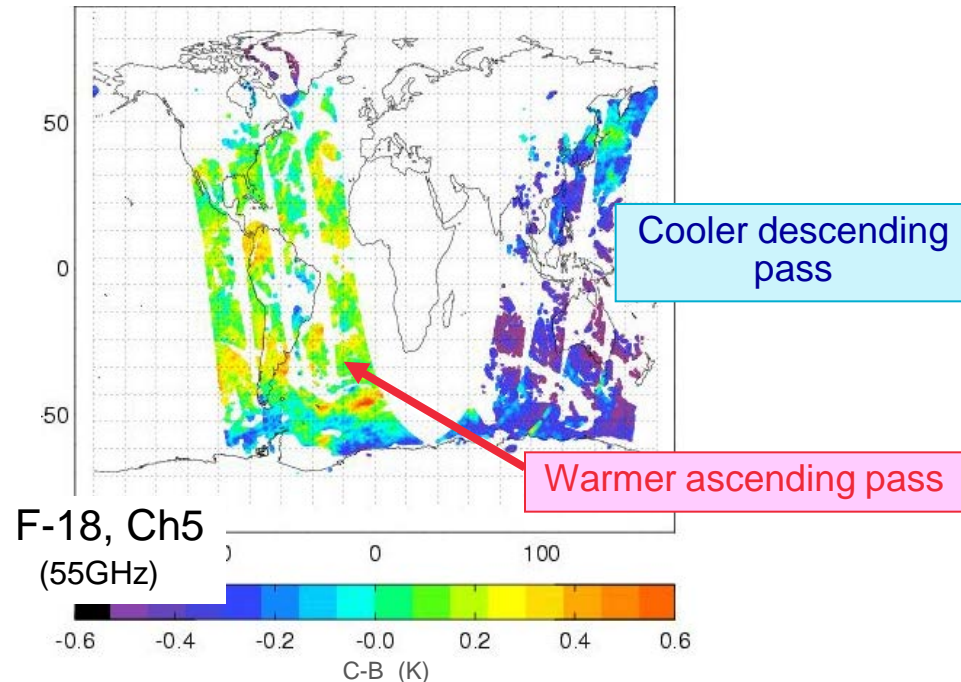
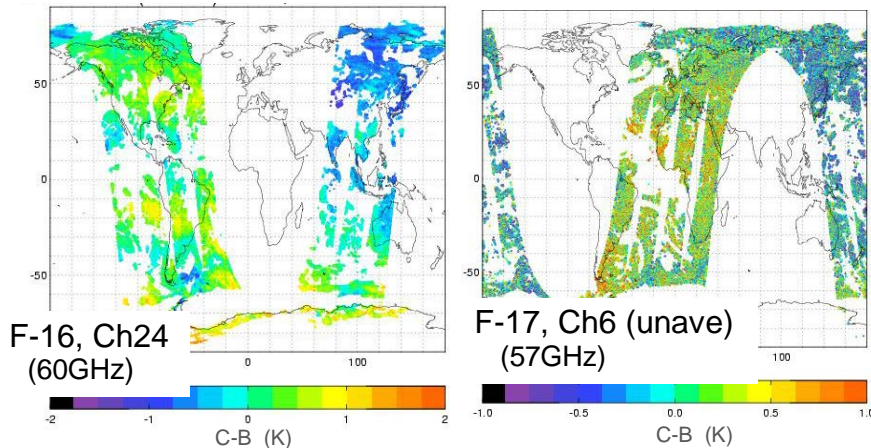


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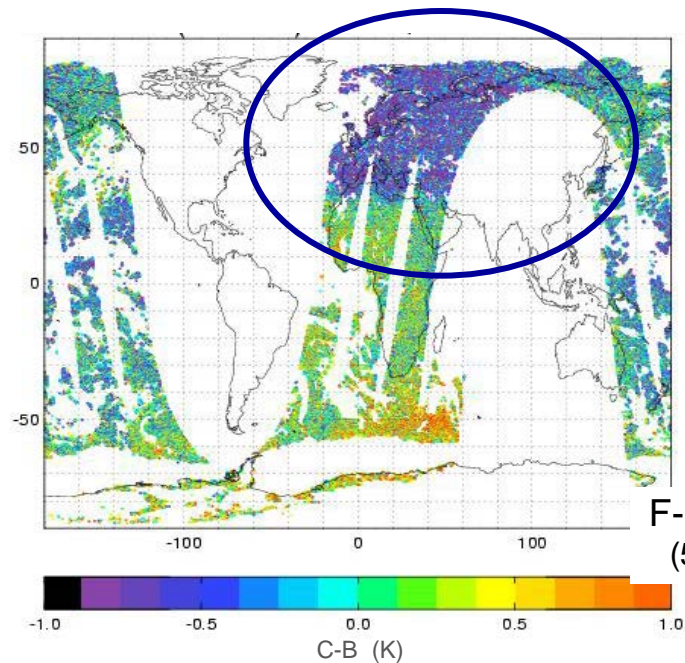
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..this data has been corrected (C) through pre-processing and using the Harris & Kelly scheme

Calibration anomalies

Also get different types of features..



Reflector emissions manifests as sudden, cold region

F-16, Ch6 (unave)
(57GHz)

These sharp transitions are particularly difficult to correct



Corrections have had some success...

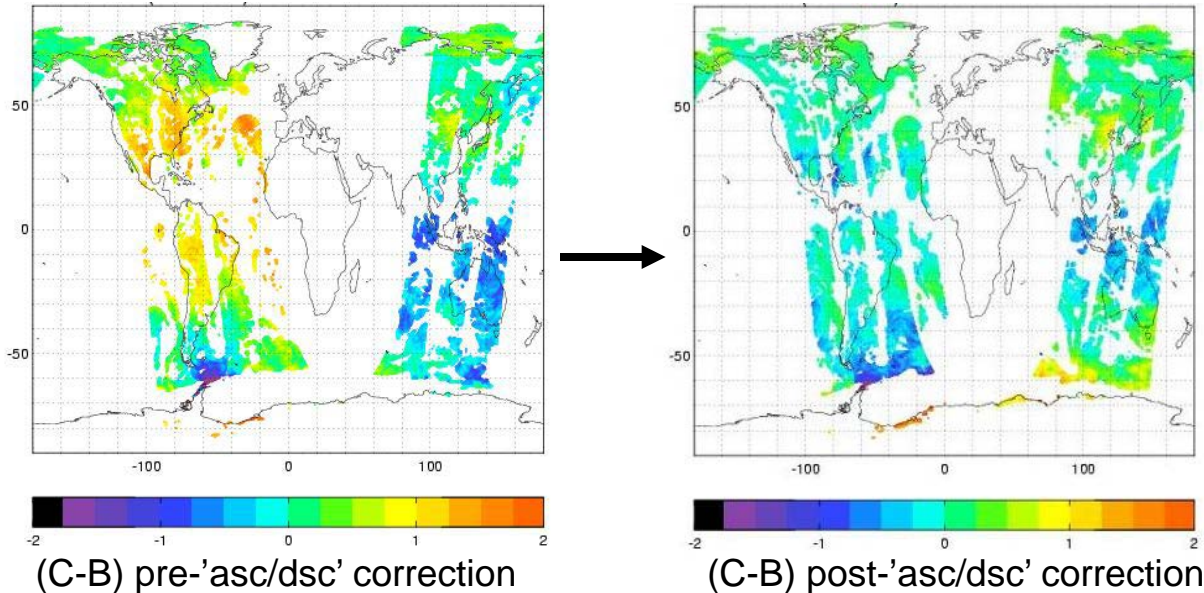
An ascending/descending bias predictor

A co-sinusoidal 'ascending/descending' bias predictor was developed:

$$\Delta T_B = \beta \cos(L + \theta)$$

that effectively reduced the departures.

β – bias coefficient
 L – latitude
 θ – phase shift
from equator





The correction was only partially successful...

