

Is Your Dark Energy Theory Falsifiable?

Andreas Albrecht (UC Davis)

DarkFun 2007, LBL

Jan 19 2007

- How can you test your DE theory?
- How can you estimate the testability of a given theory when you with proposed new data

1) Direct comparison with data

1) A “quick and dirty” approach

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How well do Dark Energy Task
Force simulated data sets constrain
specific scalar field quintessence
models?

Augusta Abrahamse

Brandon Bozek

Michael Barnard

+AA

+

DETF
Simulated data

+

Quintessence
potentials

+

MCMC

See also Dutta & Sorbo 2006

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Why now?

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Why now?

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→ State impact of future
experiments

→ Input to designing
future expts?

potentials

MCMC

See also Dutta & Sorbo 2006

How well do Dark Energy Task Force simulated data sets constrain specific scalar field quintessence models?

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In progress

quintessence potentials

+

MCMC

See also Dutta & Sorbo 2006

The potentials

Exponential (Wetterich, Peebles & Ratra)

PNGB (Frieman et al)

Exponential with prefactor (AA & Skordis)

The potentials

Exponential (Wetterich, Peebles & Ratra)

$$V(\varphi) = V_0 e^{-\lambda\varphi}$$

PNGB (Frieman et al)

$$V(\varphi) = V_0 (\cos(\varphi / \lambda) + 1)$$

Exponential with prefactor (AA & Skordis)

$$V(\varphi) = V_0 \left(\chi (\varphi - \beta)^2 + \delta \right) e^{-\lambda\varphi}$$

& others

The potentials

Exponential (Wetterich, Peebles & R

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Brandon

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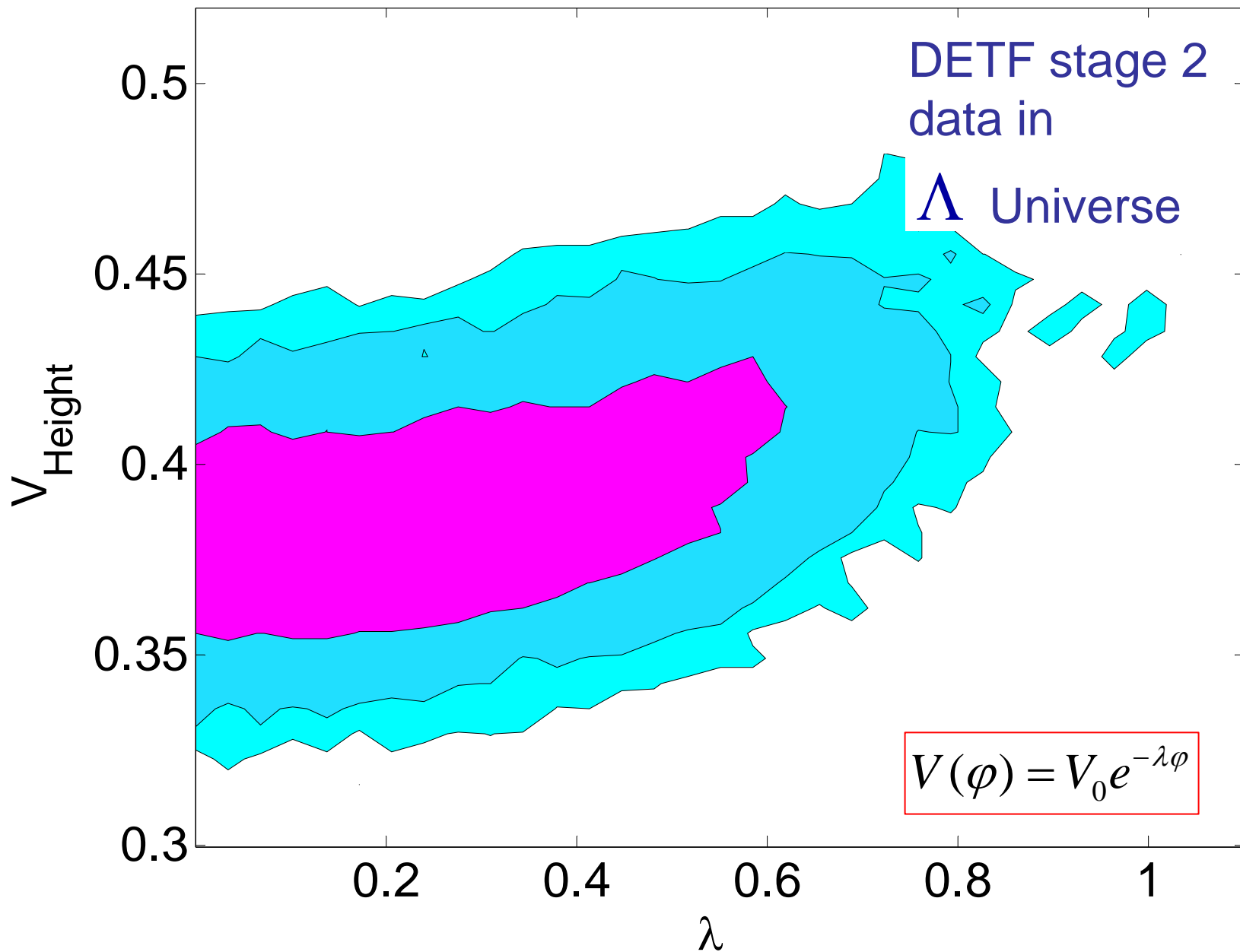
Michael

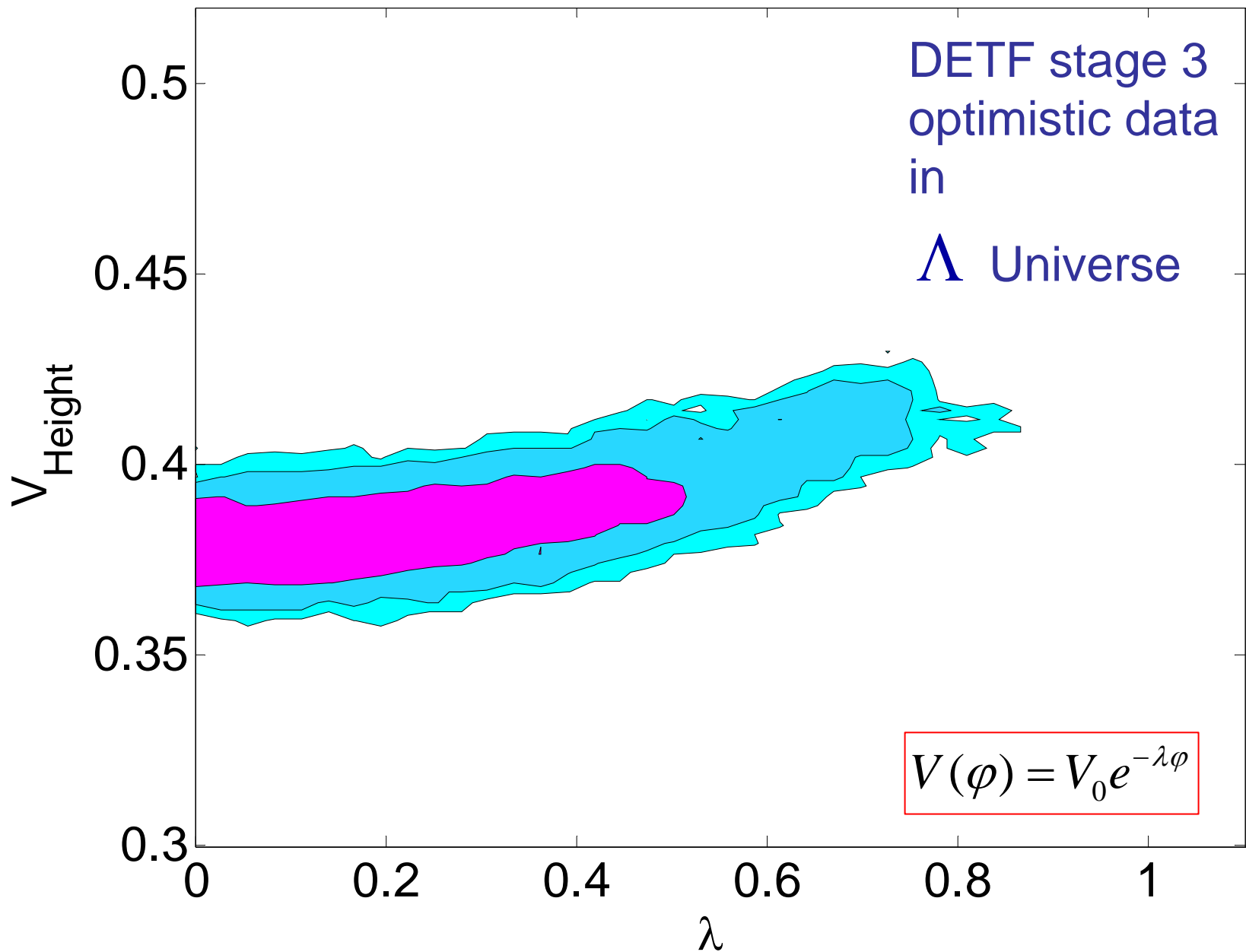
& others

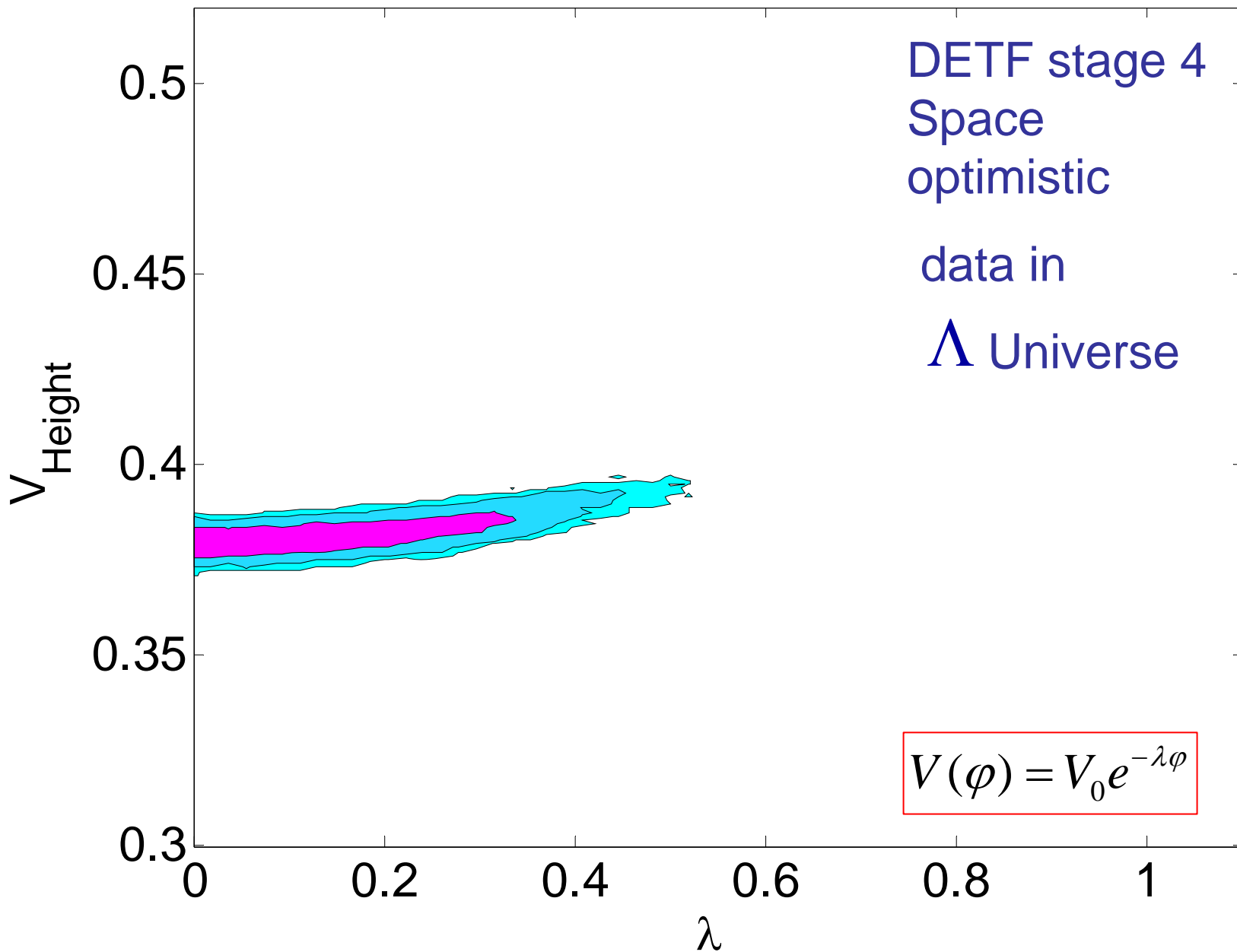
Challenges: Potential parameters can have very complicated relationships to cosmic observables

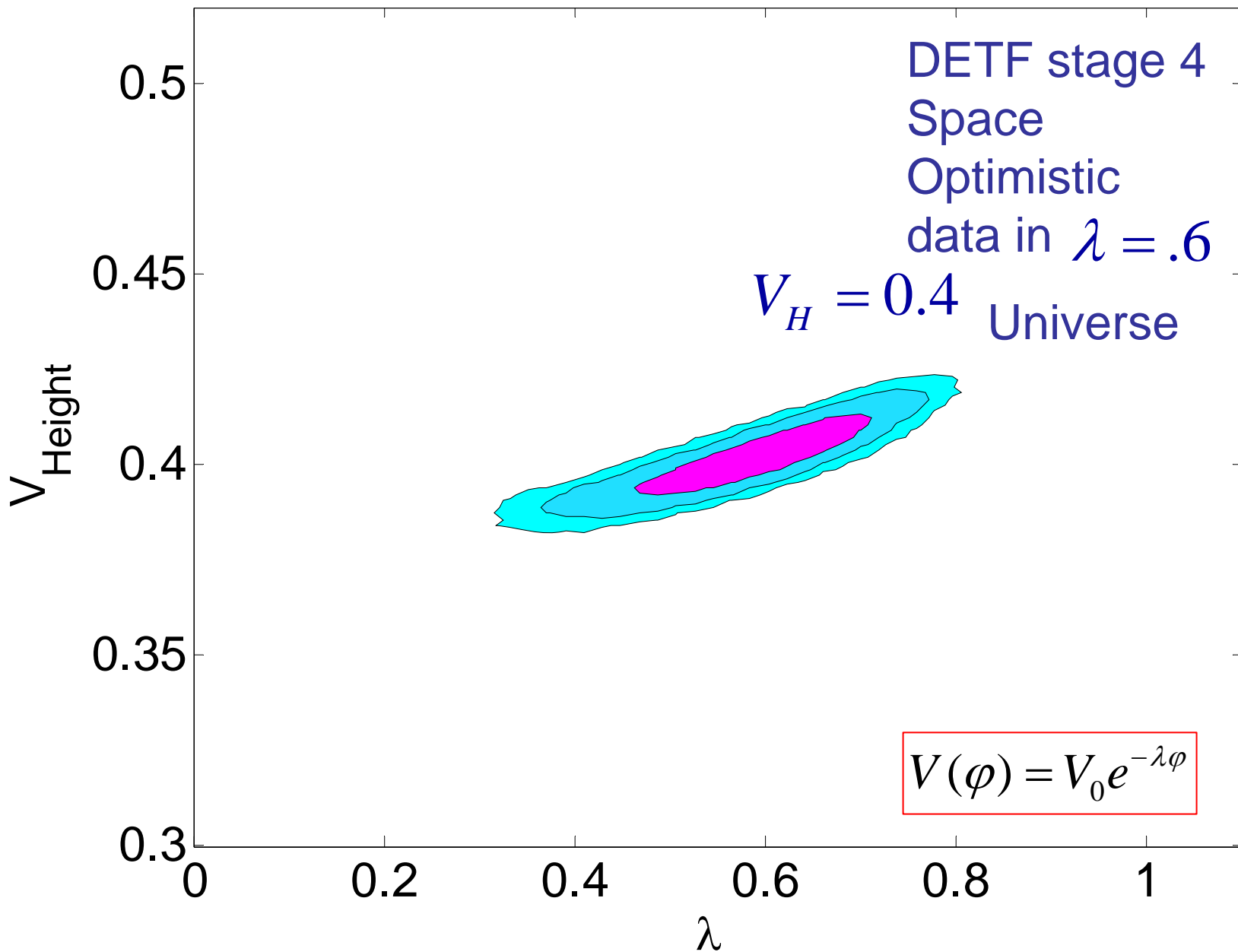
Resolved with good parameter choices (functional form and value range)

