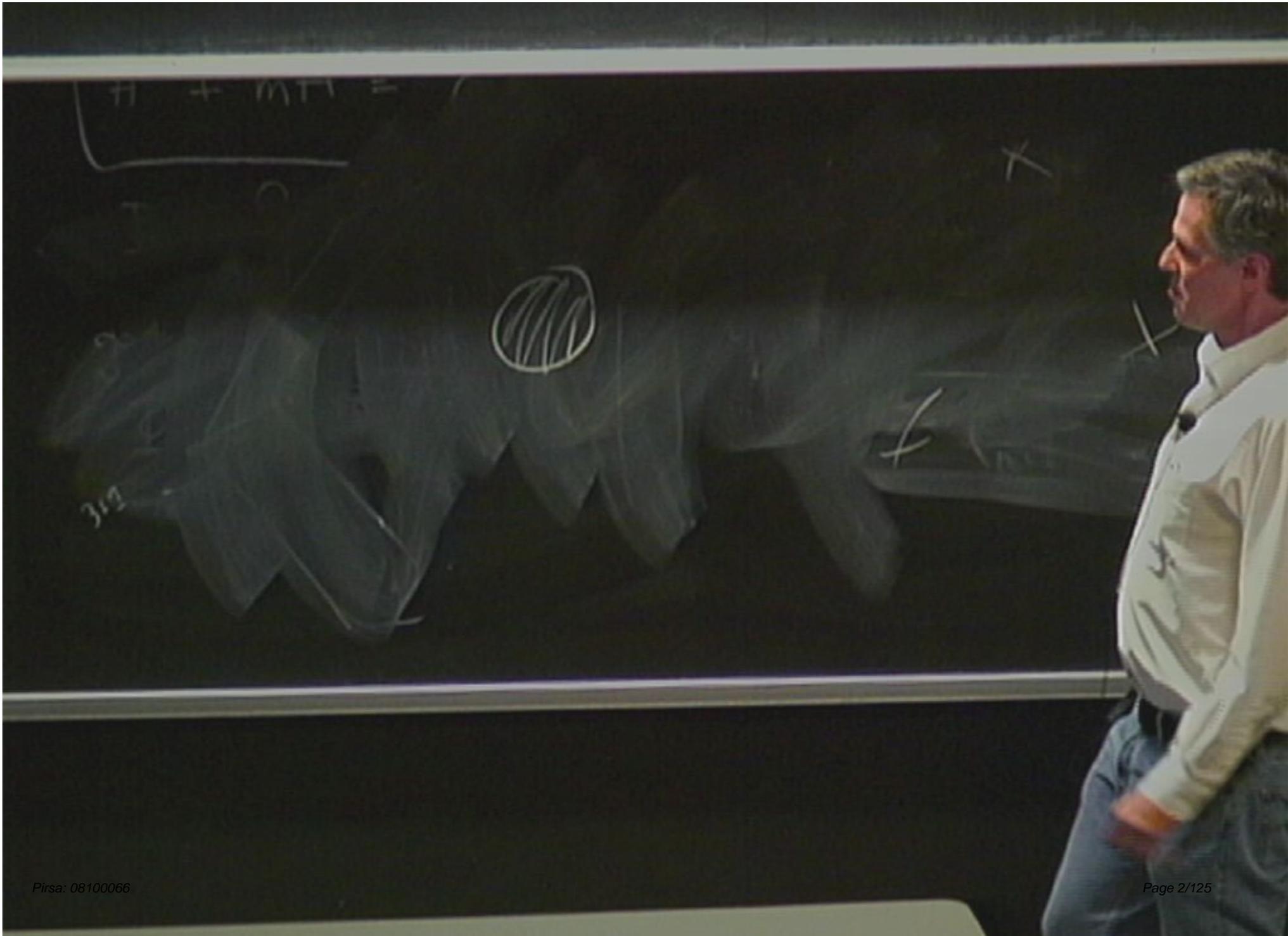


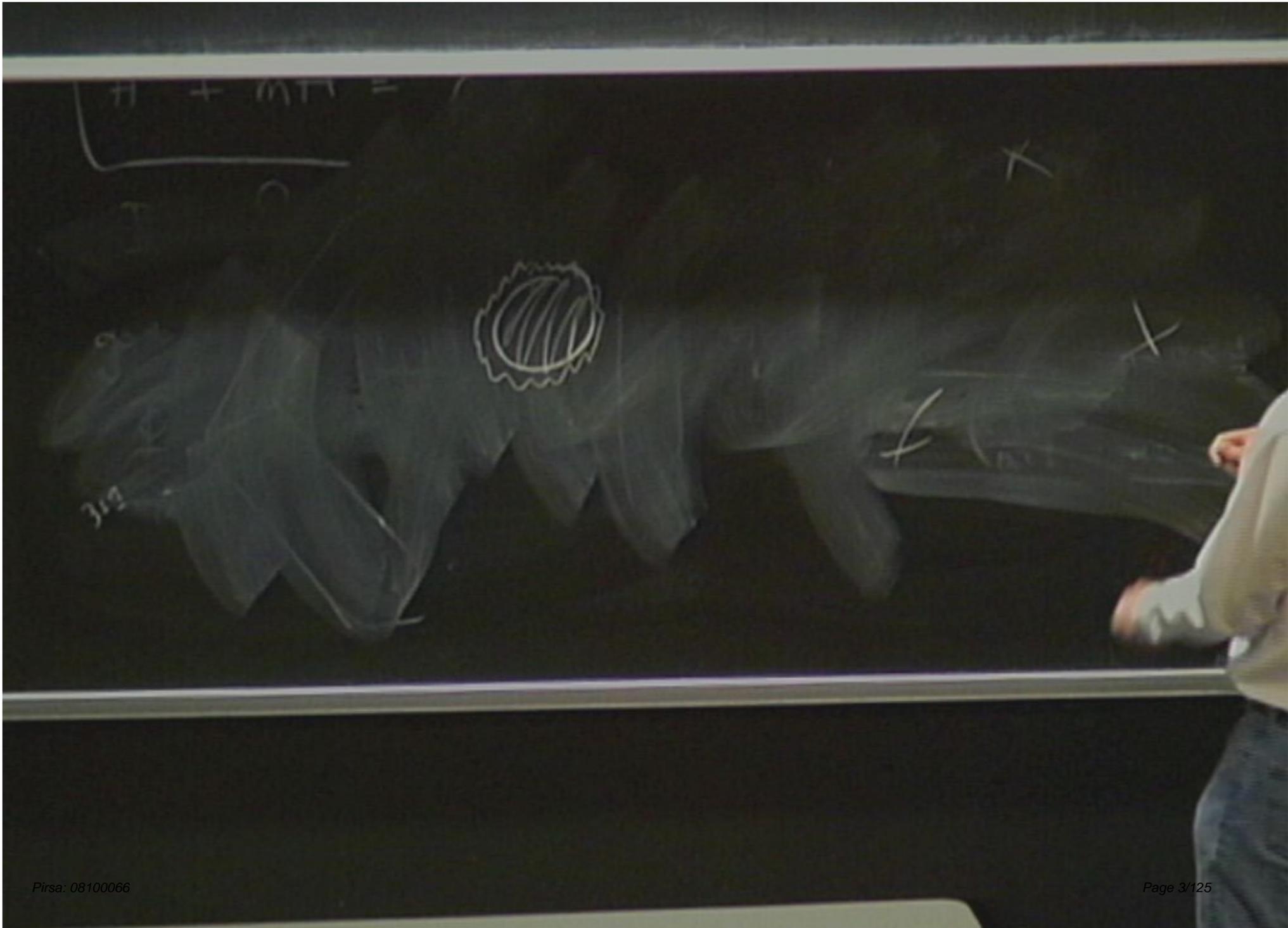
Title: What will future dark energy experiments tell us?

Date: Oct 23, 2008 11:30 AM

URL: <http://pirsa.org/08100066>

Abstract:





The Dark Energy Task Force (DETF)

- Created specific simulated data sets (Stage 2, Stage 3, Stage 4)
- Assessed their impact on our knowledge of dark energy as modeled with the w_0 - w_a parameters

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

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New work,
relevant to setting a
concrete threshold
for Stage 4

The Dark Energy Task Force

- Created specific simulations
- 4)
- Assessed their impact
- modeled with the w_0

NB: To make concrete comparisons this work ignores various possible improvements to the DETF data models.