1. Within two months the bias correction being applied was no longer applicable
 - The biases were found to be *seasonally* dependent

Ideally require the correction to be applied in a variational bias correction scheme in order for it to evolve with time

2. Worked well as a first order correction
 - Couldn't compensate for biases such as those resulting from reflector emissions

Extend the ascending/descending scheme in order to fit the complex structure



Met Office



An Orbital Bias Predictor



A Fourier Approach

An orbital predictor needs..

- to be a periodic function to represent the cyclic nature of the orbital bias

so a Fourier series expansion would be appropriate:

$$\begin{aligned}\Delta T_B &= \sum \beta P \\ &= \sum_{i=0}^N a_i \cos(i\phi) + b_i \sin(i\phi)\end{aligned}$$

β – bias coefficient
 P – bias predictor
 ϕ – orbital angle

2N coefficients (β)
 $a_0..a_N, b_0..b_N,$

2N predictors (P)
 $\cos(0\Phi).. \cos(N\Phi), \sin(0\Phi).. \sin(N\Phi),$

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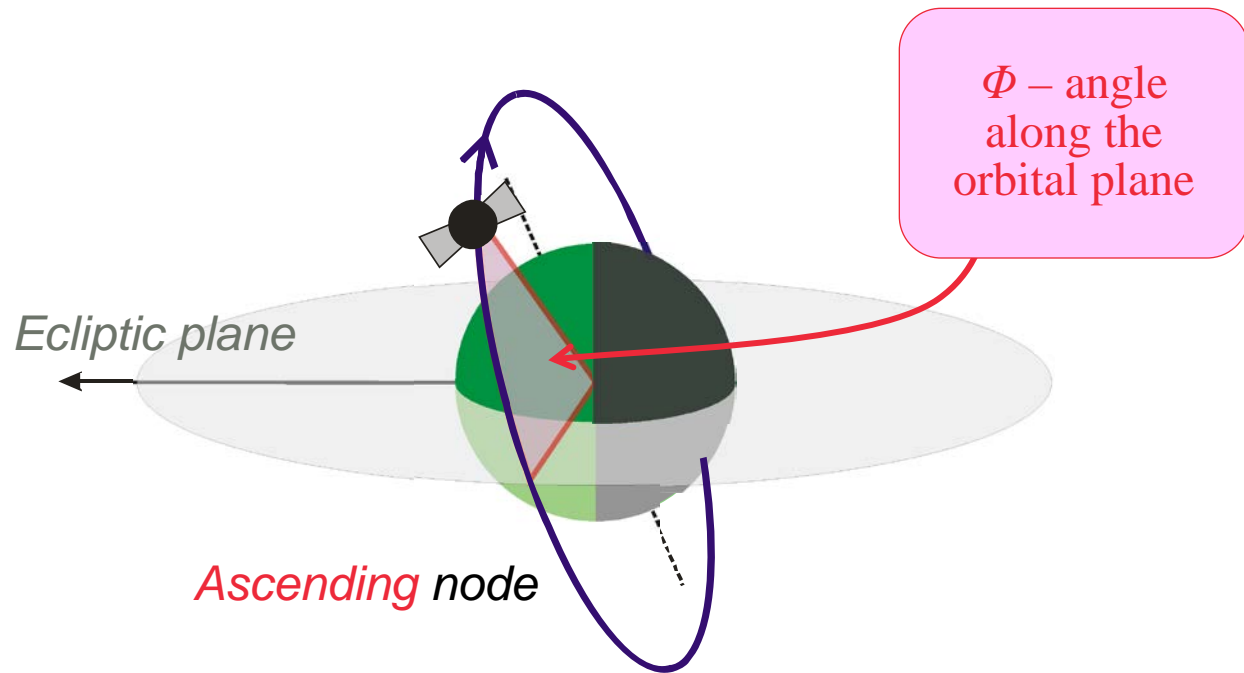
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2N predictors (P)
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- is an extension from the ‘asc/dsc’ predictor
- should capture the complex features that the original did not
- Propose the new predictor is a function of the orbital angle rather than latitude ...

The orbital angle, Φ

- Referenced from the intersection of the satellite's ascending node with the ecliptic plane



Using reference relative to the position of the sun should result in a more stable predictor



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Implementing the predictor...

within the Met Office's variational bias correction framework



Toy Model

- A toy model representative of the Met Office's variational bias correction (VarBC) data assimilation system was developed to investigate the performance of the scheme.
- Inputs to the toy model are:
 - Residual biases (O-B) vs. ϕ
 - Prior estimate of the coefficients a_i, b_i
- At each cycle:
 - The corrected 'C' temperatures are obtained
 - The coefficients a_i, b_i , are updated



Toy Model

The update is computed by minimising the cost function:

$$J = J_o + J_B$$



observations background

Toy Model

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$$\begin{aligned}
 J &= J_o + J_B \\
 &= \underbrace{\frac{\left(\sum^{All\ Obs} ((O-B) - \beta' P) \right)^2}{\sigma_o^2}}_{\text{fit of predictor to observations}} + \underbrace{\frac{(\beta - \beta')^2}{\sigma_B^2}}_{\text{fit of predictor to previous background}}
 \end{aligned}$$

(O-B) – observation
minus model background

$\beta' P$ – ‘prediction’ of the
residual bias

β – bias coefficients from
current cycle

β' – bias coefficients
from previous cycle

σ_o, σ_B – weightings

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 \end{aligned}$$

... have two unknowns, to estimate...

- the number of predictors, N and
- relative weightings of observation and background component, σ_o, σ_B



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What weighting is best?

- The magnitude of the weightings, (σ_o, σ_B) are a function of the number of observations assimilated

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- The magnitude of the weightings, (σ_o, σ_B) are a function of the number of observations assimilated
- The weightings are optimized for each channel accordingly

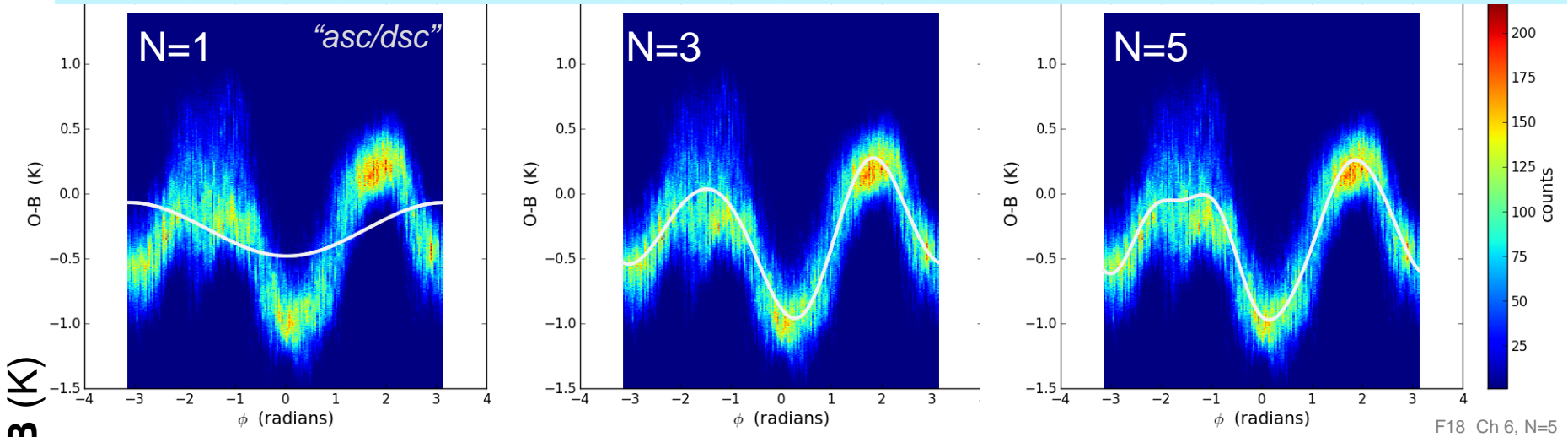


How many coefficients give a good fit?

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Before run toy model...