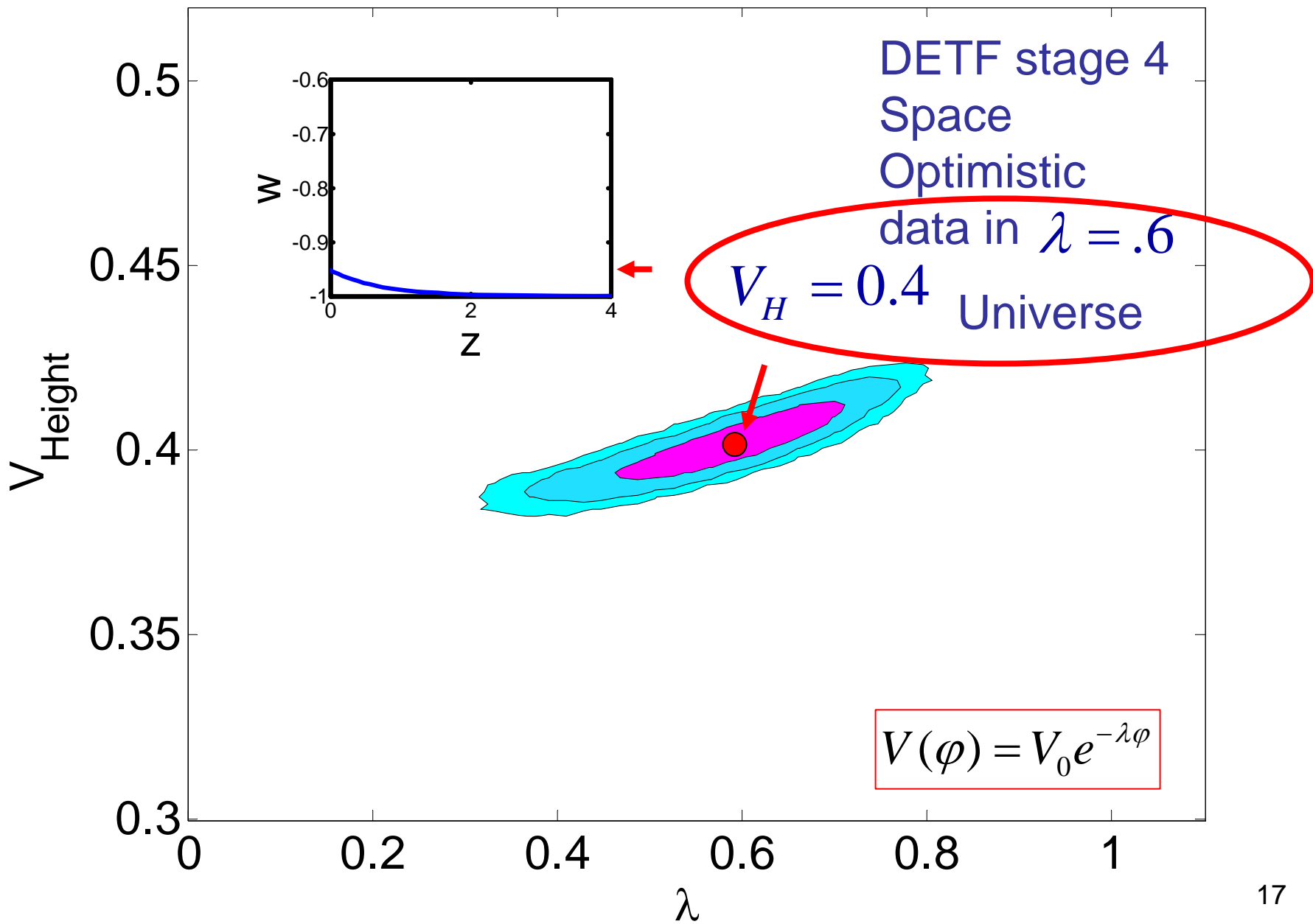
Sneak preview:

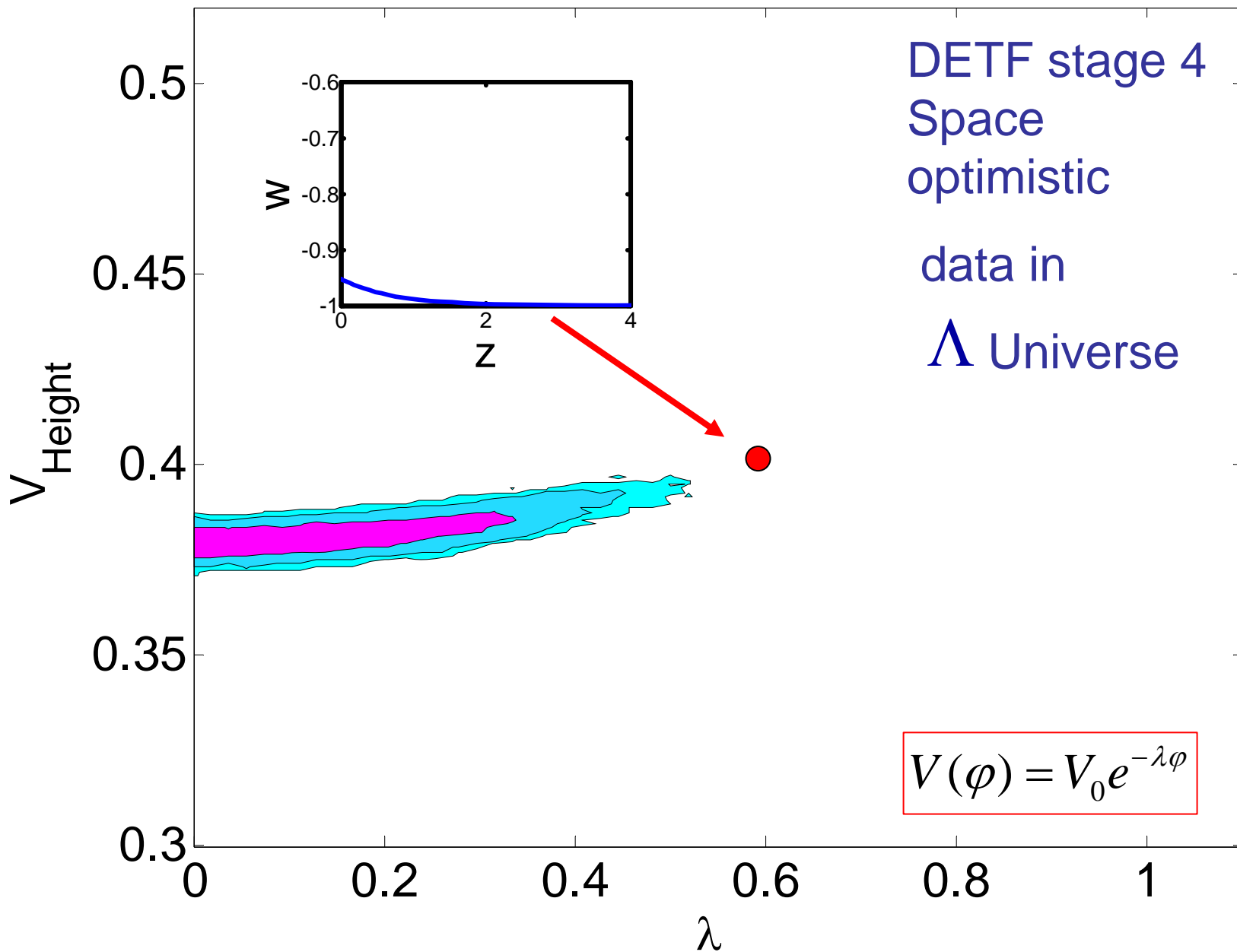


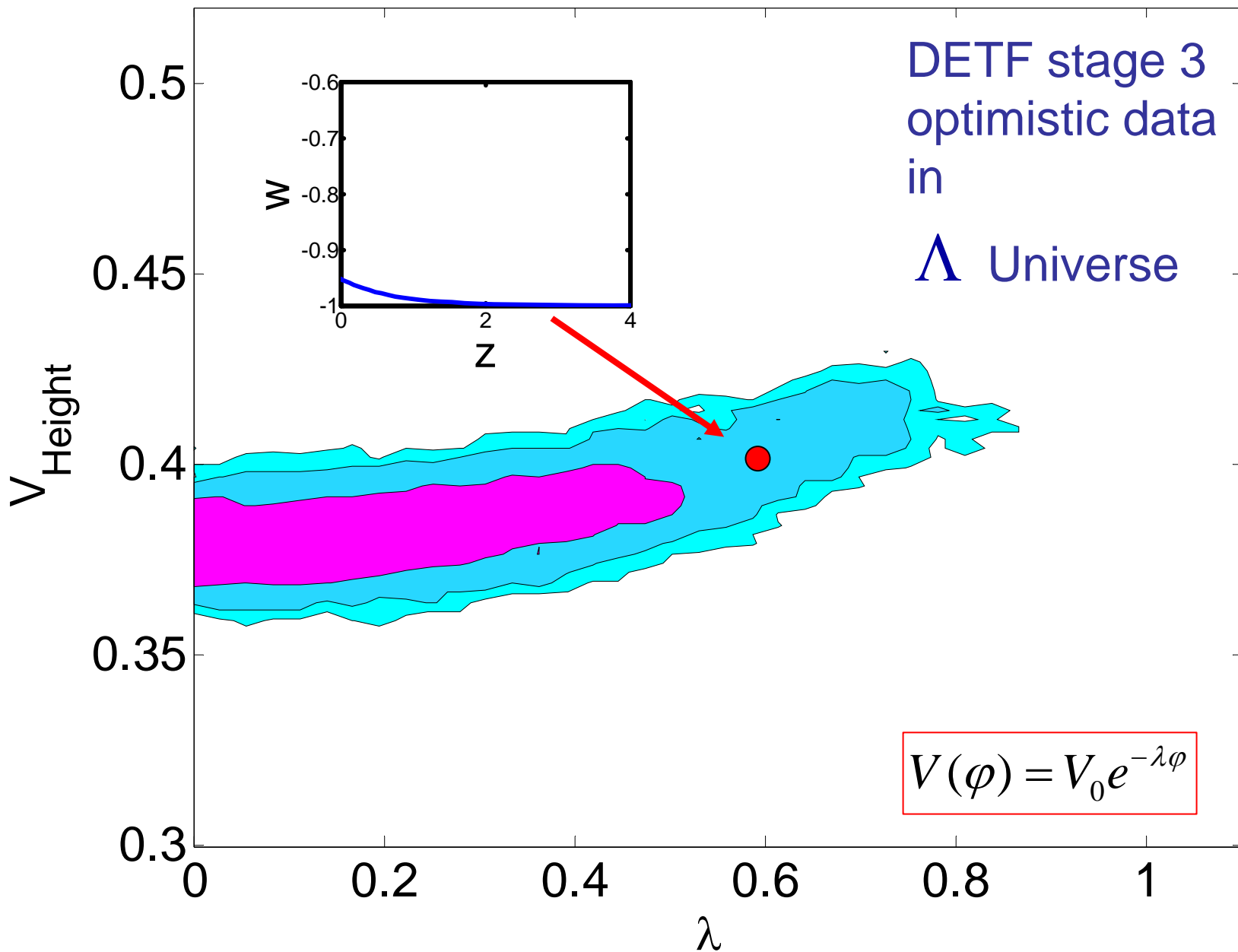


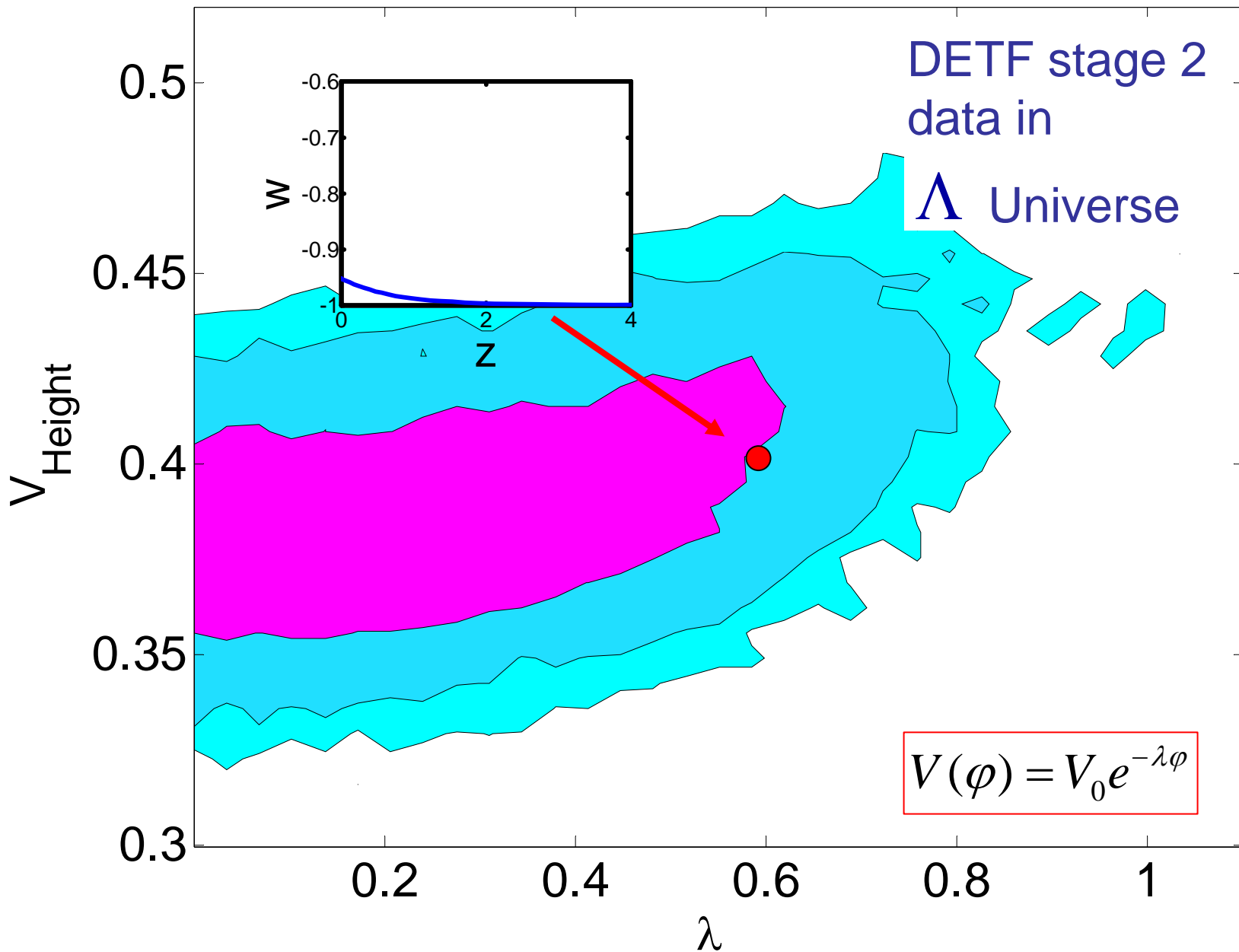












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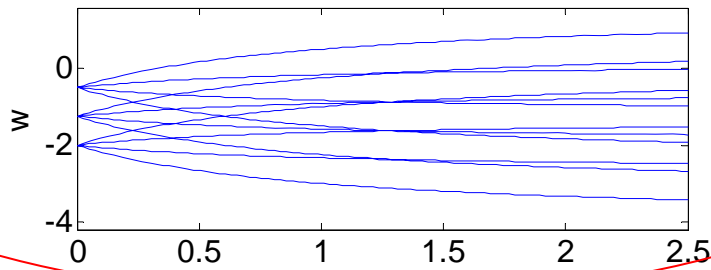
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How good is the $w(a)$ ansatz?

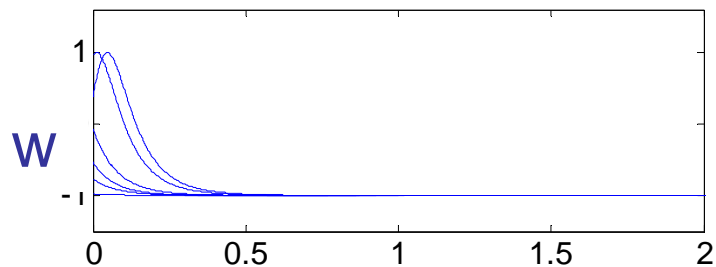
Sample $w(z)$ curves in w_0 - w_a space



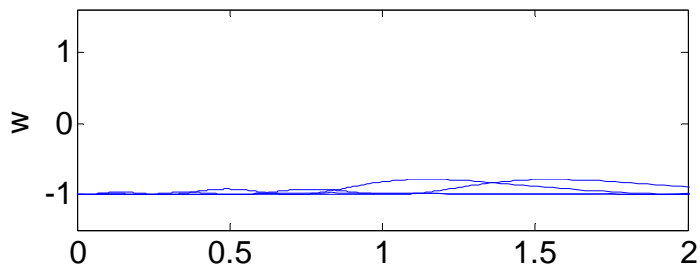
$$w(a) = w_0 + w_a (1 - a)$$

w_0 - w_a can only do these

Sample $w(z)$ curves for the PnGB models



Sample $w(z)$ curves for the EwP models

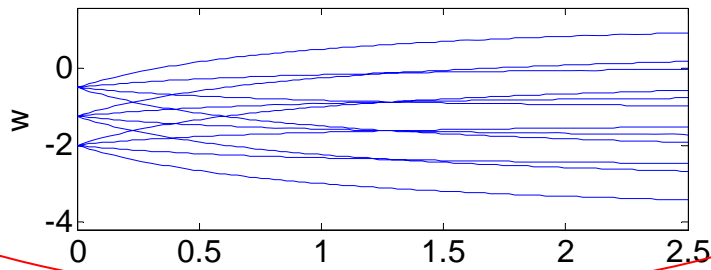


DE models can do this
(and much more)

z

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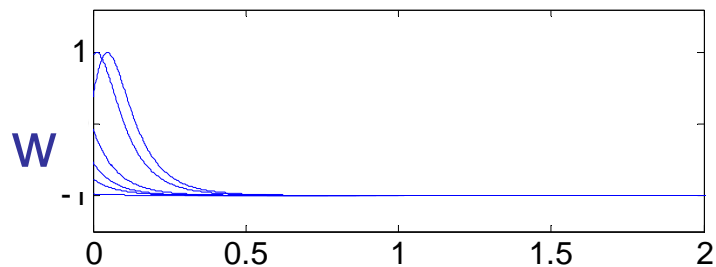
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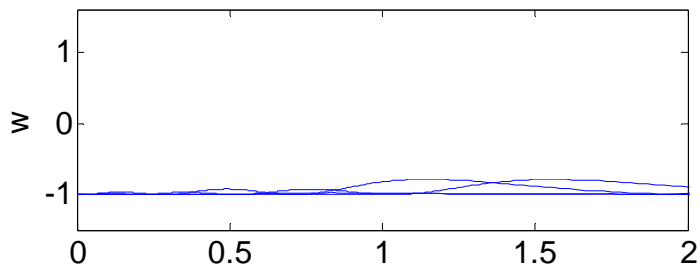
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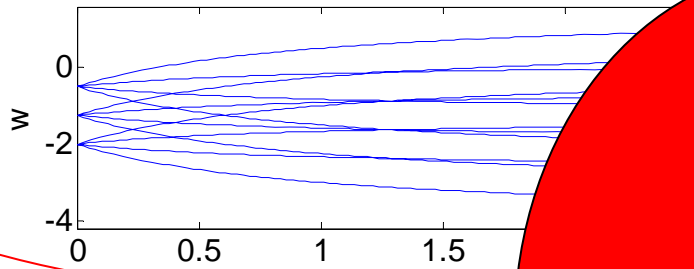