(see for example J Newman, H Zhan *et al* & Schneider *et al*)

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DETF

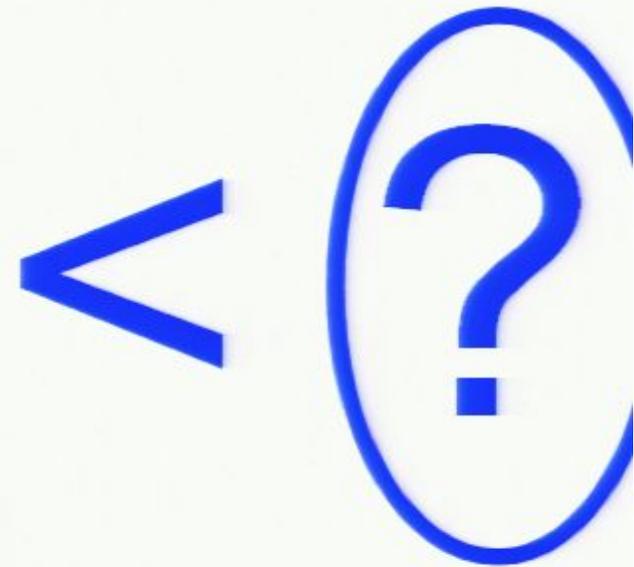
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