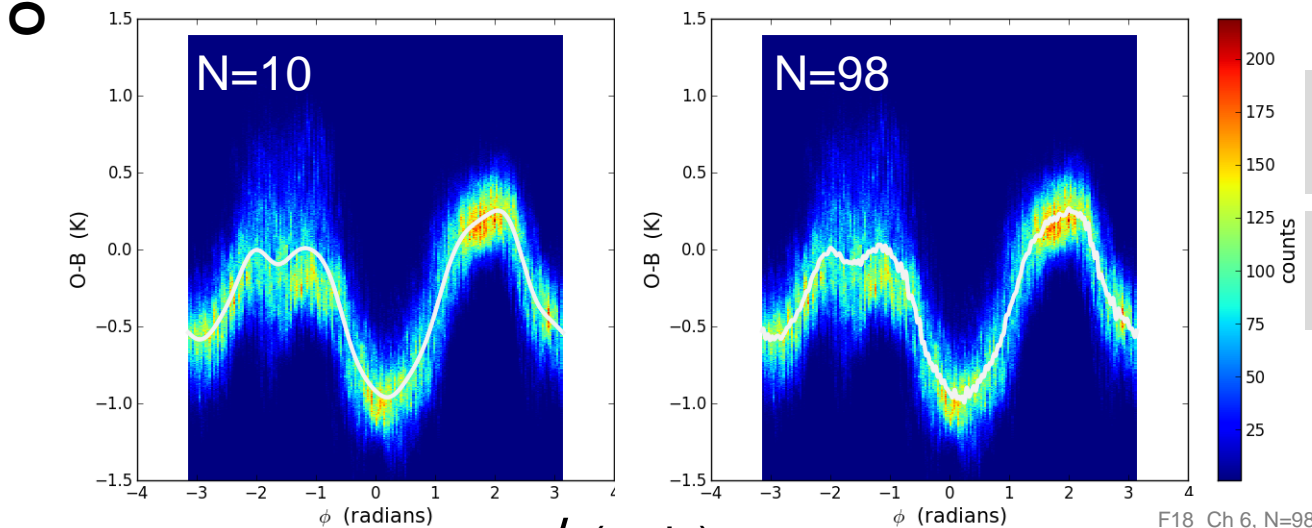
Collected (O-B) data over a 3 day period (13 cycles) and fit a Fourier Series to it..



F18 Ch 6, N=5

Cumulative counts for 201309200000 - 201309231800

Cumulative counts for 201309200000 - 201309231800



F18 Ch 6, N=98

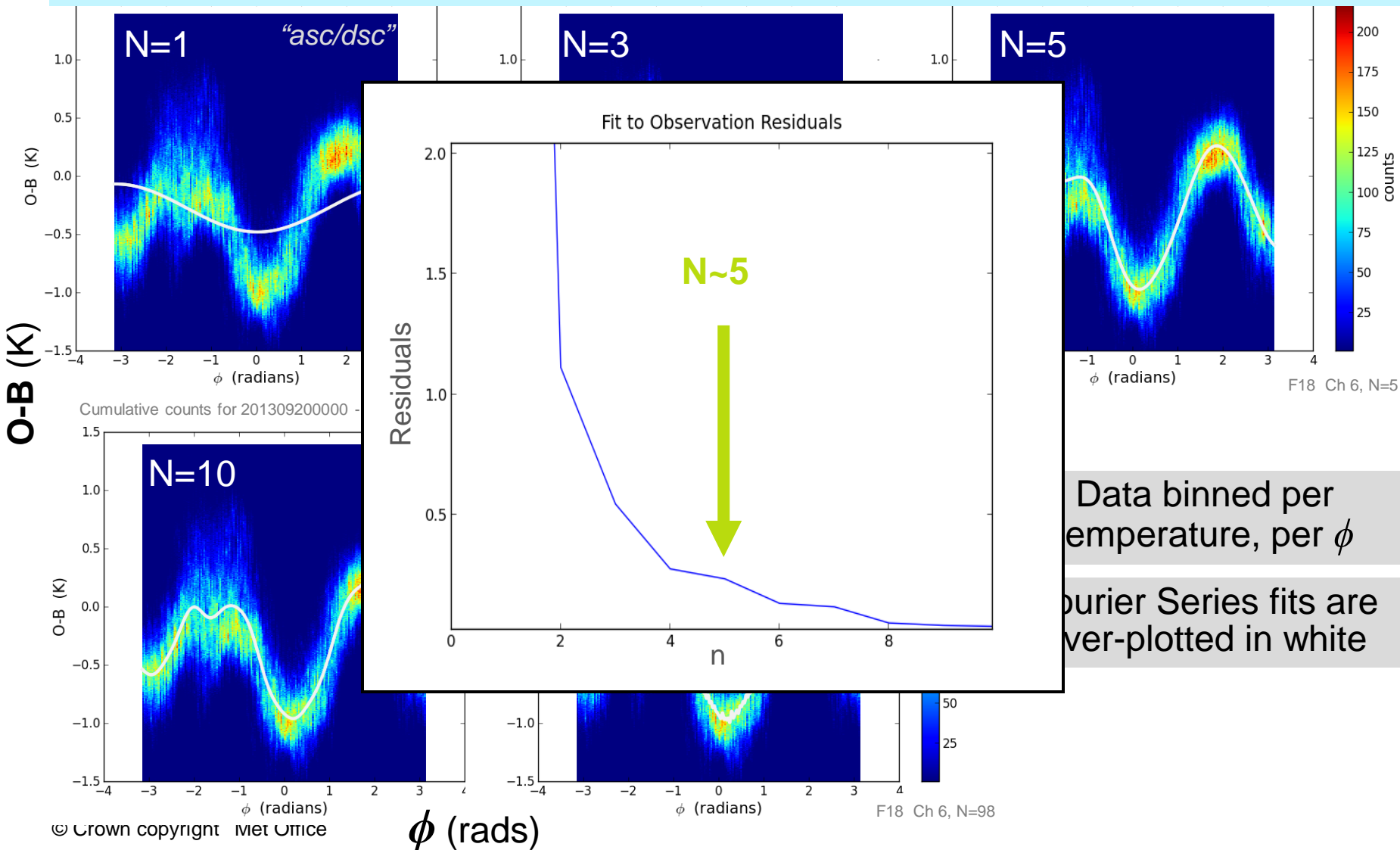
Data binned per temperature, per ϕ

Fourier Series fits are over-plotted in white

How many coefficients give a good fit?