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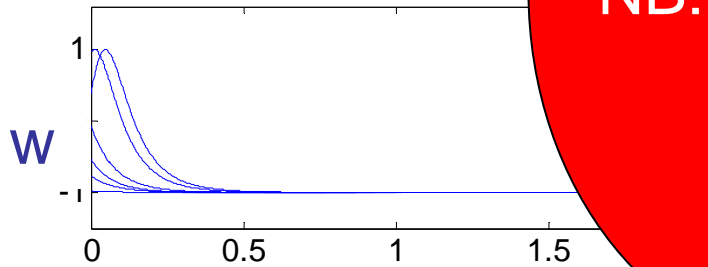
Z

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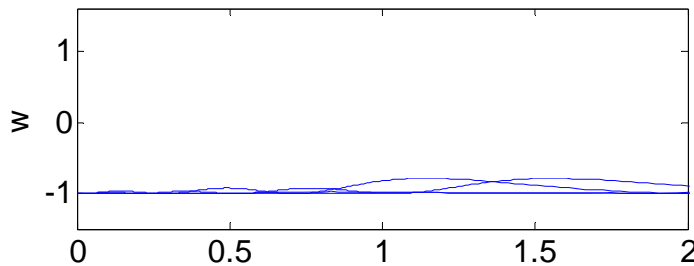
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Sample $w(z)$ curves for the PN



Sample $w(z)$ curves for the EwP models



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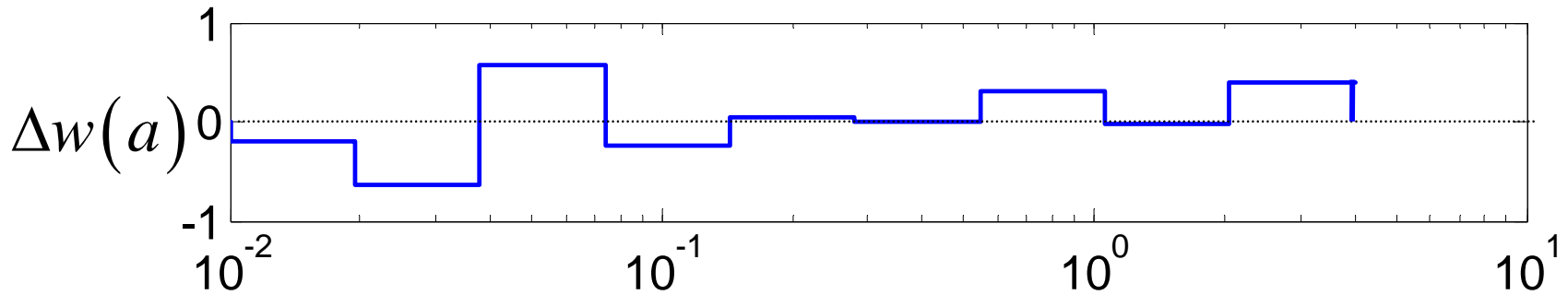
... can only do these

NB: Better than

$$w(a) = w_0$$

DE models can do this
(and much more)

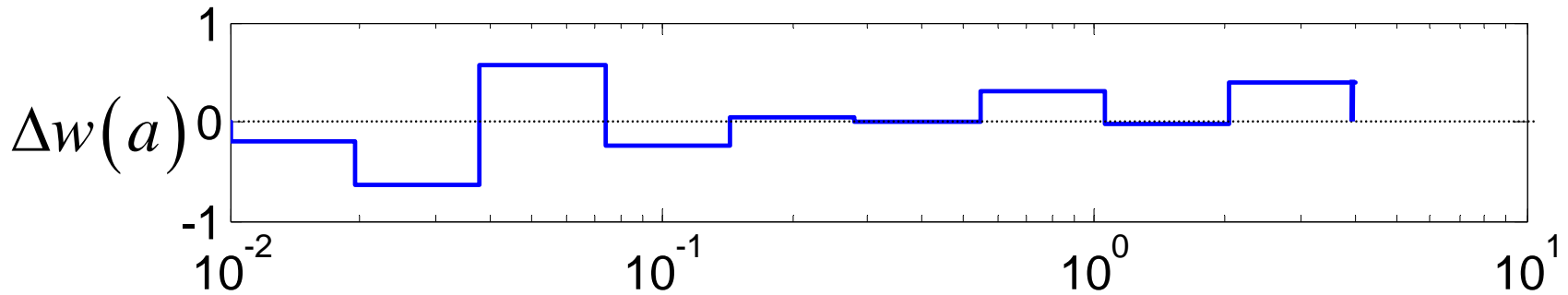
Try 9D stepwise constant $w(a)$



$$w(a) = -1 + \sum_{i=1}^9 w_i T(a_i, a_{i+1})$$

9 parameters are coefficients of the “top hat functions”
 $T(a_i, a_{i+1})$

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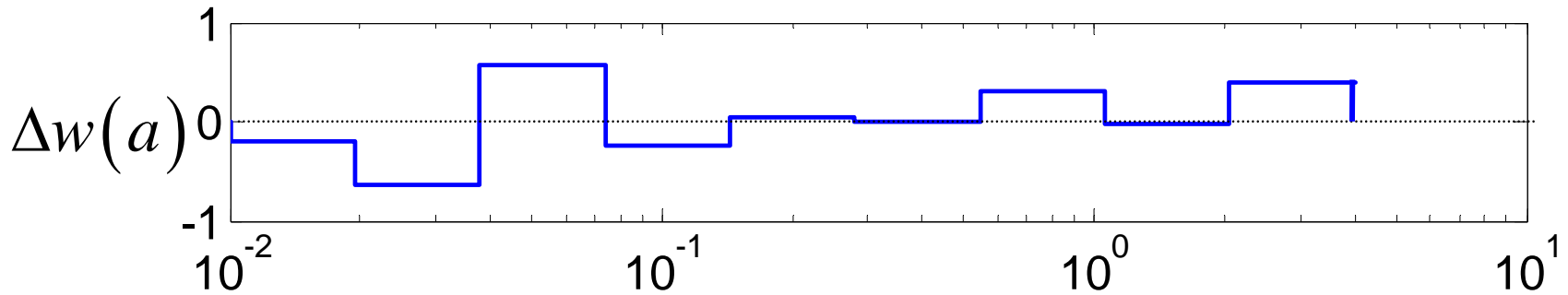
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→ Allows greater variety of $w(a)$ behavior

→ Allows each experiment to “put its best foot forward”

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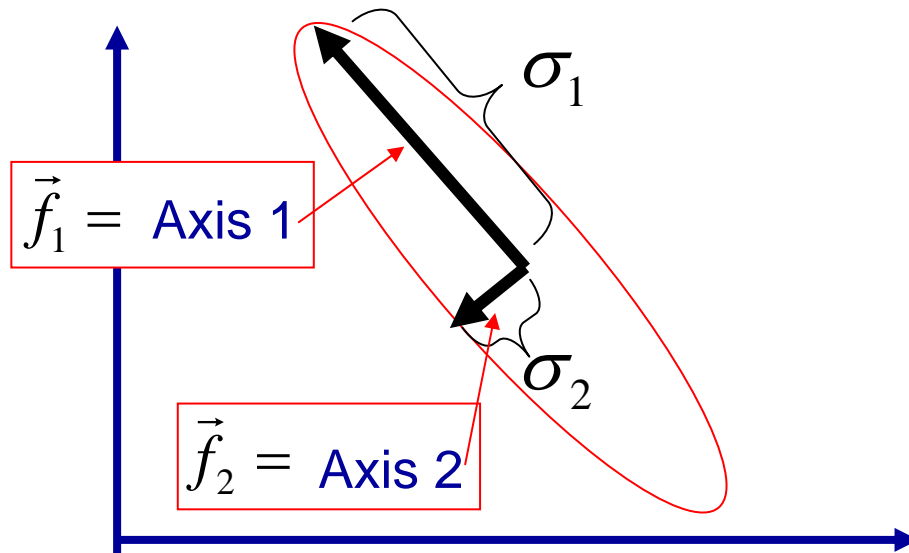
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Q: How do you describe error ellipsis in 9D space?

A: In terms of 9 principle axes \vec{f}_i and corresponding 9 errors σ_i :

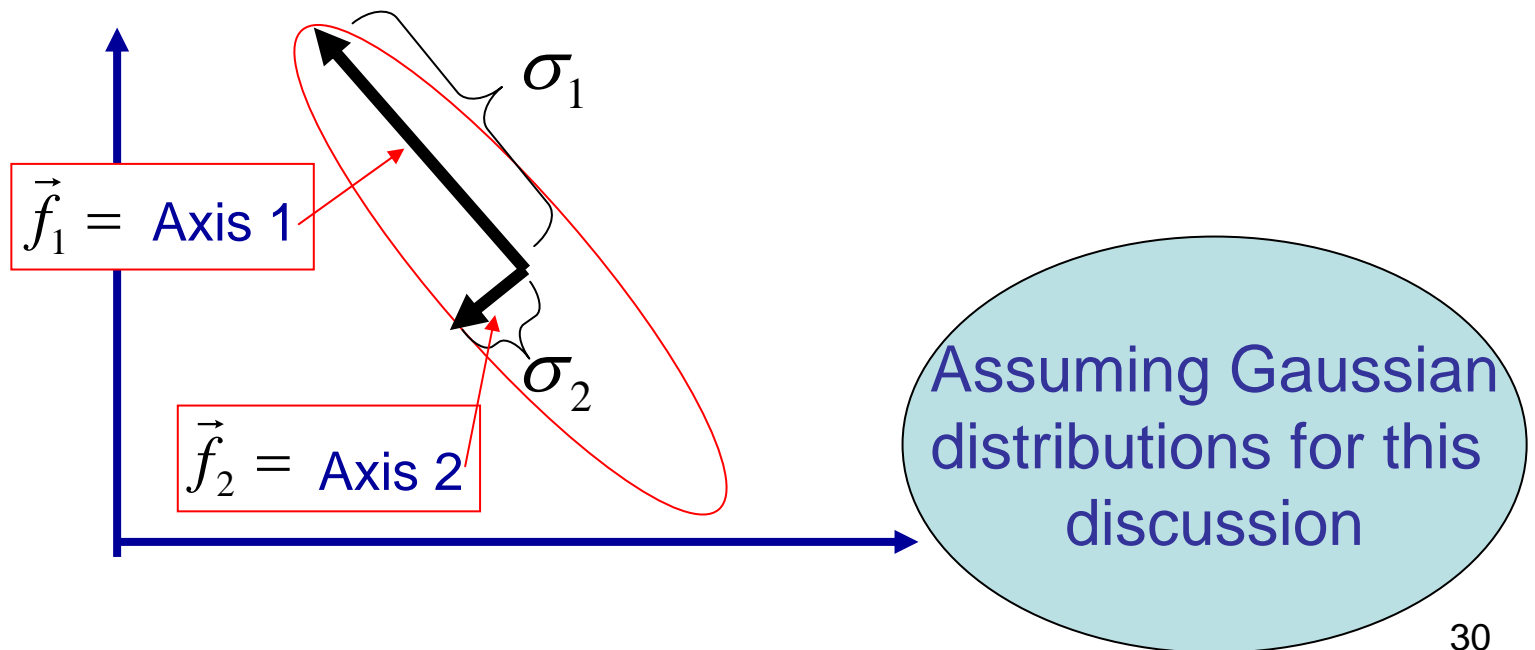
2D illustration:



Q: How do you describe error ellipsis in 9D space?

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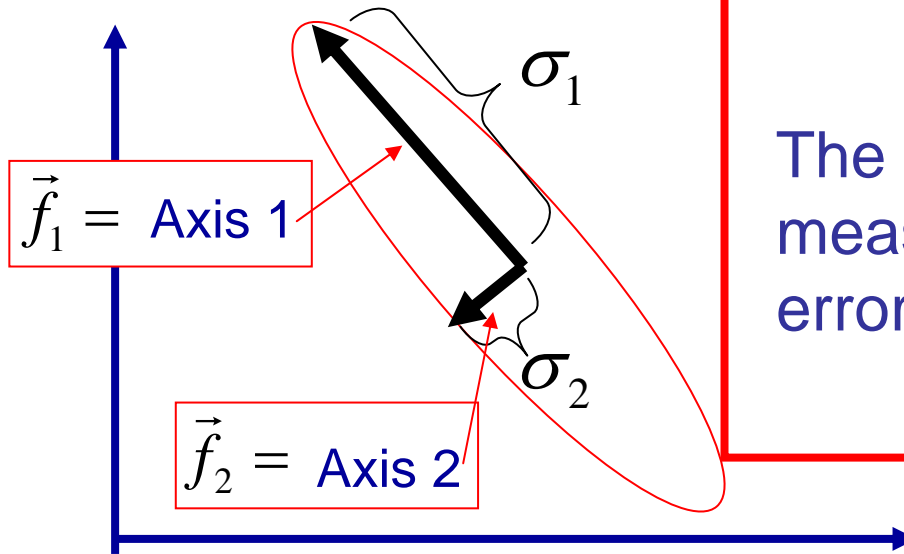
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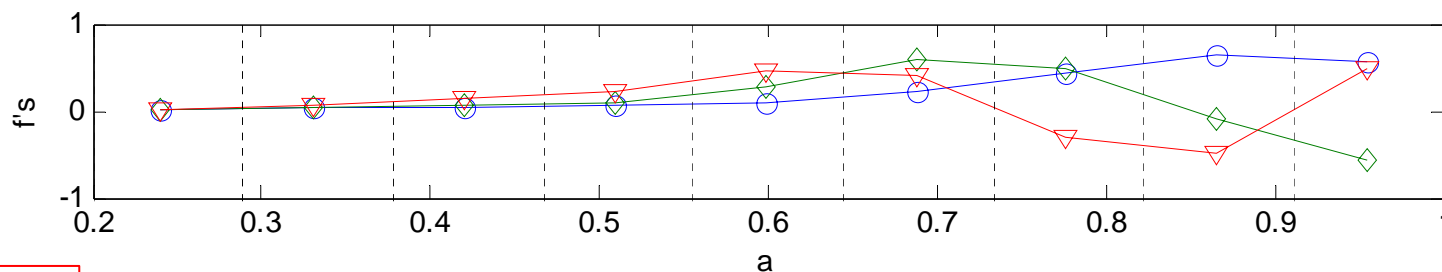
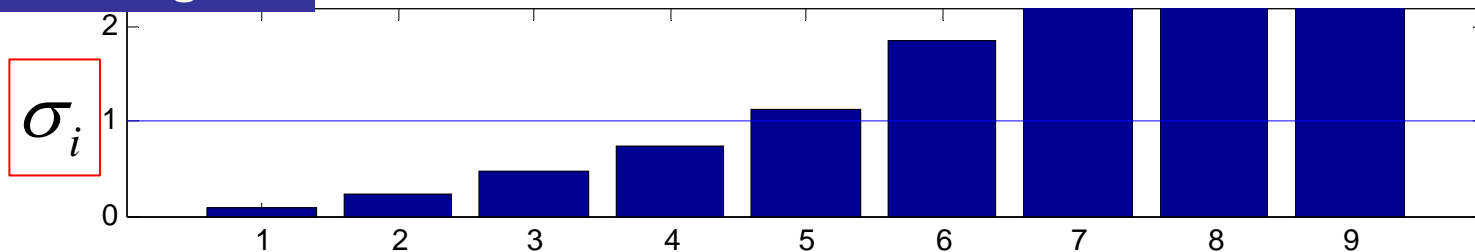
NB: in general the \vec{f}_i s form a complete basis:

$$\vec{w} = \sum_i \alpha_i \vec{f}_i$$

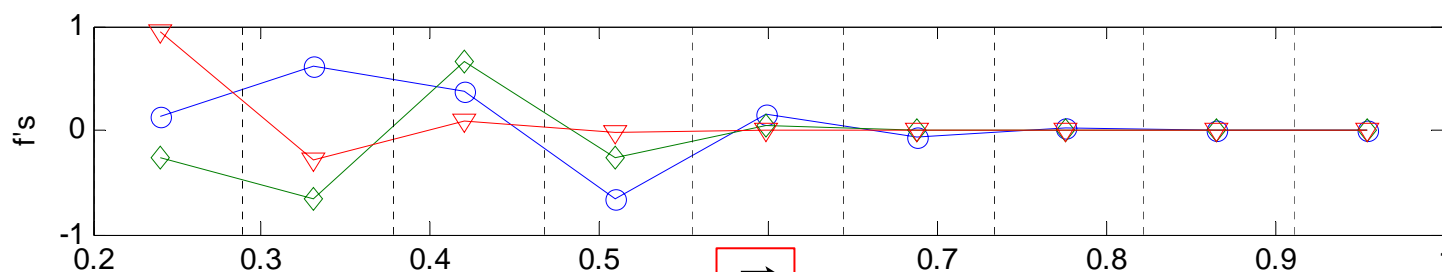
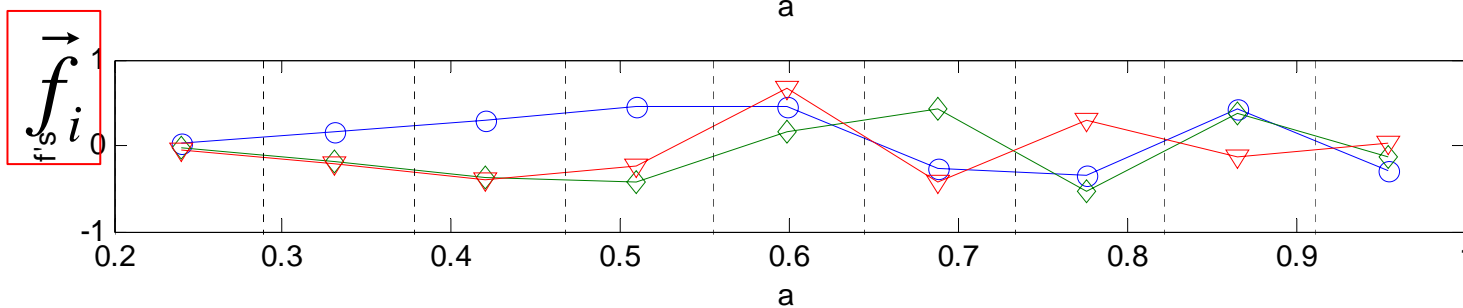
The α_i are independently measured qualities with errors σ_i