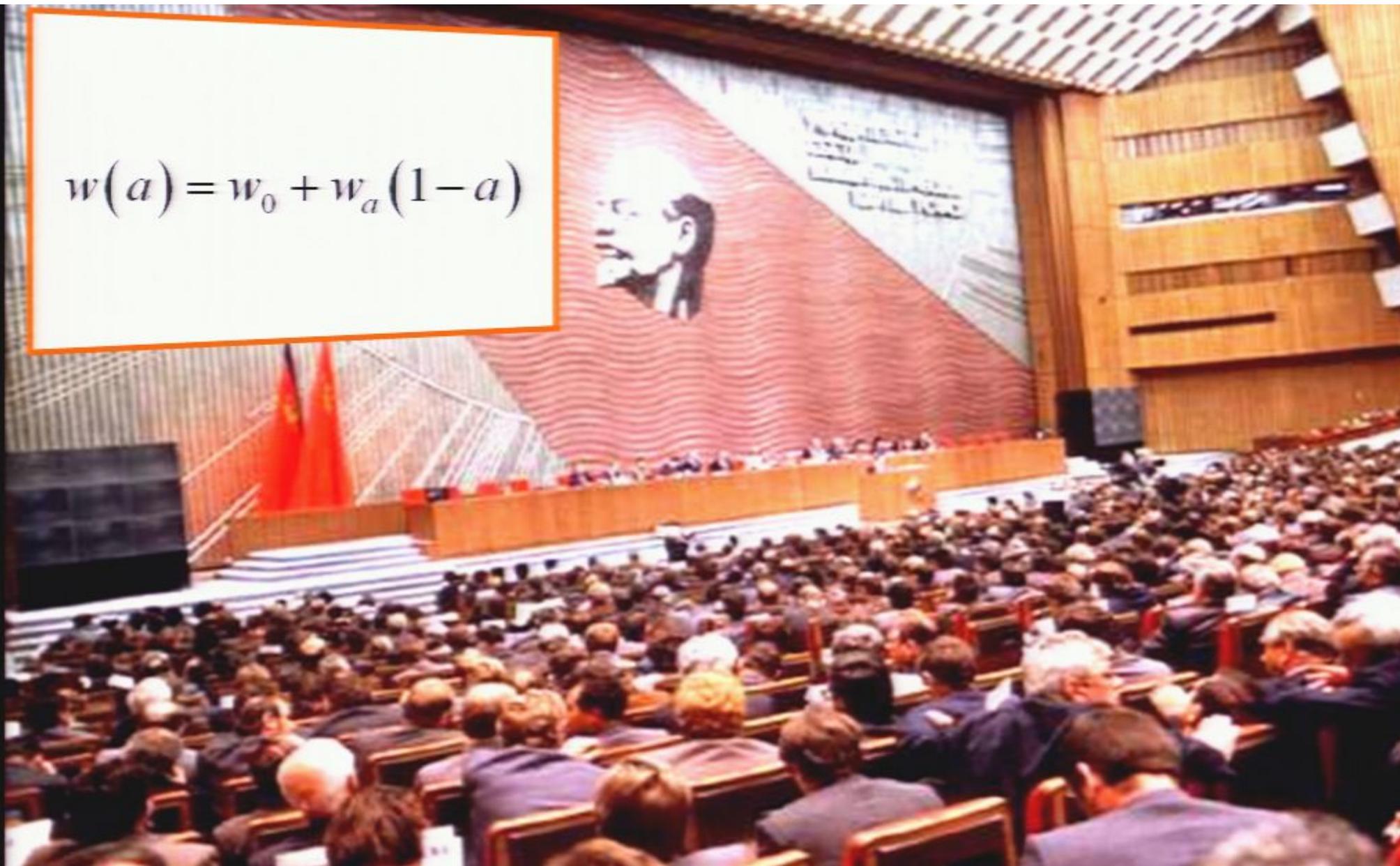
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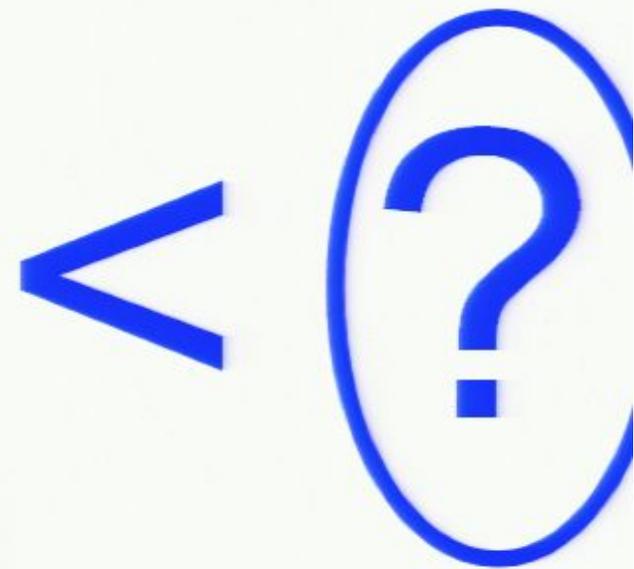
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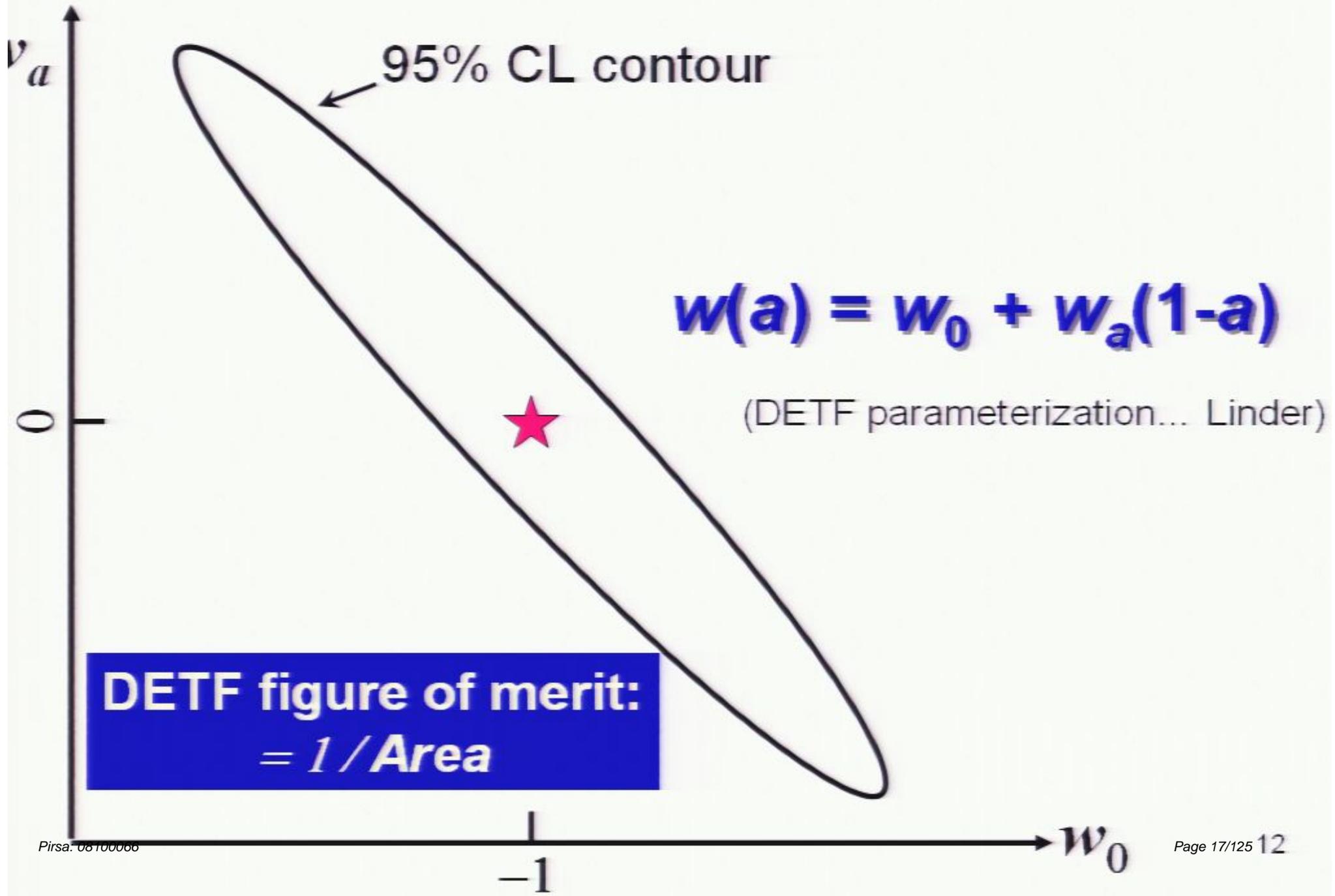