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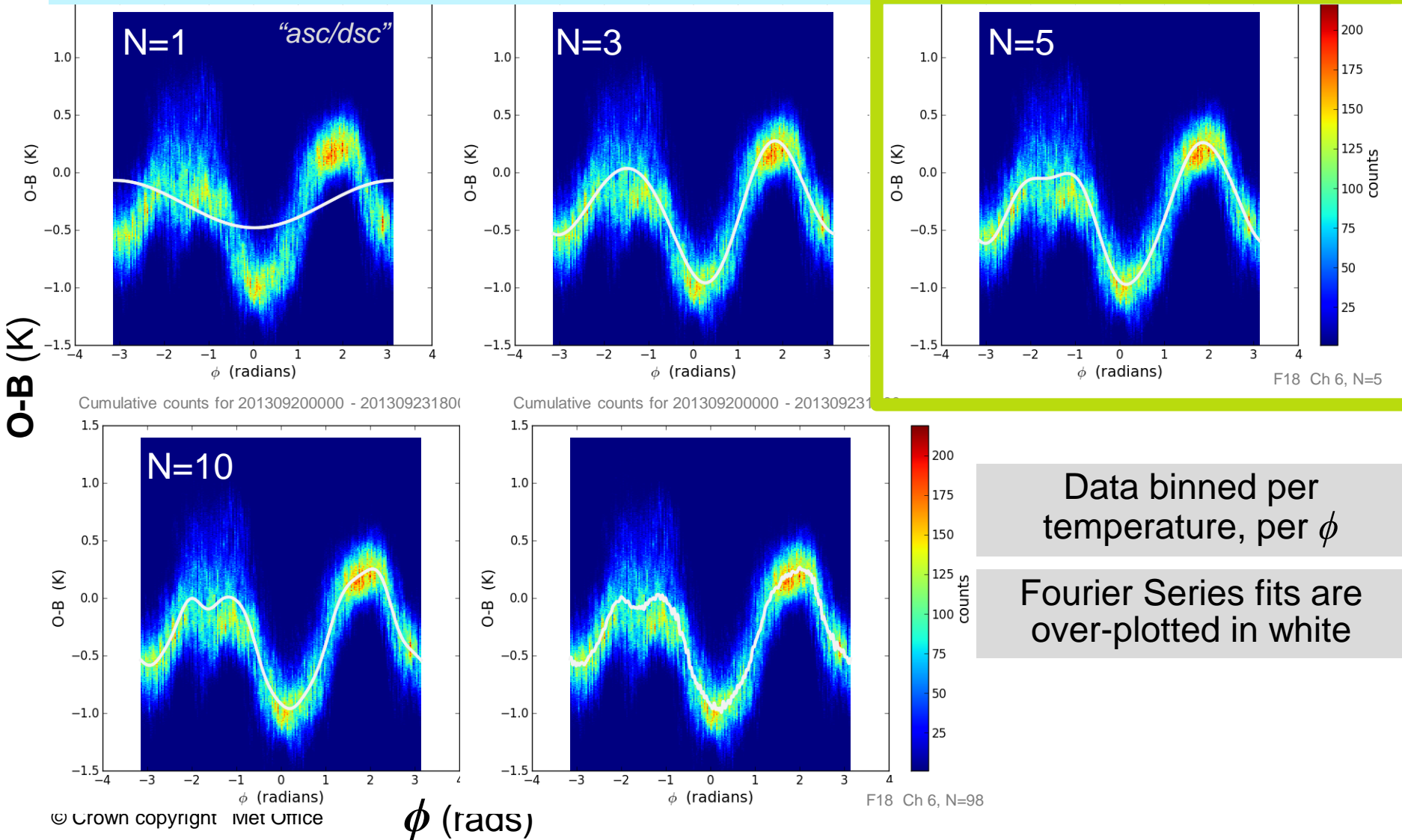
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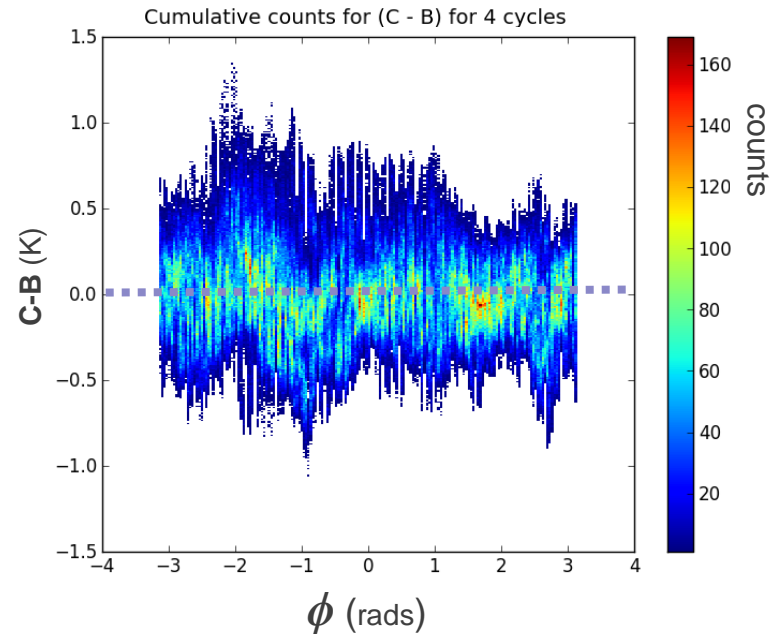
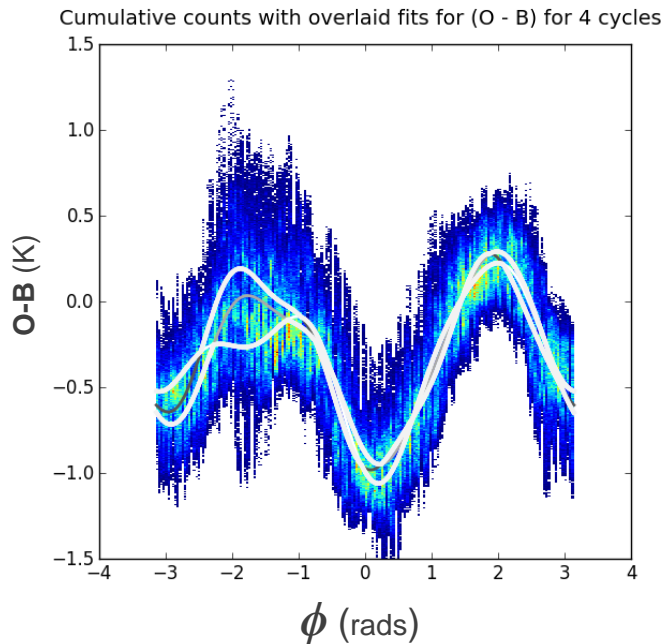
How well does the predictor work within a VarBC framework?

Using the toy model...

- Using a predictor comprised of 10 components ($N=5$), and optimally tuned background weighting

For 24 hours (4 cycles)...

F-18, Ch6
(57GHz)