Characterizing 9D ellipses by principle axes and corresponding errors

DETF stage 2



Principle Axes

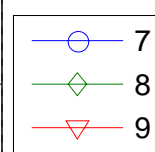
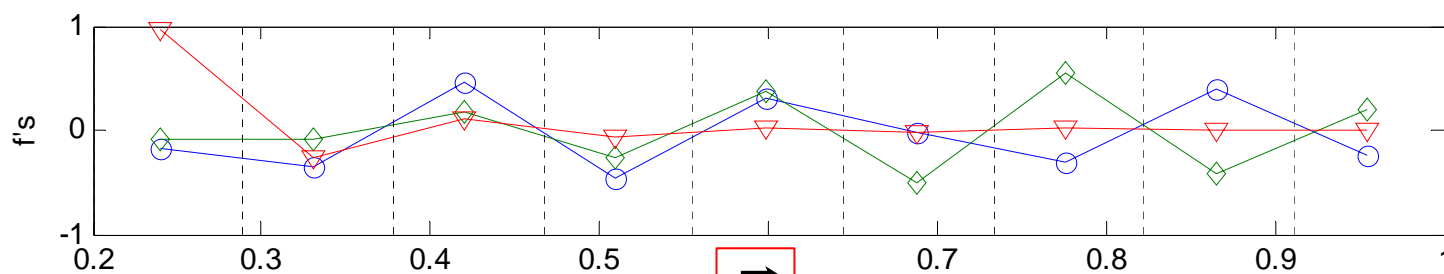
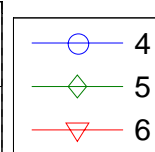
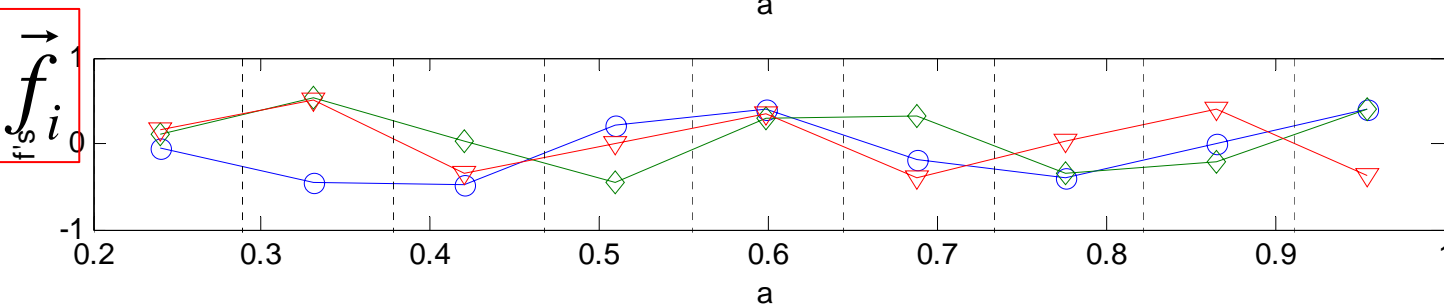
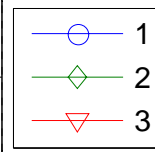
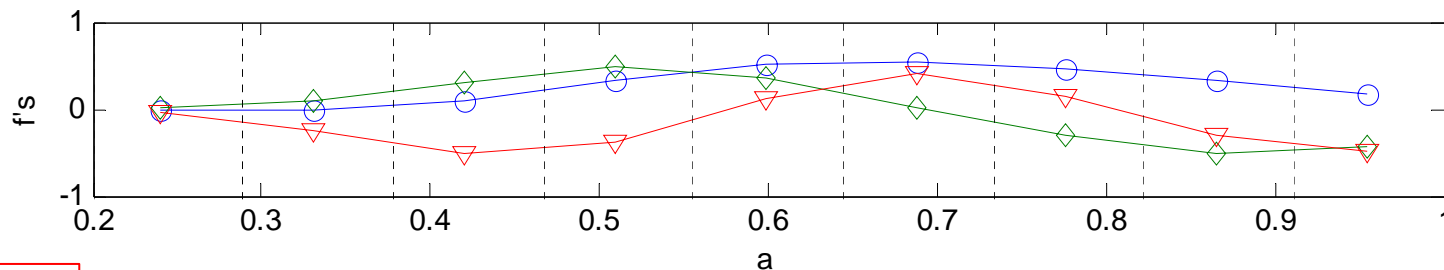
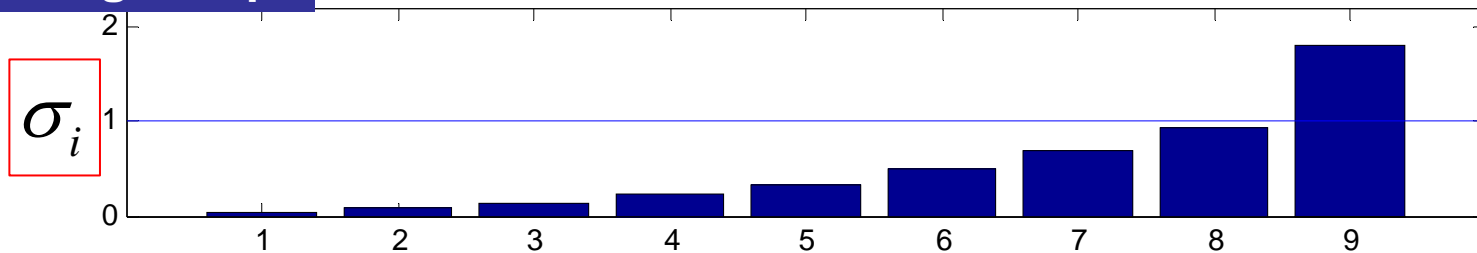


\vec{a}

i

Characterizing 9D ellipses by principle axes and corresponding errors

WL Stage 4 Opt



\vec{a}

i

Principle Axes

σ_i

$f's_i$

f's

2

0

1

2

3

4

5

6

7

8

9

f's

-1

0

1

a

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f's

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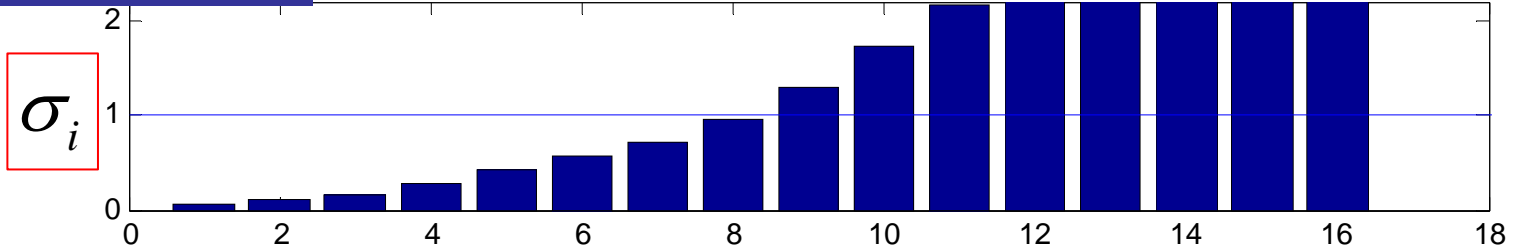
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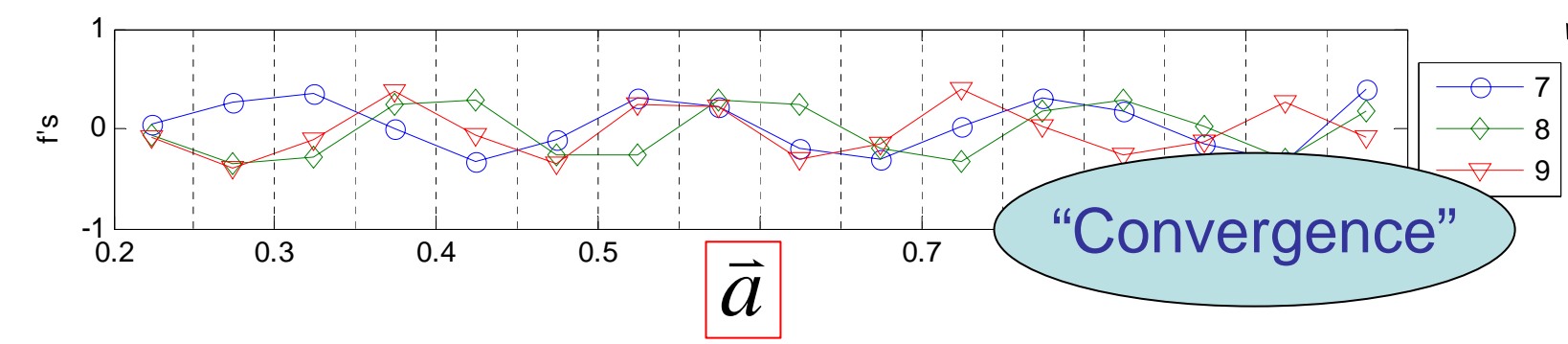
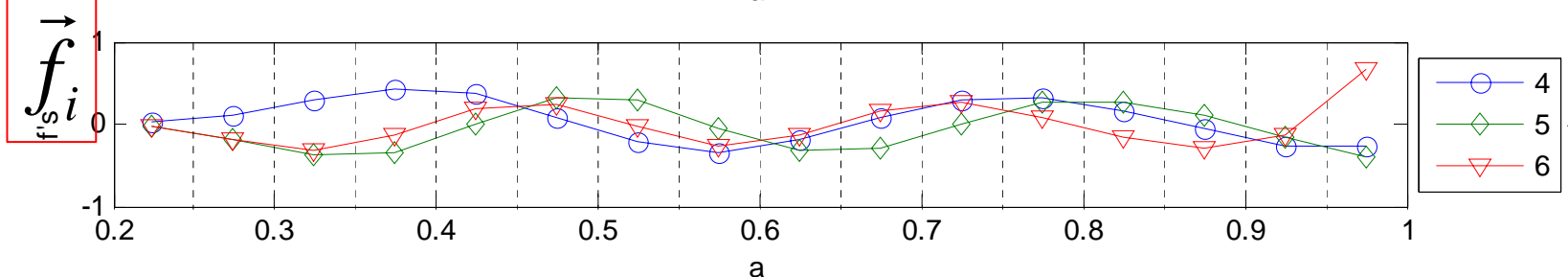
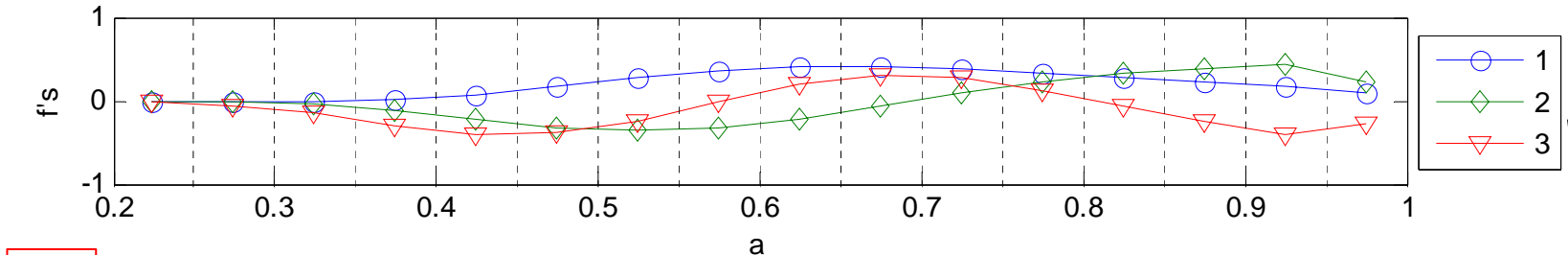
0.3

Characterizing 9D ellipses by principle axes and corresponding errors

WL Stage 4 Opt



Principle Axes



i

DETF Figure of Merit:

$$\mathcal{F}_{\text{DETF}} = \frac{1}{\sigma_1 \sigma_2}$$

9D Figure of Merit:

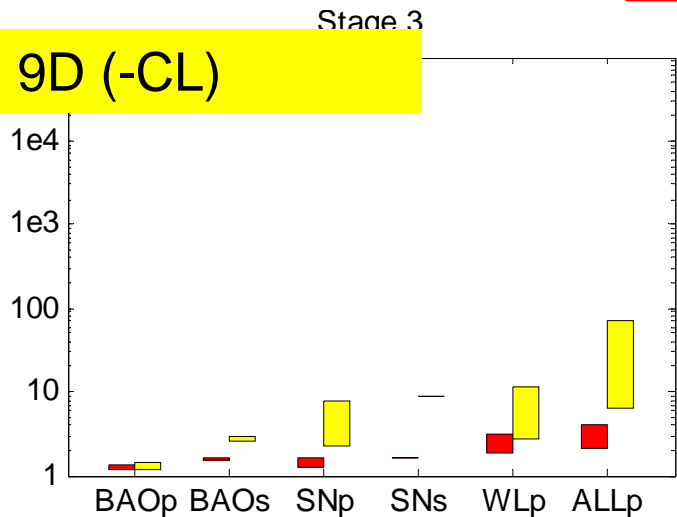
$$\mathcal{F}_{9\text{D}} = \frac{1}{\prod_{i=1}^9 \sigma_i}$$

If $\sigma_i > 1$ we set $\sigma_i = 1$

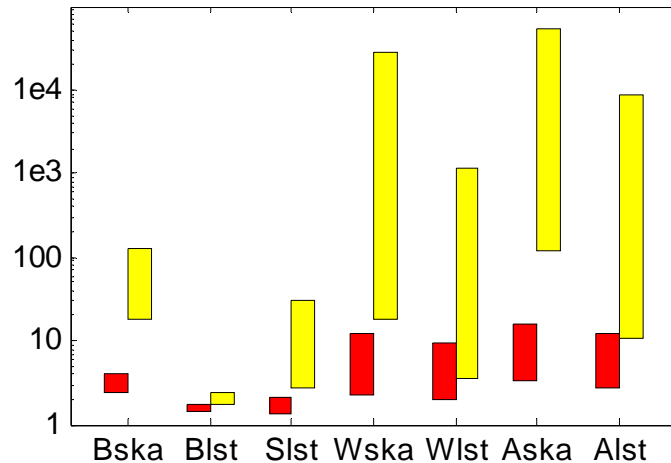
DETF(-CL)

$\mathcal{F}_{\text{DETF/9D}}$

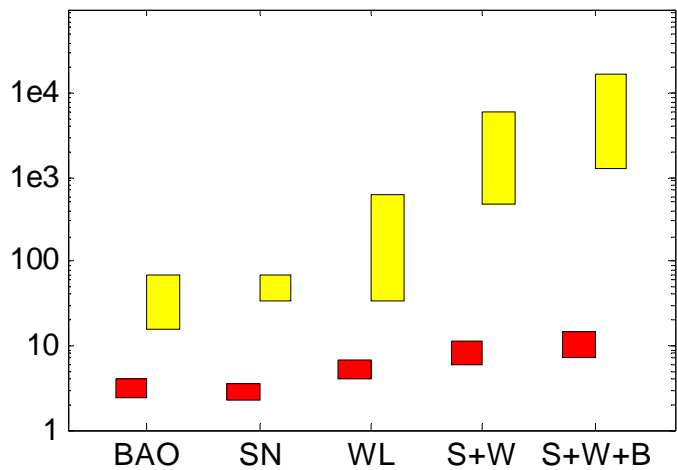
9D (-CL)