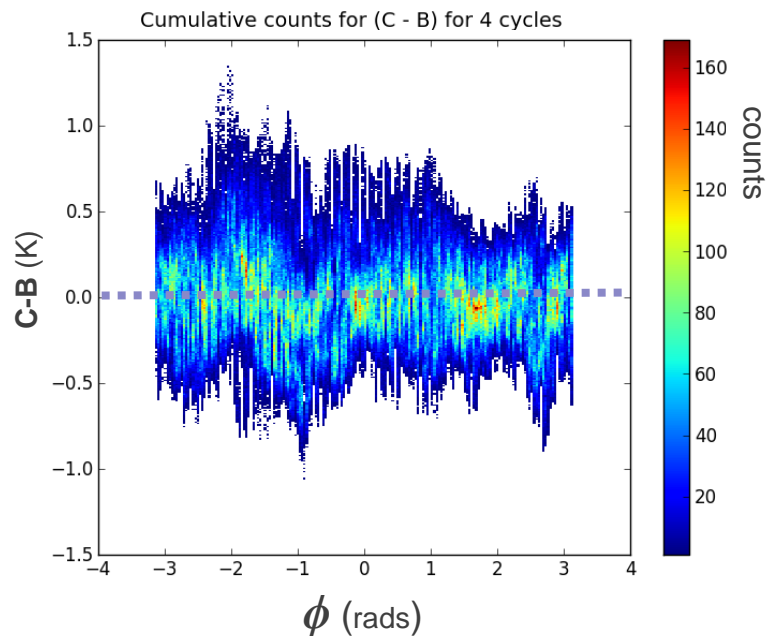
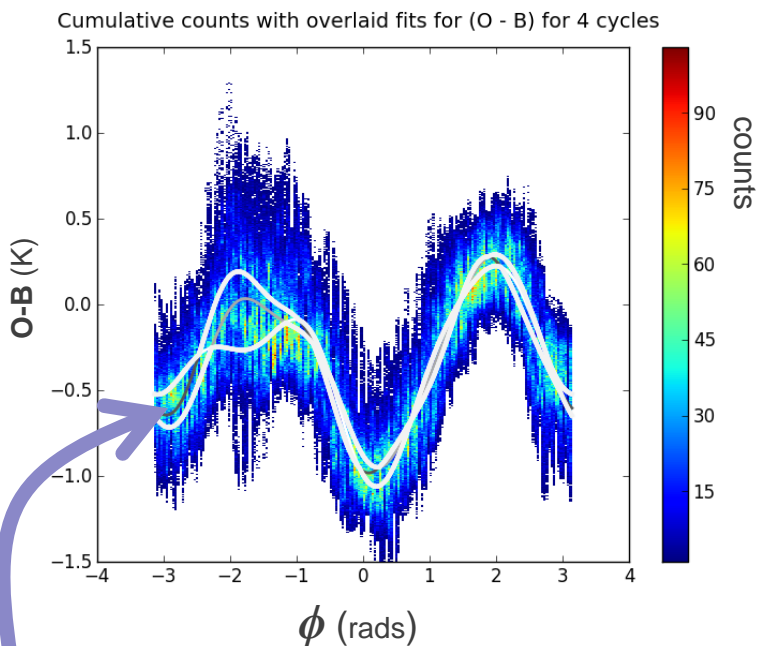
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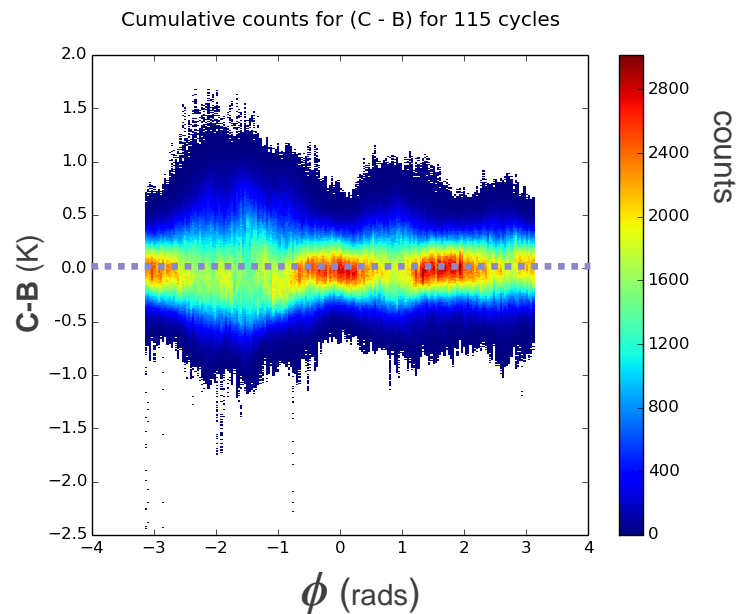
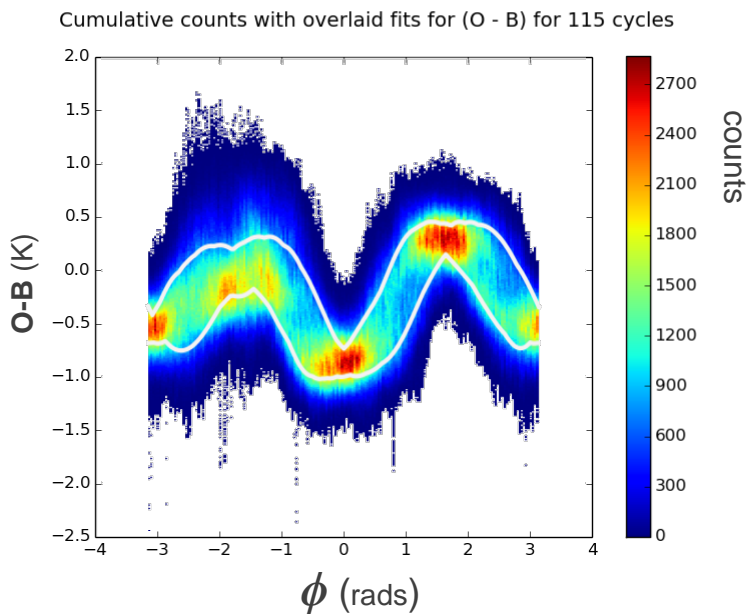
Envelope of the variational fits (white) shows a reasonable fit to the (O-B)'s

Corrected brightness temperatures are reduced to near zero

But was there any evolution in the bias over time?

Over a 30 day period (115 cycles)...
(20th Sept – 20th Oct 2013)

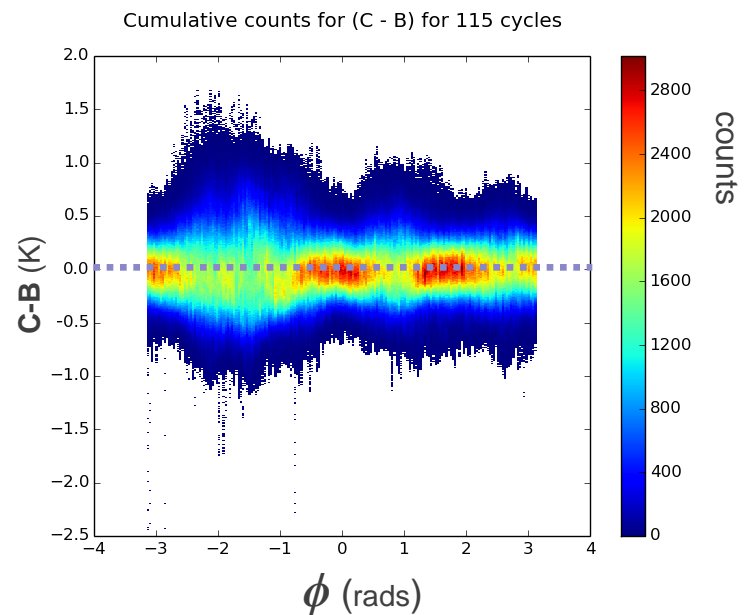
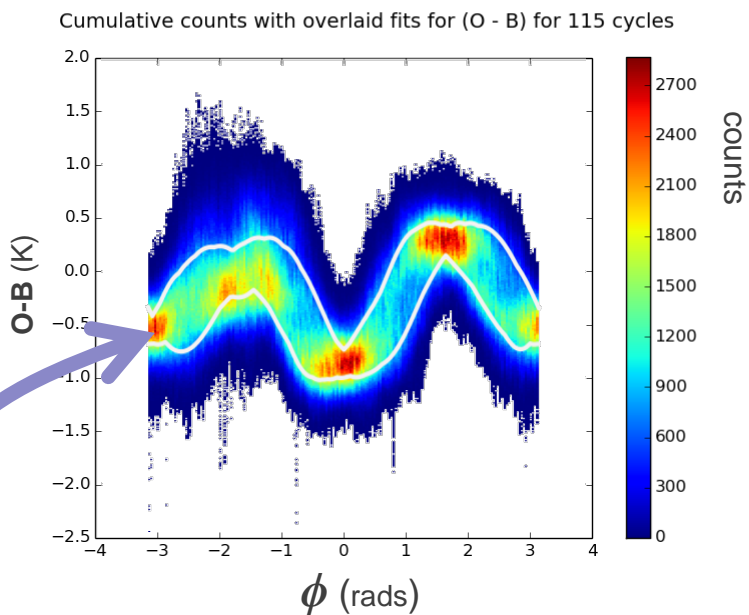
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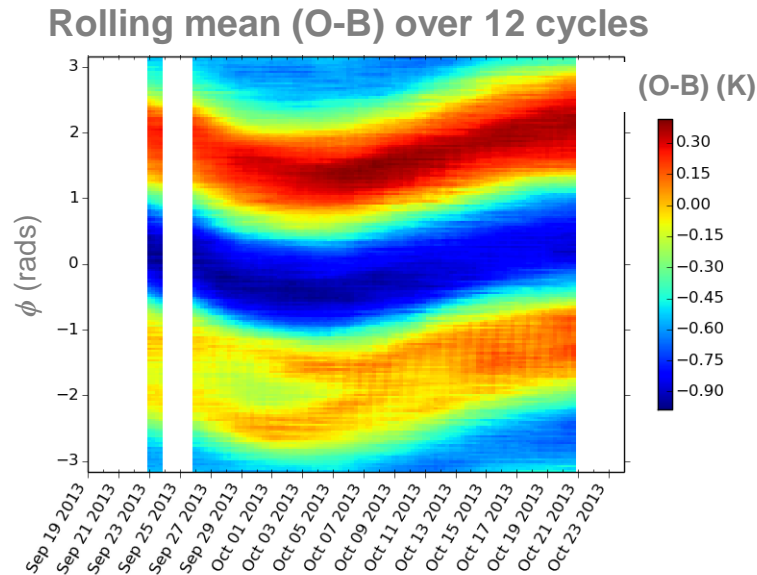
F-18, Ch6
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The general orbital structure is 'predicted' well