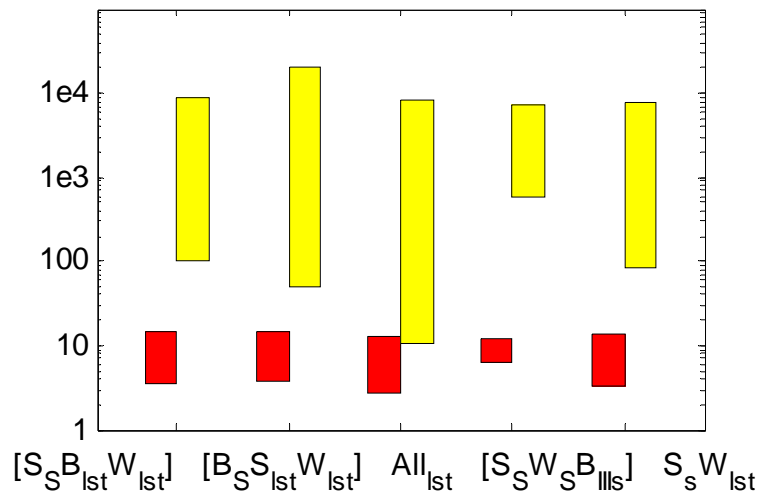
Stage 4 Ground



Stage 4 Space

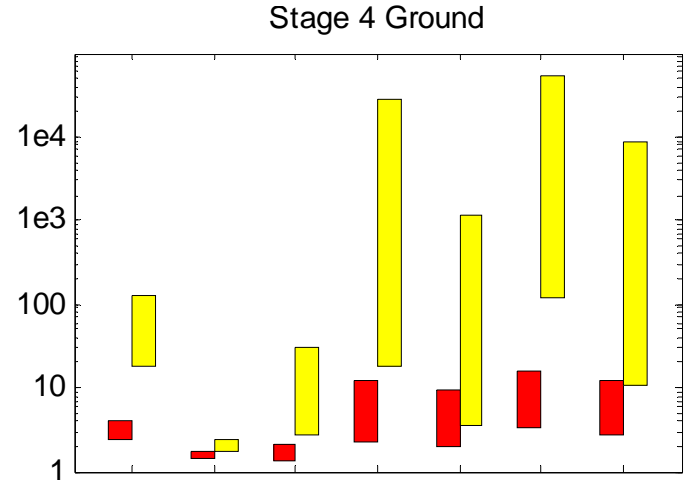
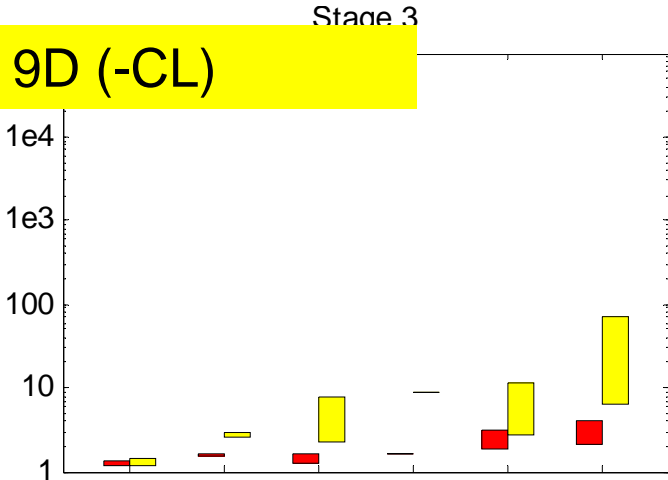


Stage 4 Ground+Space

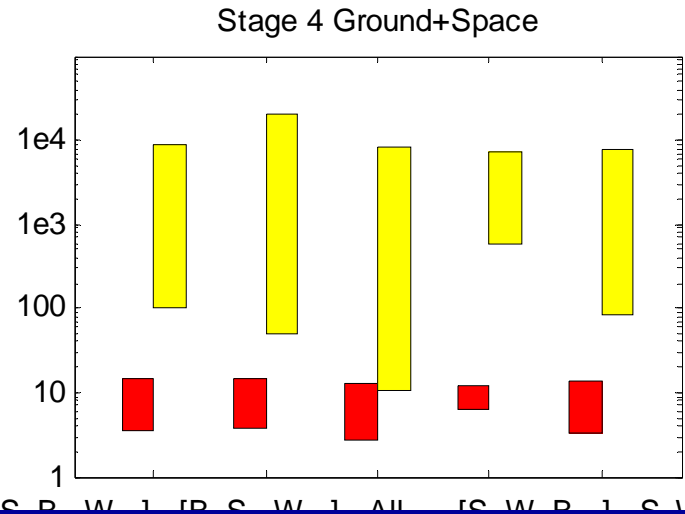
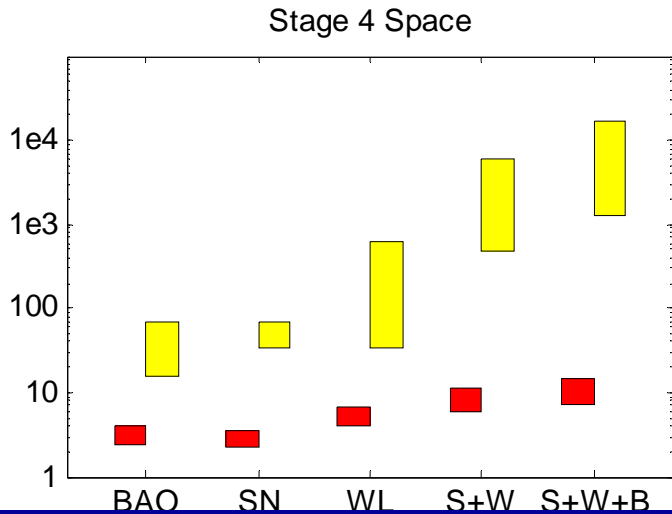


DETF(-CL)

$\mathcal{F}_{\text{DETF/9D}}$



Stage 2 \rightarrow Stage 3 = 1 order of magnitude (vs 0.5 for DETF)



Stage 2 \rightarrow Stage 4 = 3 orders of magnitude (vs 1 for DETF)

Define the “scale to 2D” function

$$\mathcal{S}_{2D}(\mathcal{F}) \equiv \mathcal{F}^{2/D_e}$$

$$\mathcal{S}_{2D}(\mathcal{F}) \equiv \frac{1}{\sigma_{ave}^2}$$

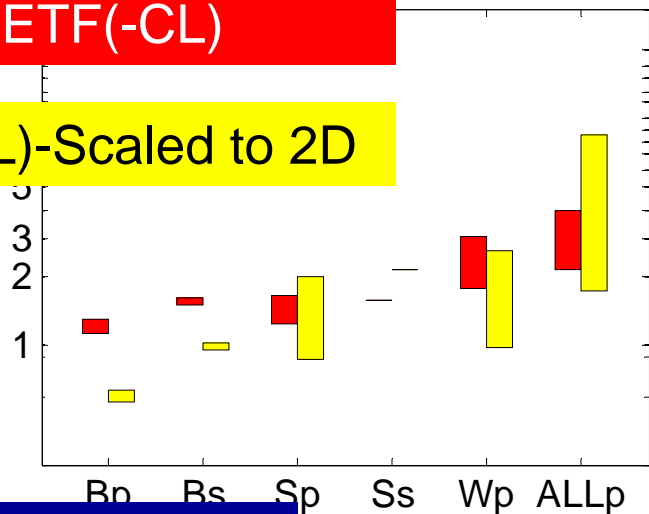
→ The idea: Construct an effective 2D FoM by assuming two dimensions with “average” errors (~geometric mean of 9D errors)

→ Purpose: Separate out the impact of higher dimensions on comparisons with DETF, vs other information from the D9 space (such relative comparisons of data model).

DETF(-CL)

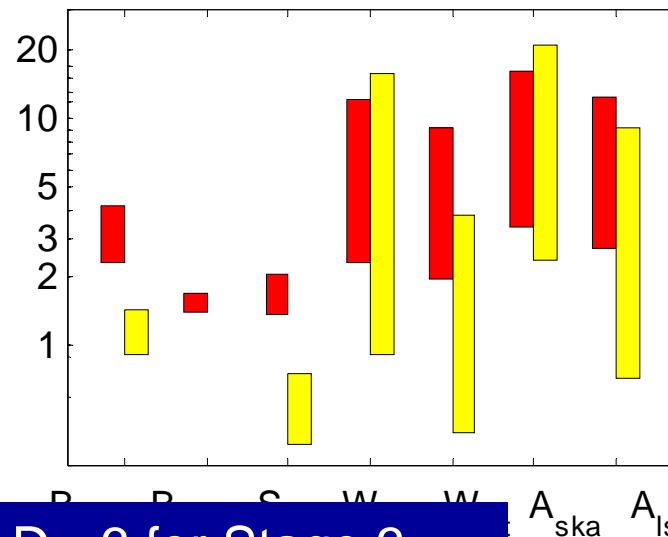
9D (-CL)-Scaled to 2D

Stage 3



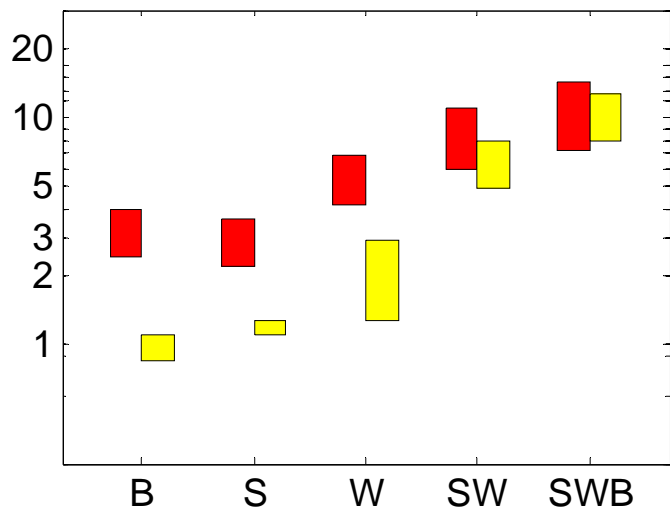
$D_e=2.5$ for Stage 2

Stage 4 Ground

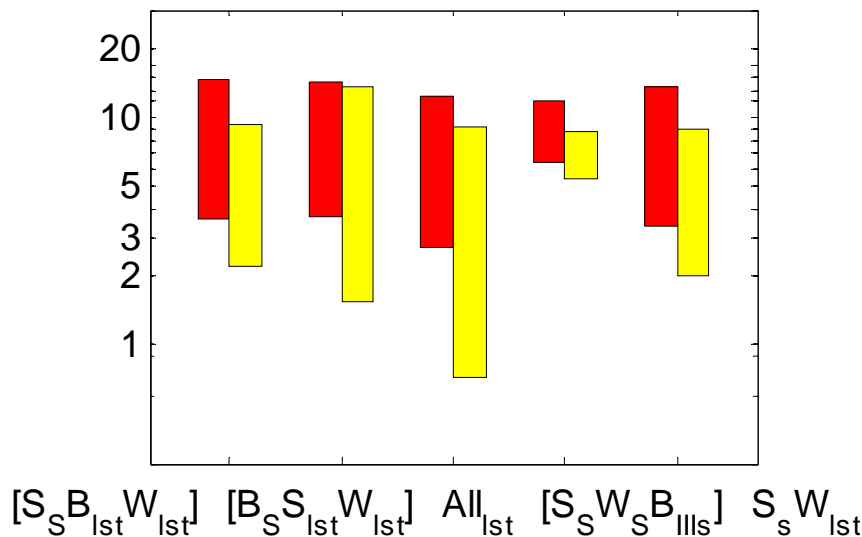


$D_e=3$ for Stage 3

Stage 4 Space

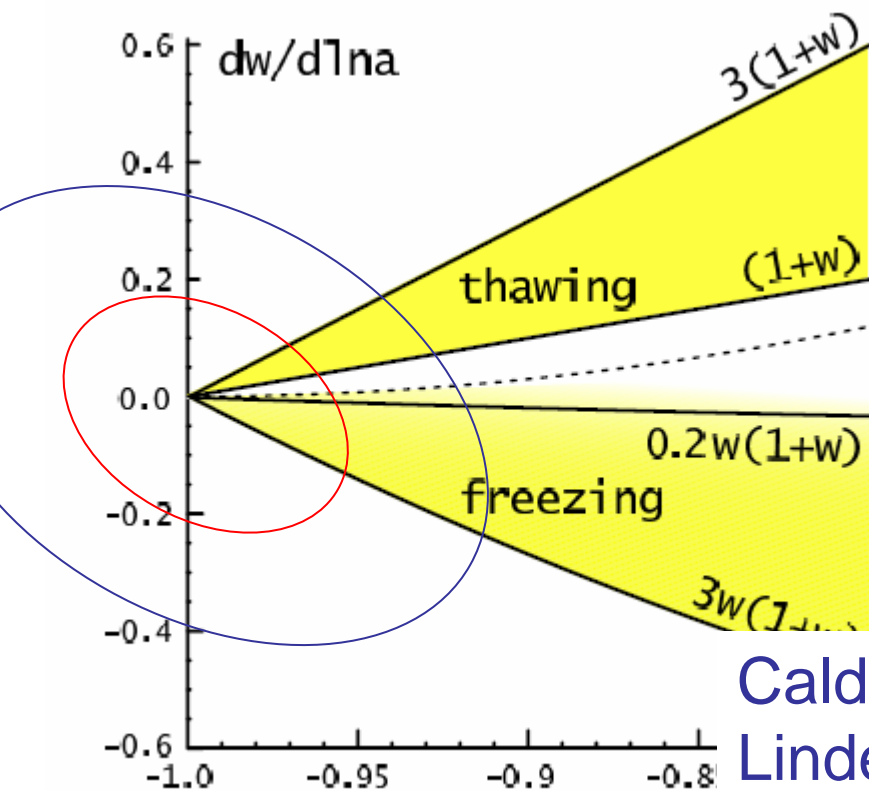


Stage 4 Ground+Space

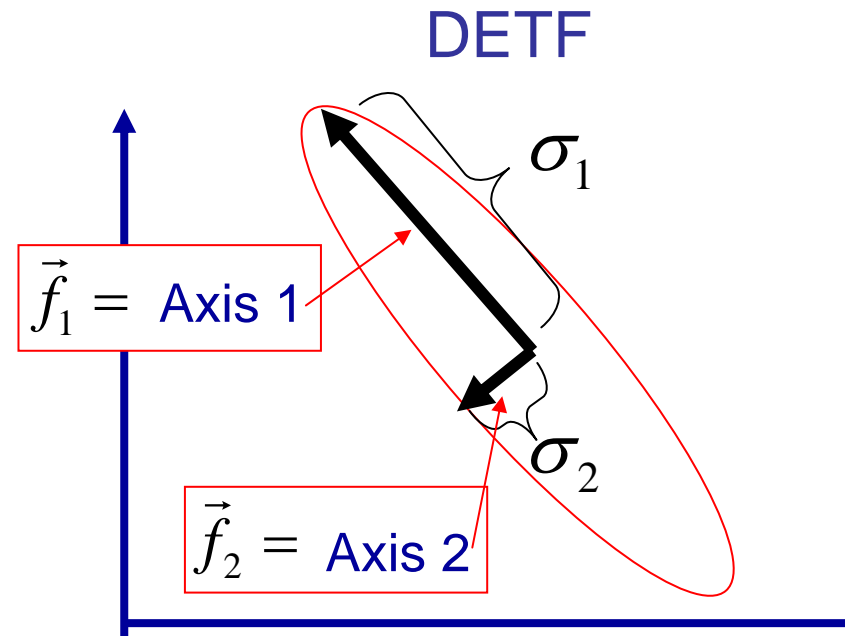


$D_e= 4$ for Stage 4 Pes, ; $D_e= 4.5$ for Stage 4 Opt,

Discussion of cost/benefit analysis should take place in higher dimensions (vs current standards)



Caldwell & Linder



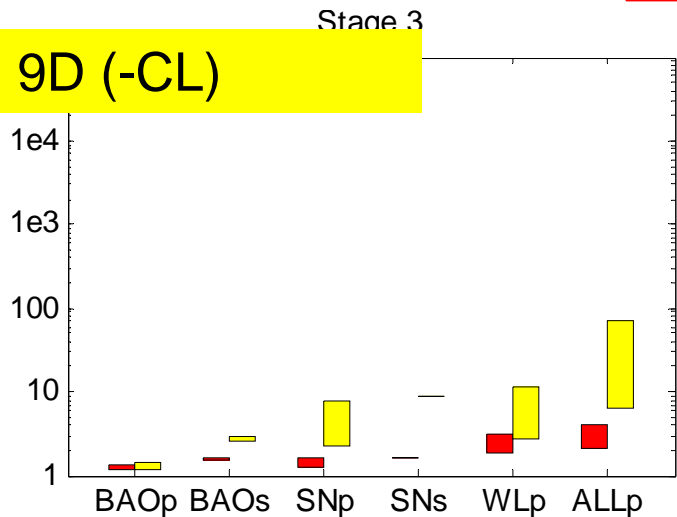
An example of the power of the principle component analysis:

Q: I've heard the claim that the DETF FoM is unfair to BAO, because w_0 - w_a does not describe the high- z behavior which to which BAO is particularly sensitive. Why does this not show up in the 9D analysis?

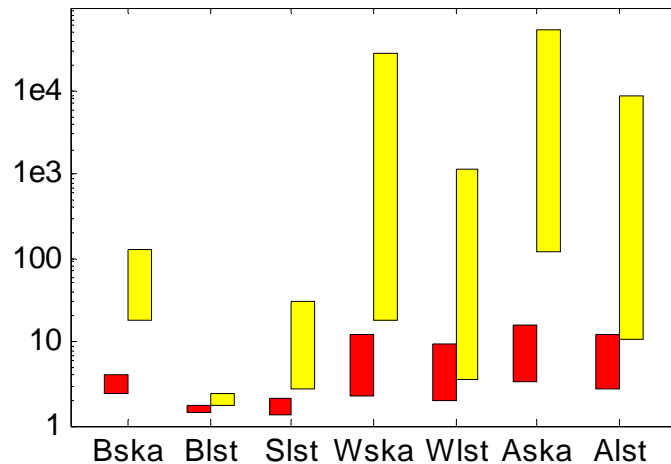
DETF(-CL)

$\mathcal{F}_{\text{DETF/9D}}$

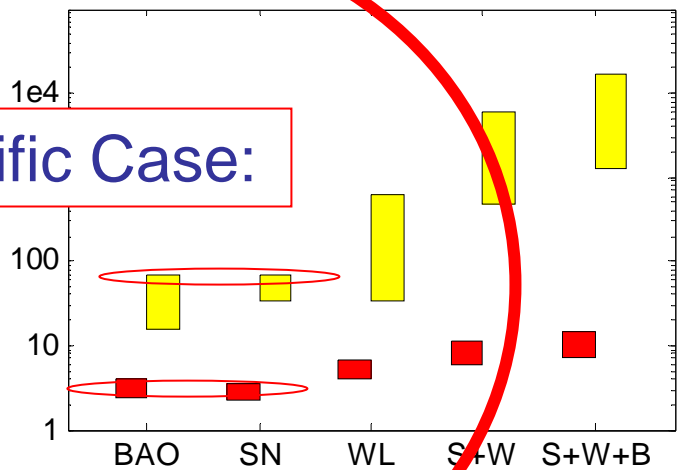
9D (-CL)



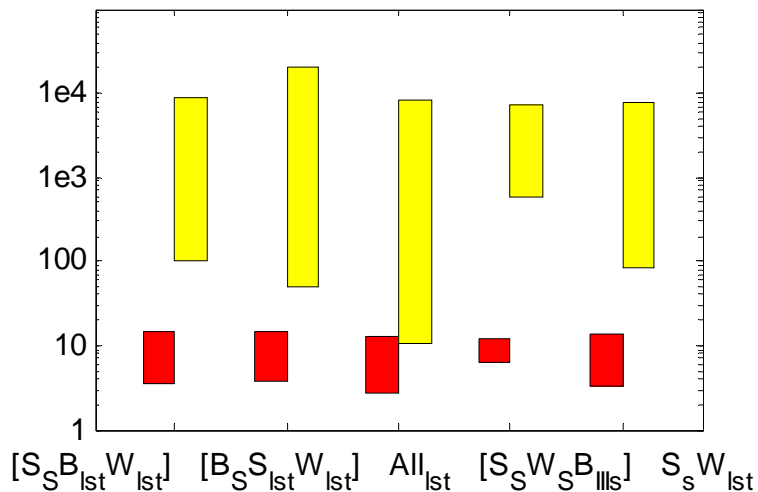
Stage 4 Ground



Stage 4 Space



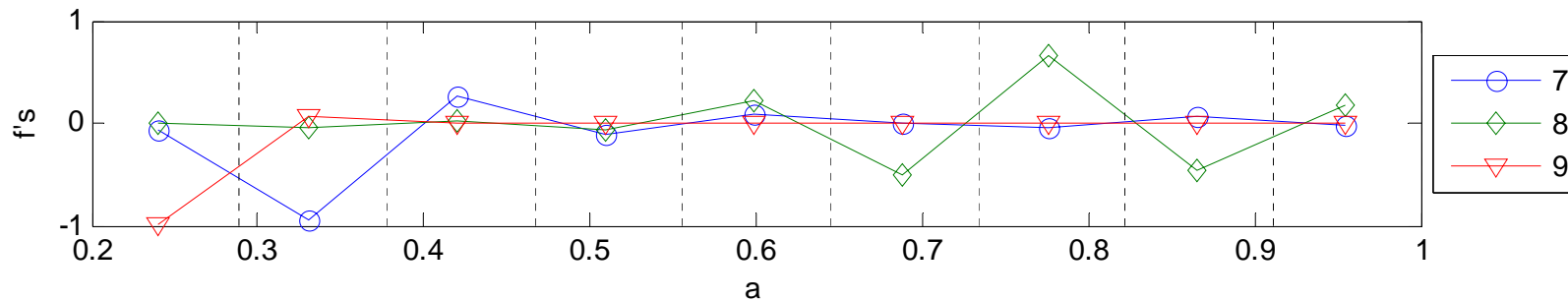
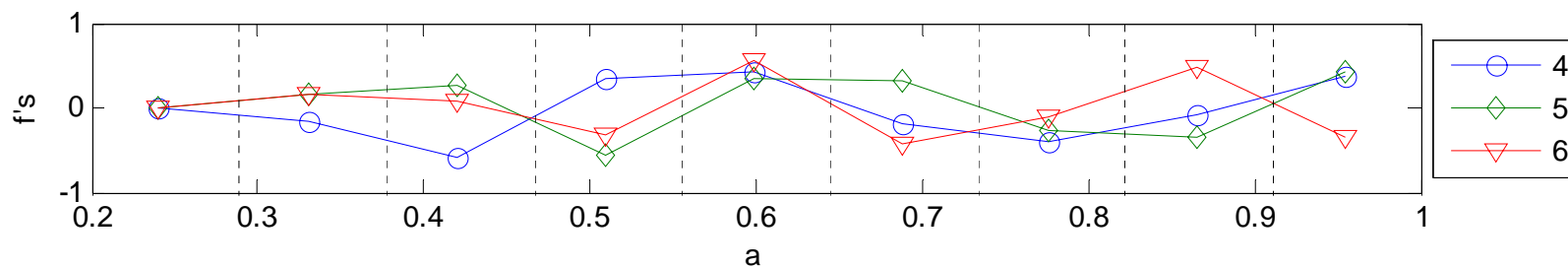
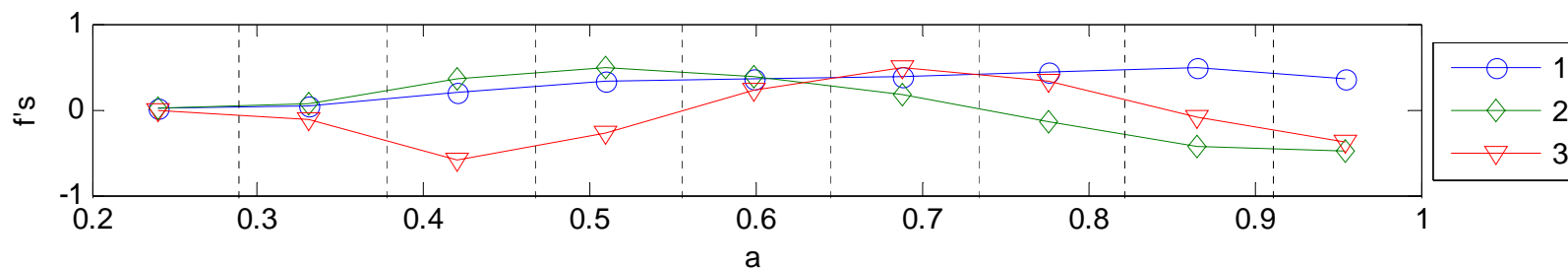
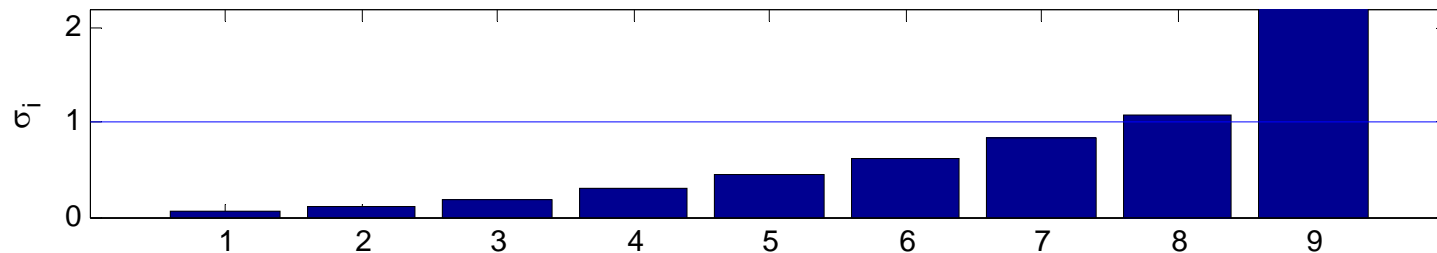
Stage 4 Ground+Space

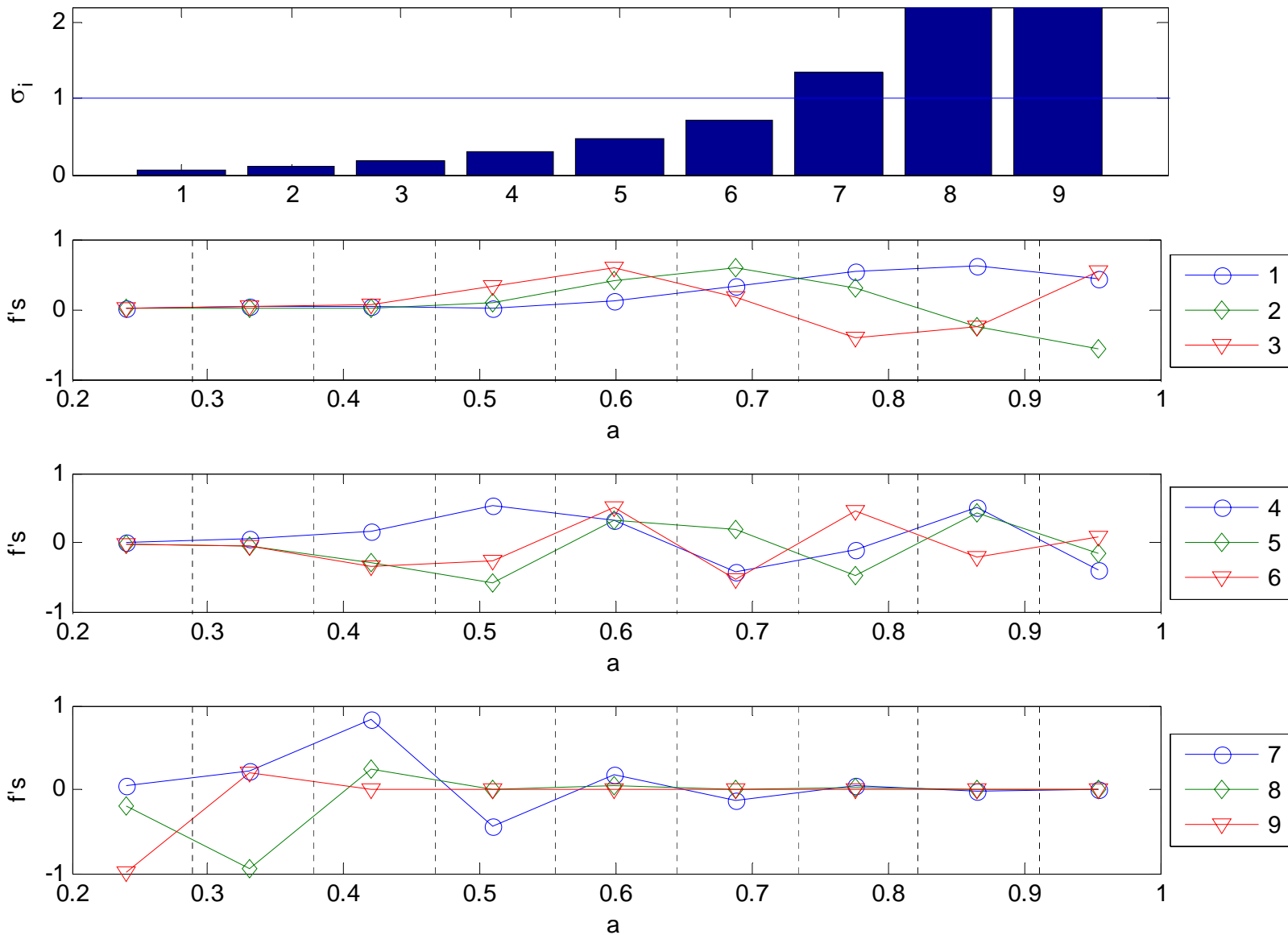


Specific Case:

BAO

Stage 4 Space BAO Opt; lin-a $N_{\text{Grid}} = 9$, $z_{\text{max}} = 4$, Tag = 044301

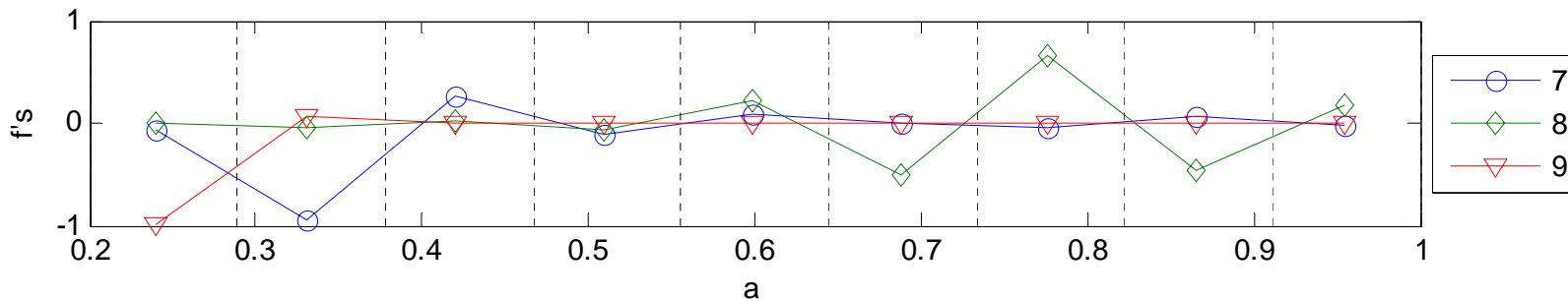
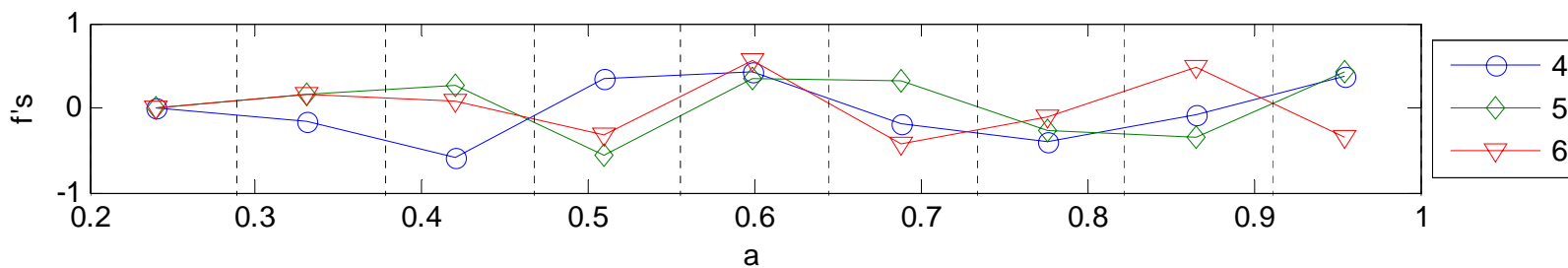
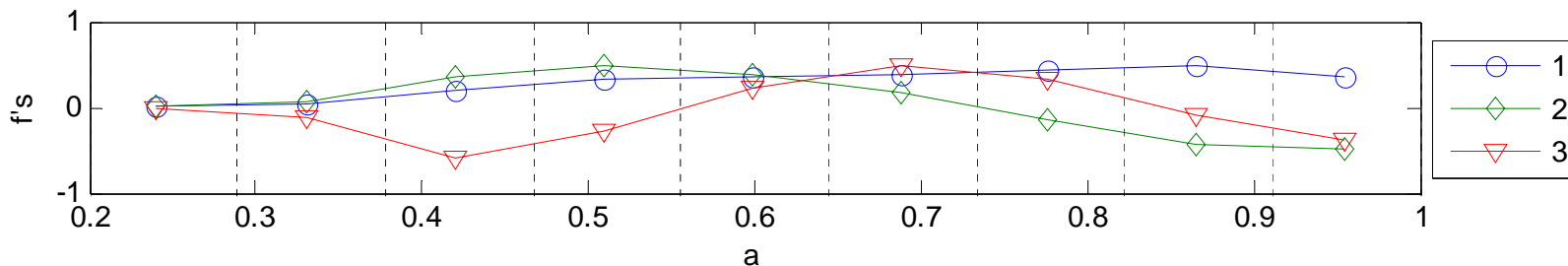
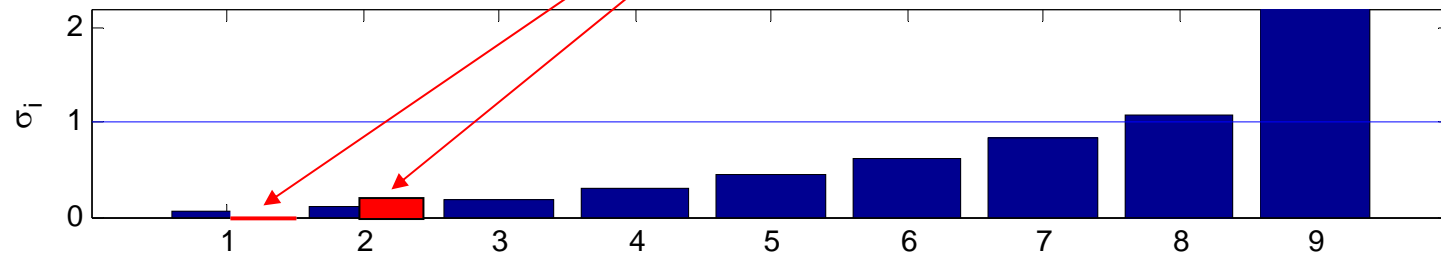