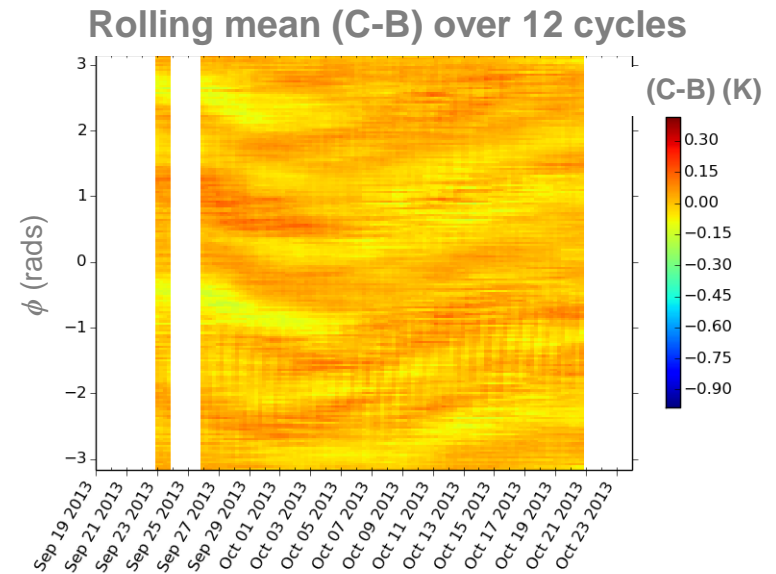
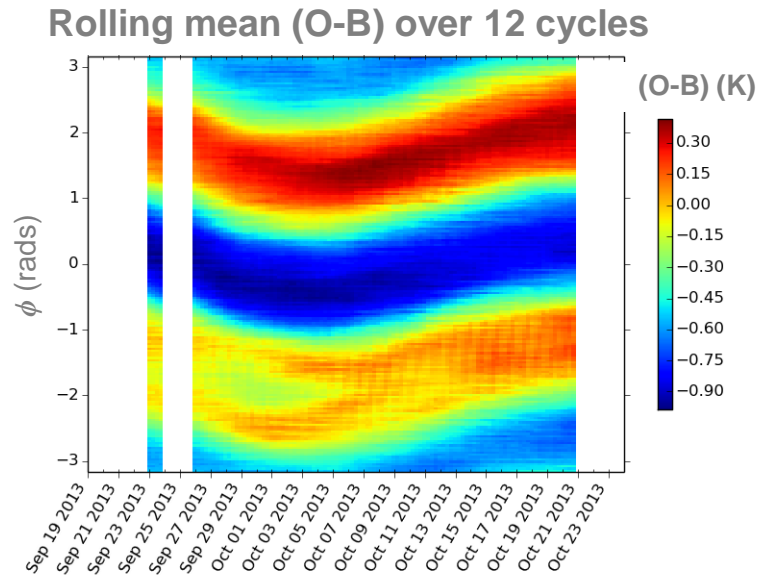
Resulting C-B exhibit little structure

More detailed analysis...



The structure of the orbital biases
has evolved with time

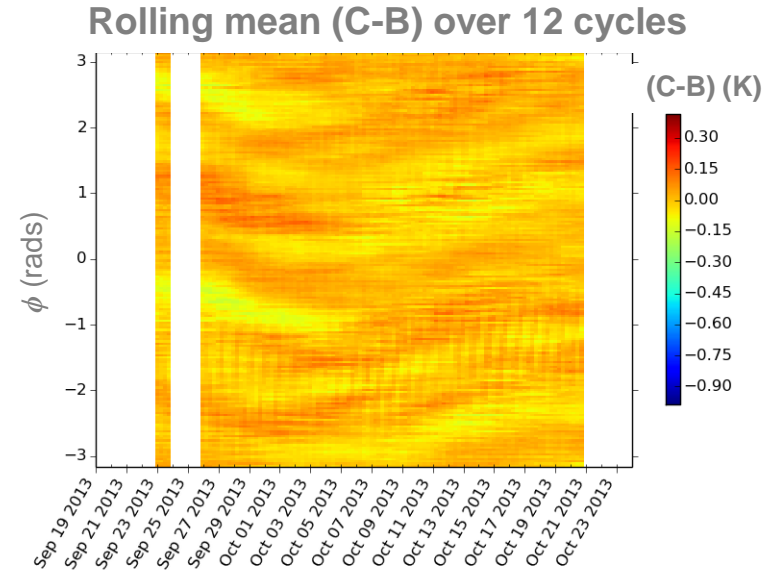
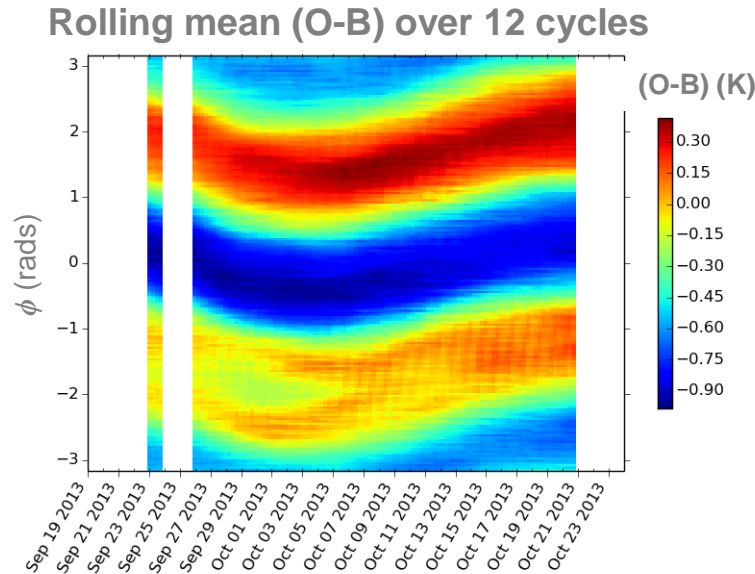
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More detailed analysis

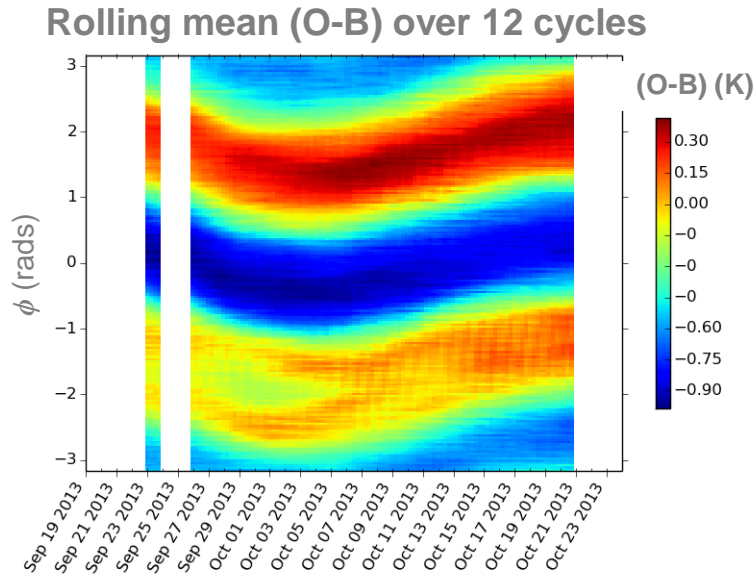
The Fourier Series fit has also evolved with time – residual biases reduced to ~50mK