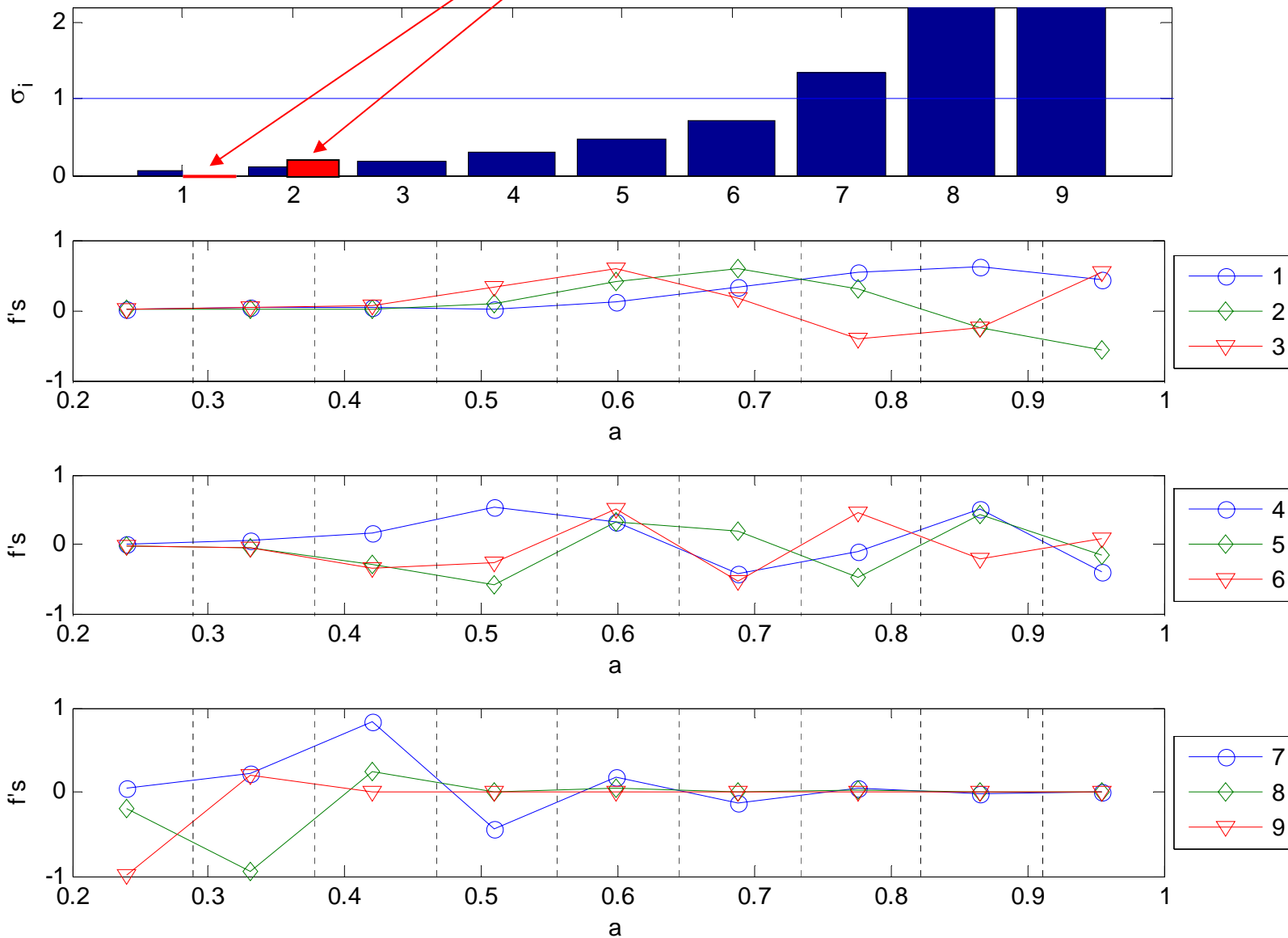
Stage 4 Space SN Opt; lin-a $N_{\text{Grid}} = 9$, $z_{\text{max}} = 4$, Tag = 044301

BAO

DETF σ_1, σ_2

Grid = 9, $z_{\max} = 4$, Tag = 044301

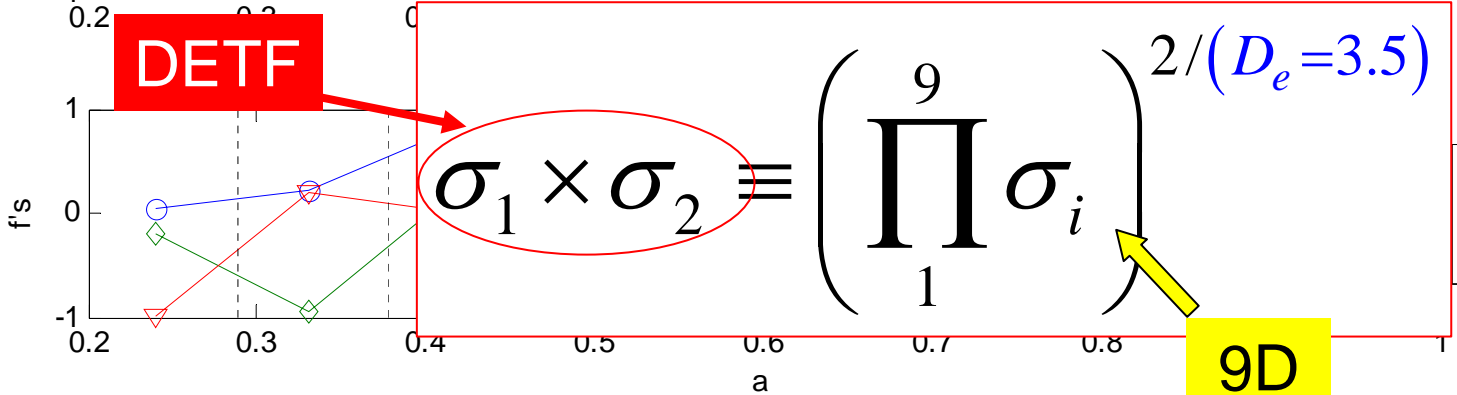
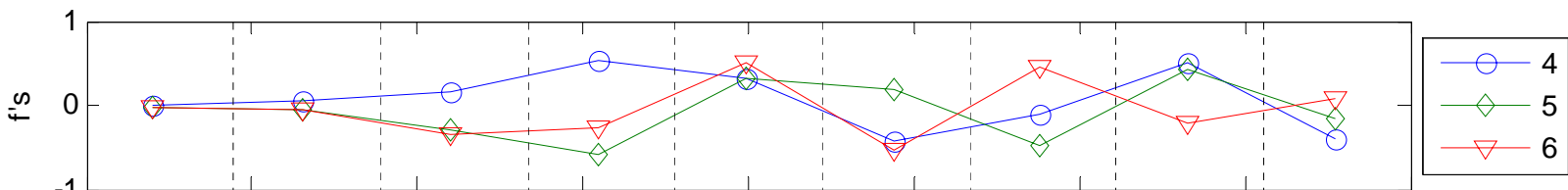
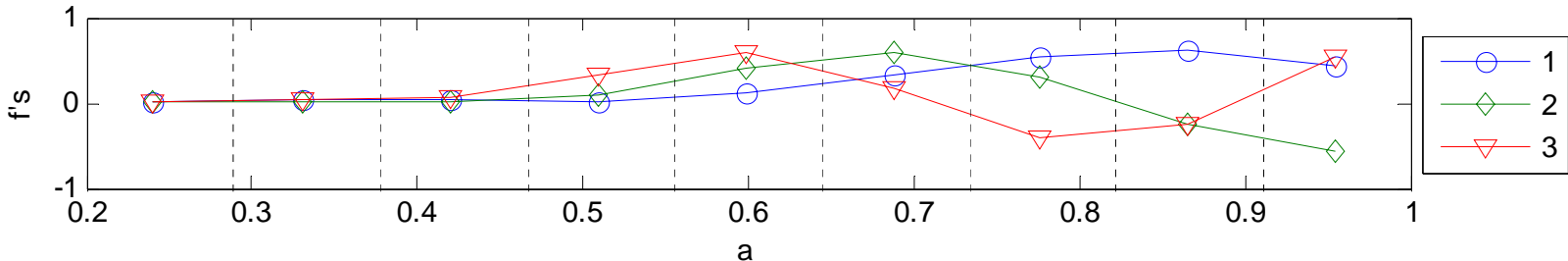
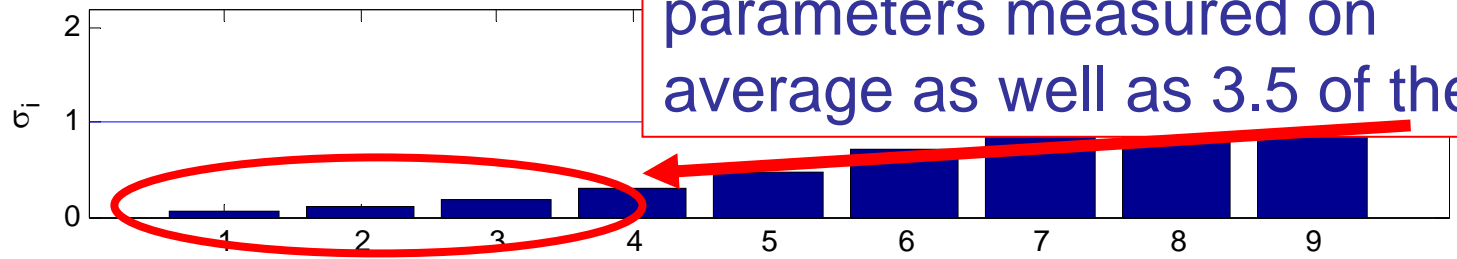


Stage 4 Space SN Cpr; lin-a $N_{\text{Grid}} = 9$, $z_{\text{max}} = 4$, Tag = 044301

SN

Stage 4 Space SN Op

w0-wa analysis shows two parameters measured on average as well as 3.5 of these



DETF

$$\sigma_1 \times \sigma_2 \equiv \left(\prod_{i=1}^9 \sigma_i \right)^{2/(D_e=3.5)}$$

9D

Quick and dirty check on the impact of future data on your DE model:

- 1) Choose a data model and generate \vec{f}_i 's
- 2) Choose representative functions $w(a)$ for your model
- 3) Find α_i coefficients in $\vec{w} = \sum_i \alpha_i \vec{f}_i$
- 4) The σ_i tell you how well the α_i will be measured

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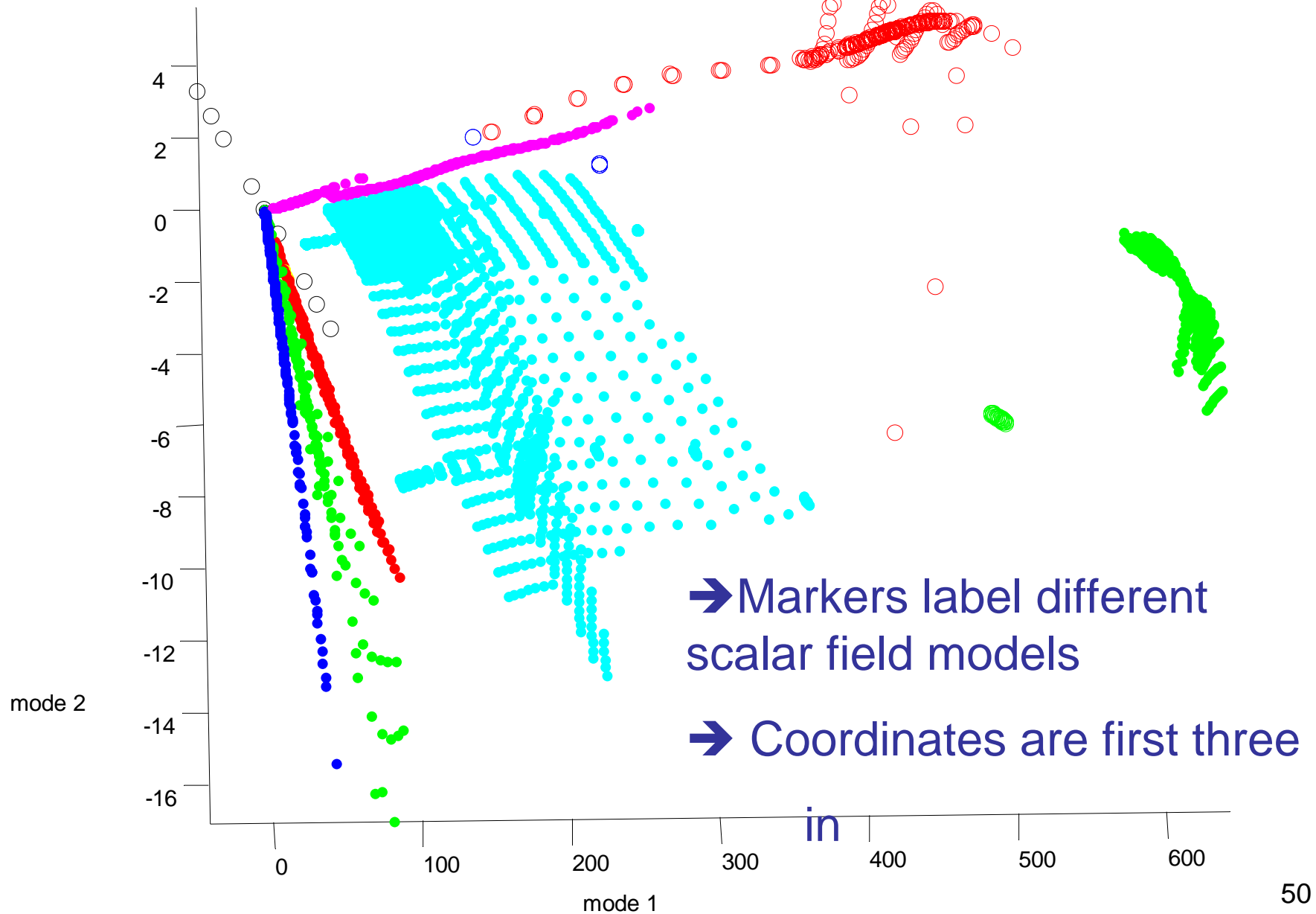
2) Choose representative functions $w(a)$ model

Or grab them from our web site

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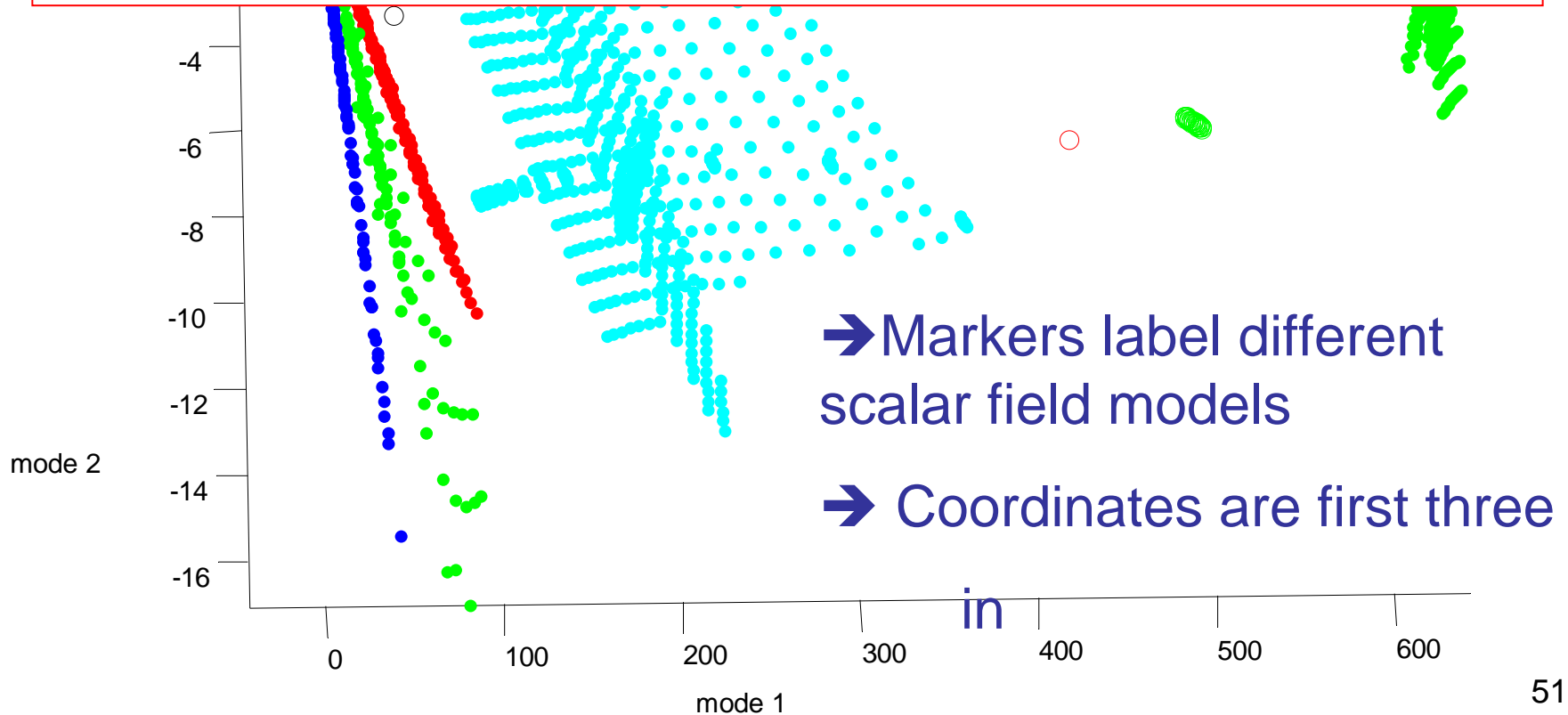
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sigMax = 4 sigMin = 0 OneModel = 0 OneVersionP1 = 0 OneRun = 0 EigenSR14 AllSolsV2₂