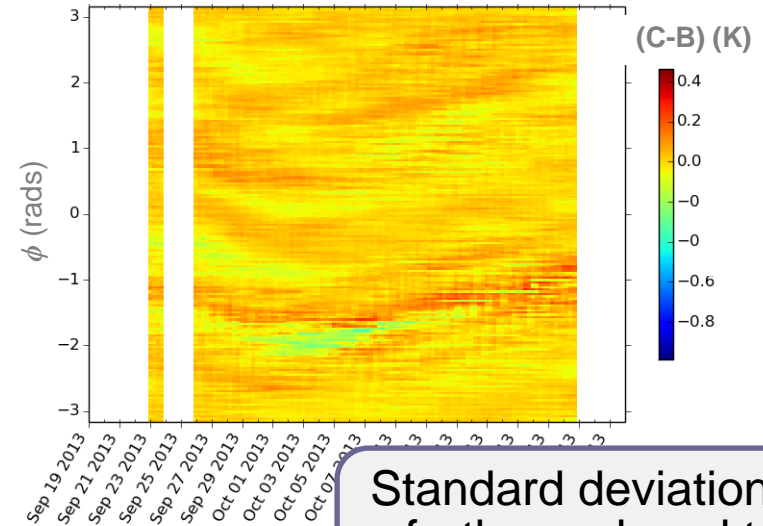
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More detailed analysis

The Fourier Series fit has also evolved with time –
Introduce QC ~50mK

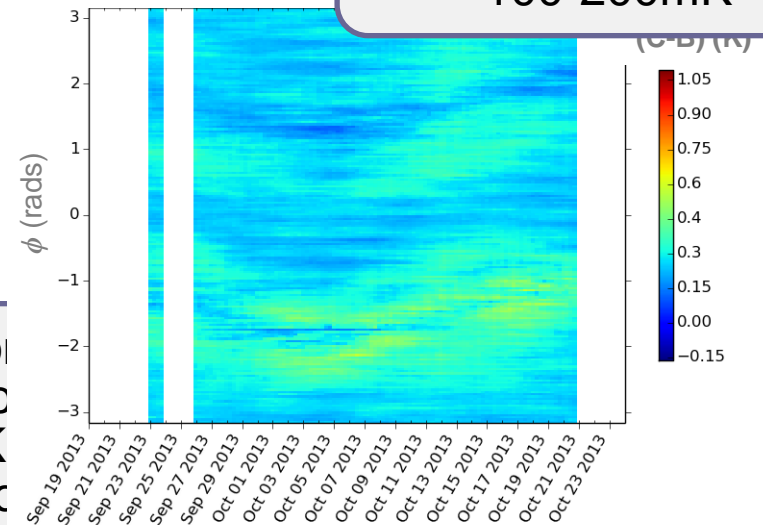


Rolling mean (C-B) over 12 cycles



Standard deviation is further reduced to ~100-200mK

Rolling standard deviation



The structure of the orbital biases has evolved with time

Post-correction standard deviation ~200-300mK over majority of c



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But...

Can it be applied to surface sensitive channels?



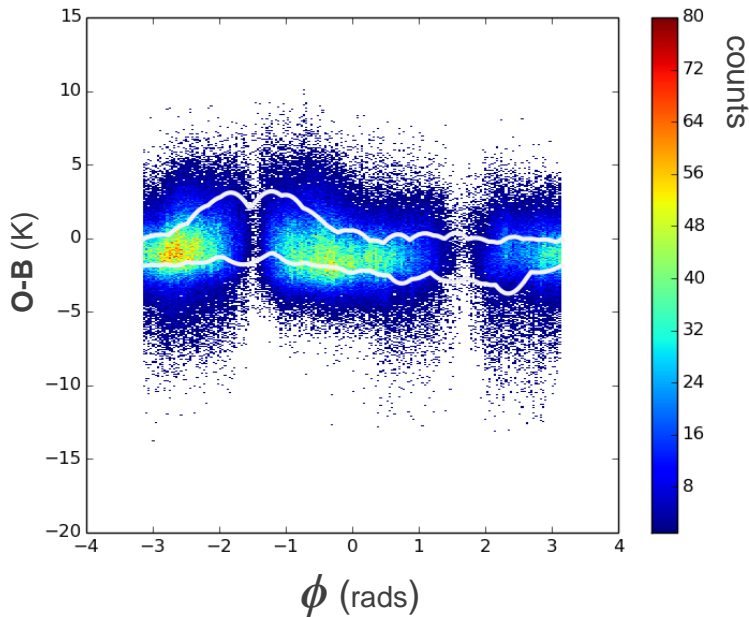
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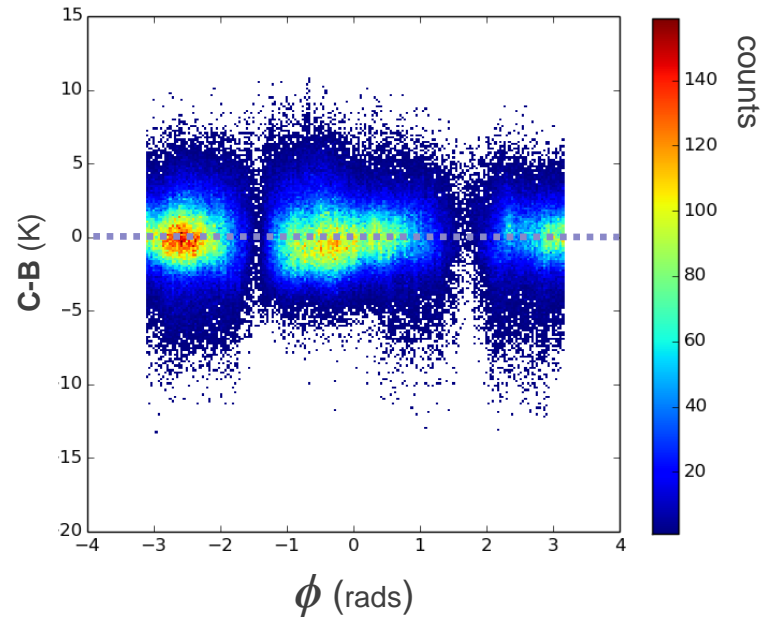
Including QC, and selecting sea-only observations...

F-18, Ch16
(37GHz)

Cumulative counts with overlaid fits for (O - B) for 115 cycles



Cumulative counts for (C - B) for 115 cycles



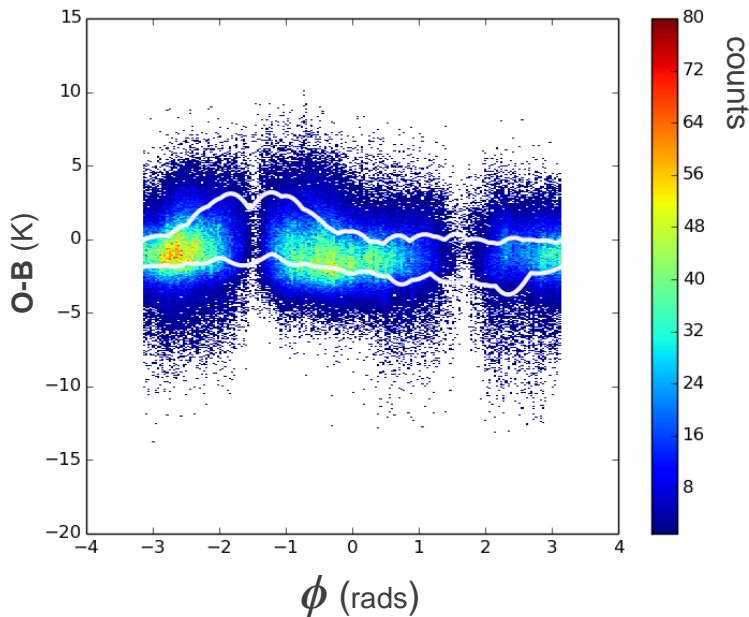
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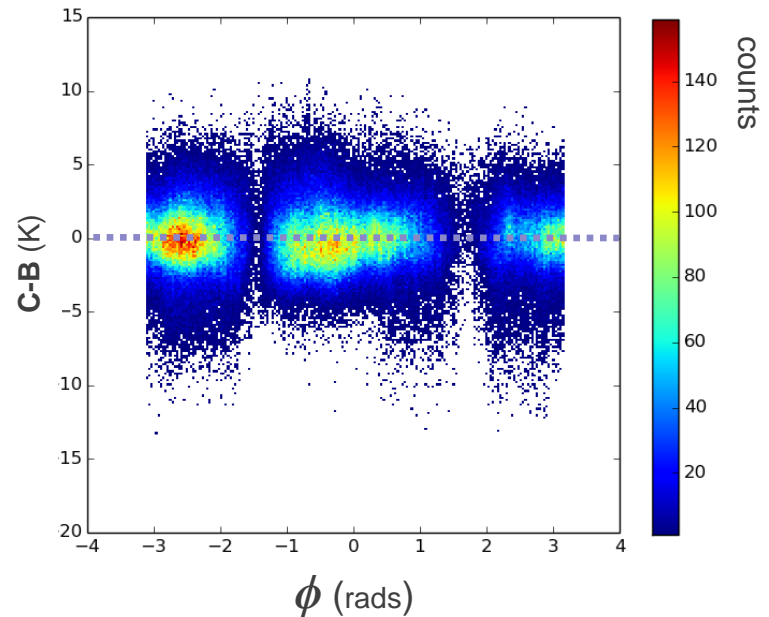
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Cumulative counts for (C - B) for 115 cycles



Corrected brightness temperatures are again reduced to zero – it is providing some benefit

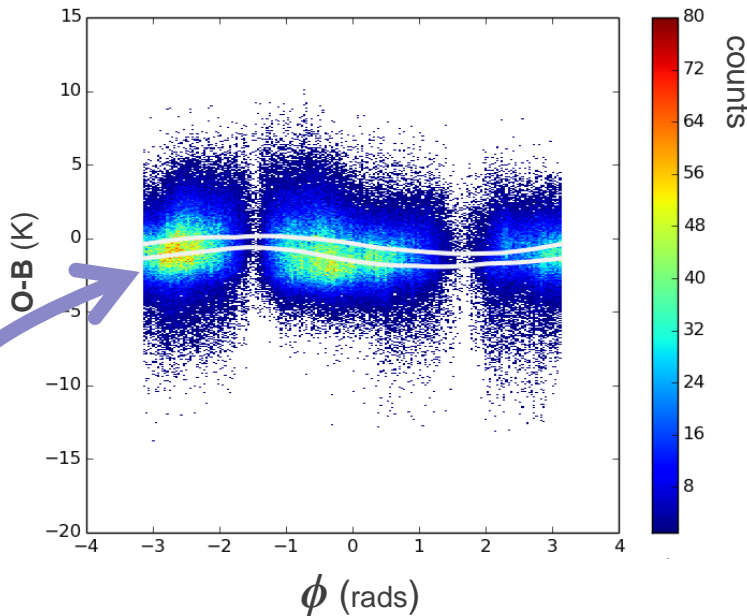
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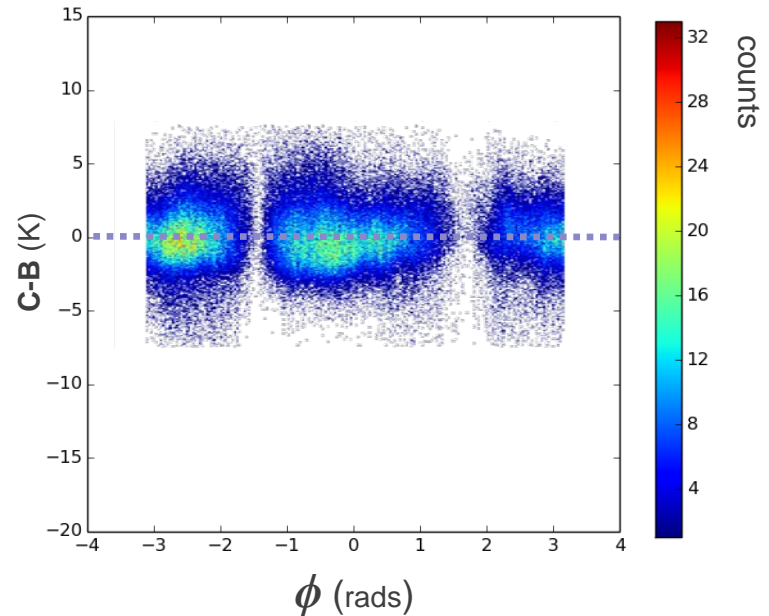
F-18, Ch16
(37GHz)

Cumulative counts with overlaid fits for (O - B) for 115 cycles



Using a correction of two components ($N=1$) fits this structure

Cumulative counts for (C - B) for 115 cycles



Correction still provides slight benefit



Conclusions

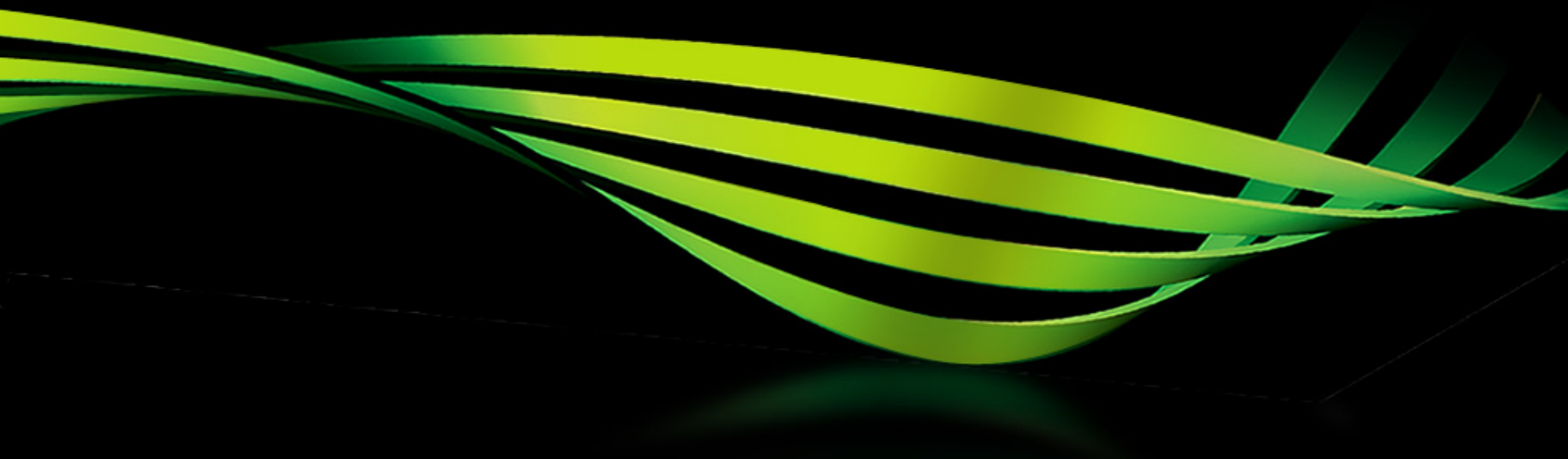


Conclusions

- SSMIS exhibits complex, orbital biases – even after physical corrections
- A new orbital bias correction scheme is under development using a Fourier Series in ϕ
- Investigations with a variational toy model show:
 - Good fits obtained using 10 components, 2 components for surface sensitive channels
 - Tuning of background/observation weightings gives good stability in time
 - Residual biases reduced to ~50mK over a month of orbits
 - for a 30 day test with real F-18, Ch6 data
 - Surface sensitive channels benefit slightly
- Results are consistent for F16, F17 and F18
- Next steps include:
 - Implementing in operational system



Met Office



Thank you for listening!