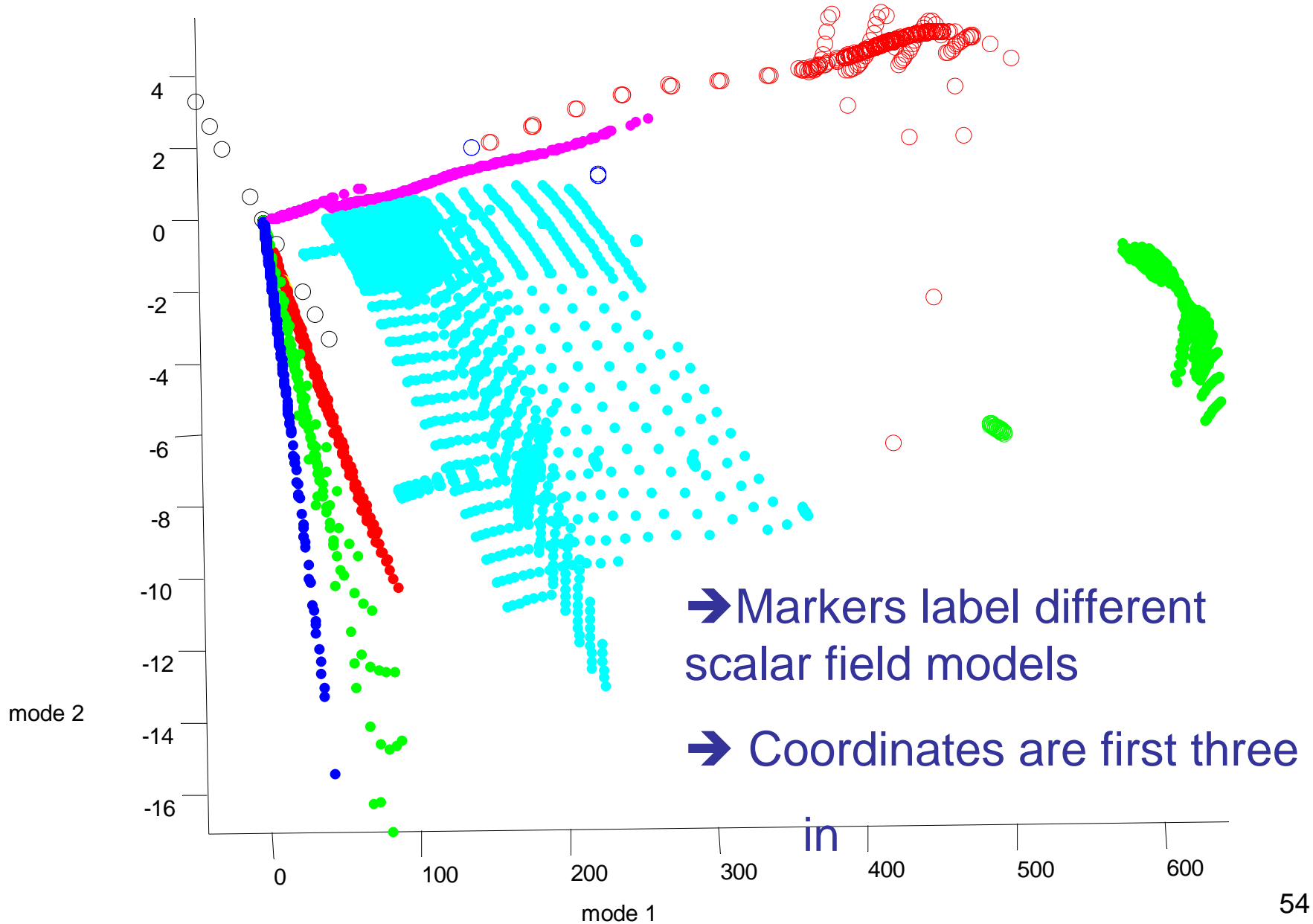
Implication: New experiments will have very significant discriminating power among actual scalar field models.
(See Augusta Abrahamse, Michael Barnard, Brandon Bozek & AA, to appear soon)



END

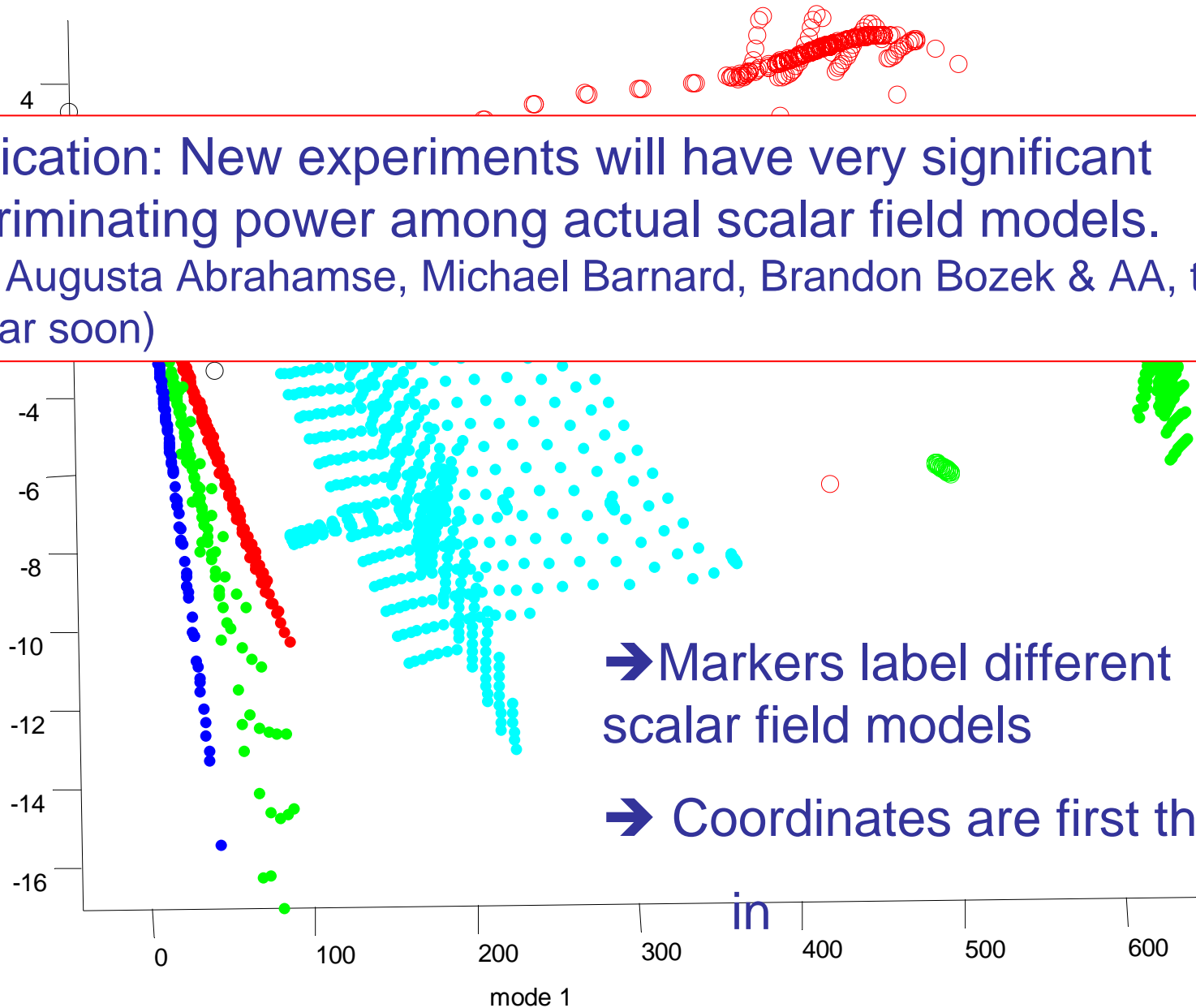
Extra material

sigMax = 4 sigMin = 0 OneModel = 0 OneVersionP1 = 0 OneRun = 0 EigenSR14 AllSolsV2₂

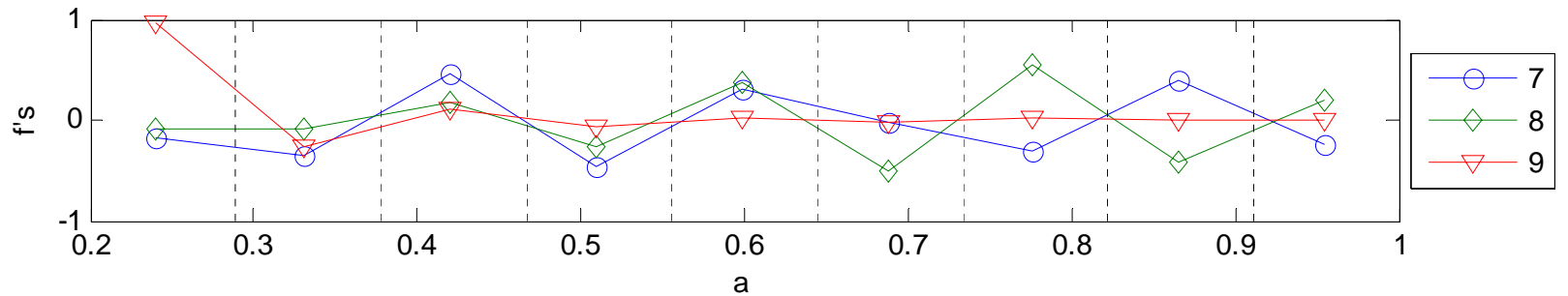
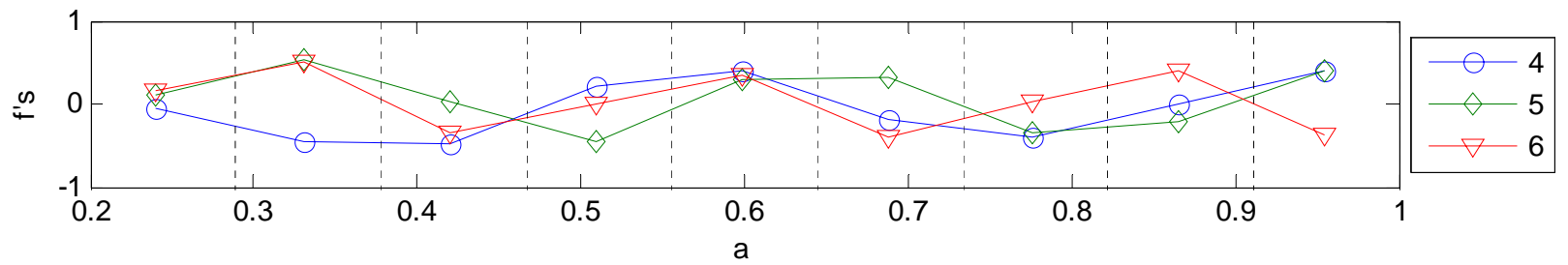
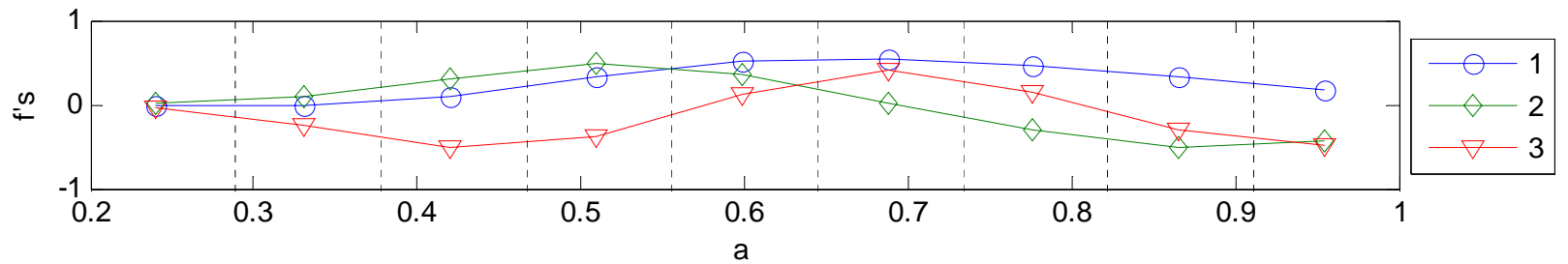
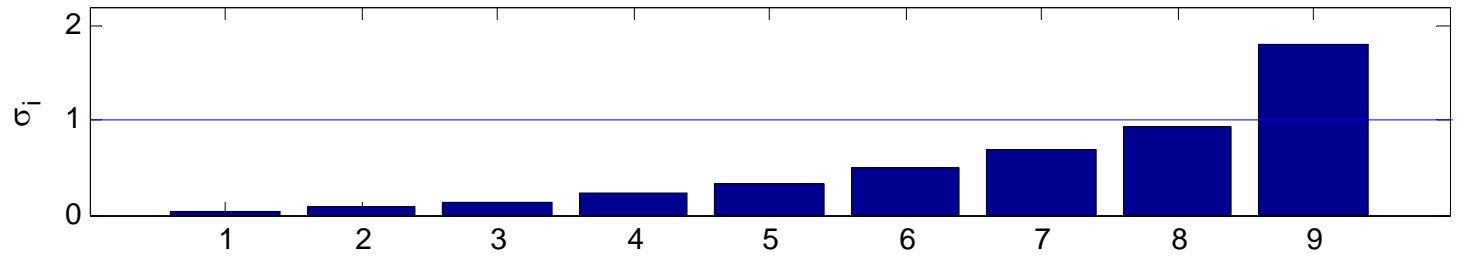




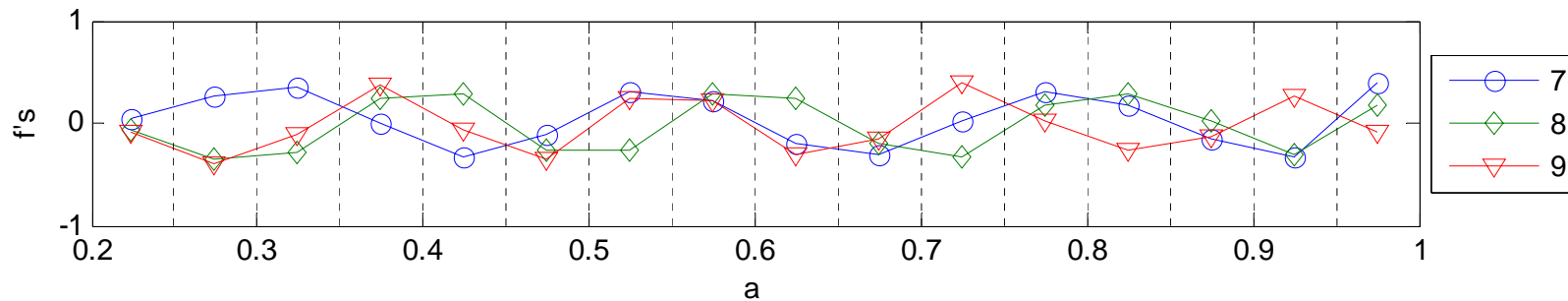
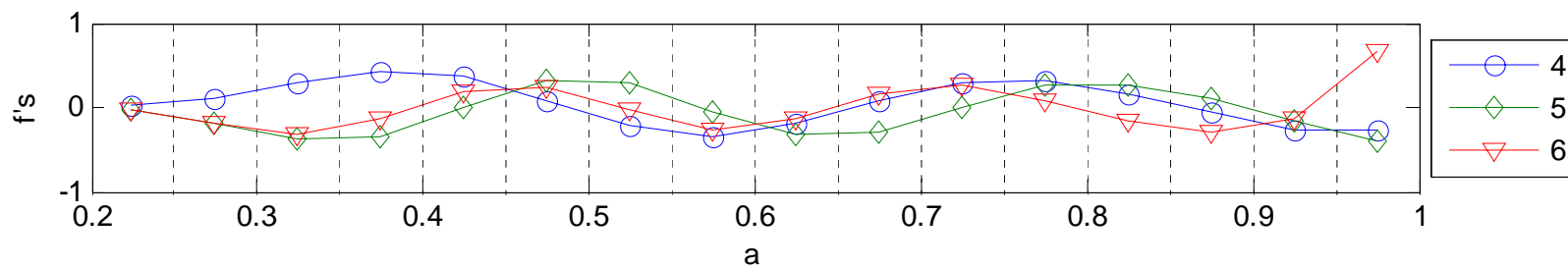
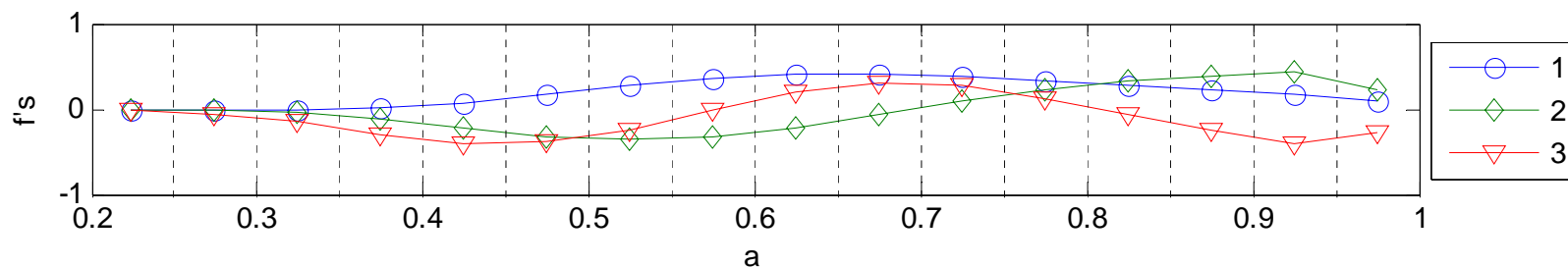
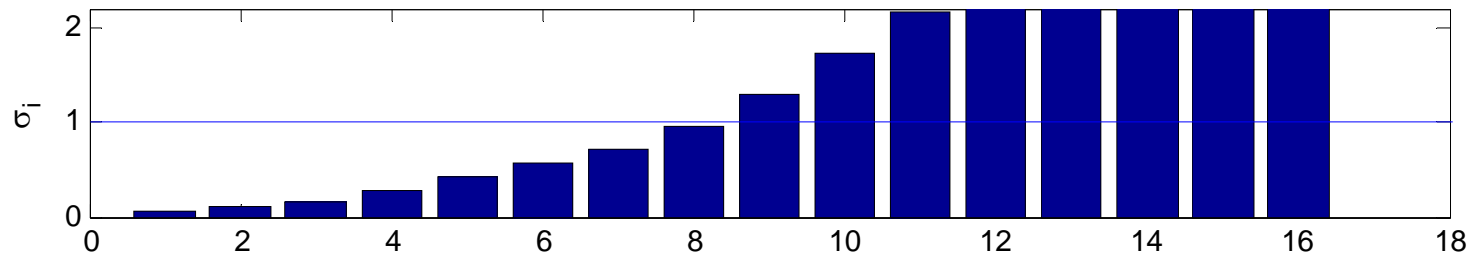
Implication: New experiments will have very significant discriminating power among actual scalar field models.
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Stage 4 Space WL Opt; lin-a $N_{\text{Grid}} = 9$, $z_{\text{max}} = 4$, Tag = 044301



Stage 4 Space WL Opt; lin-a $N_{\text{Grid}} = 16, z_{\text{max}} = 4, \text{Tag} = 054301$



chainlength = 20000 /home/bozek/Results/mcmc-lambda-VHeight-omegaM-omegak-M03-R1-AllStage2

gran = 32

covfactor = 1

steps = 52842

initial pars = 0.32199 0.394 0.1461 0.00061067 0.0040025 0.0061012 0.00092421 -0.00065905

1