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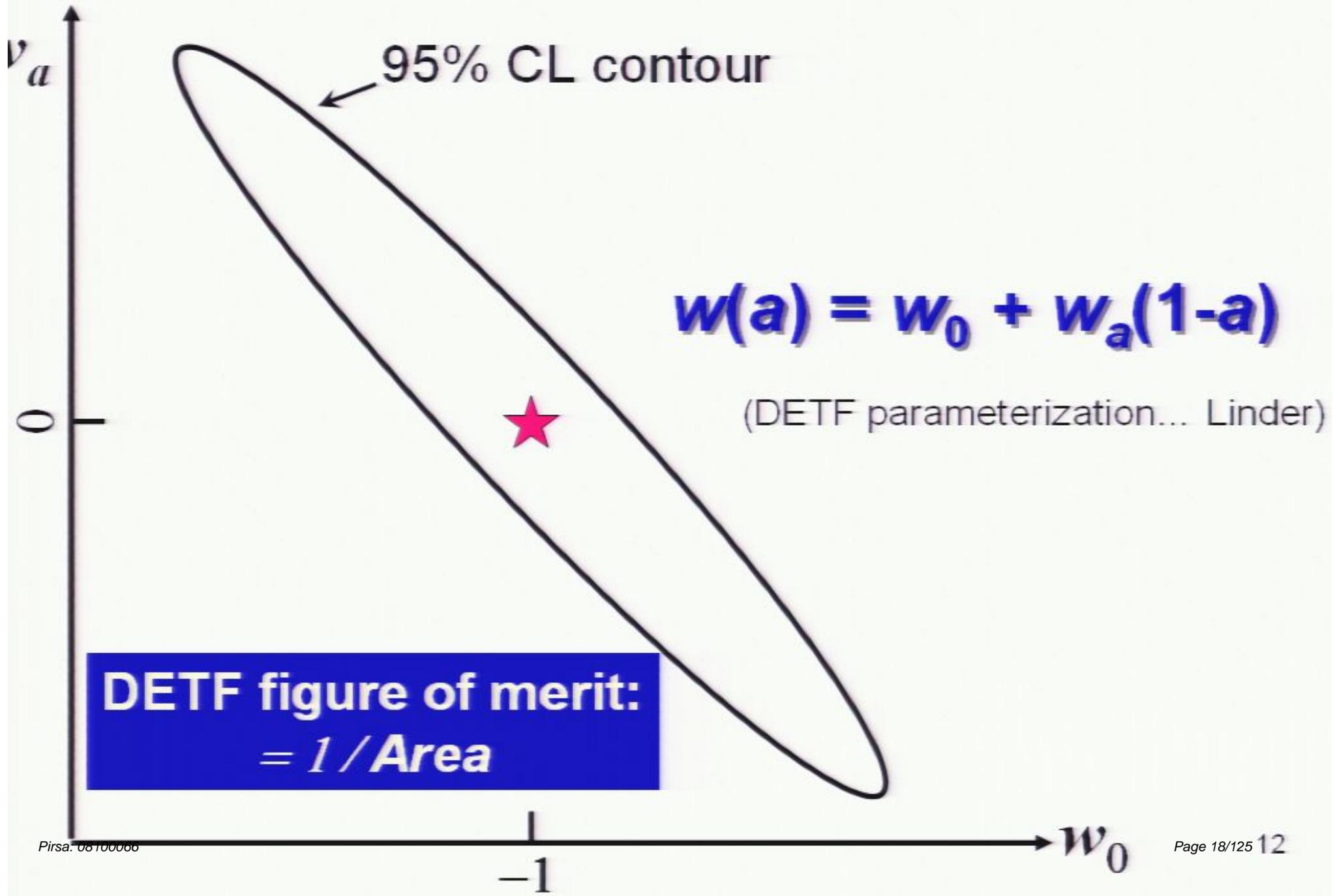




...but these parameters come out very well from this analysis

Review





The DETF stages (data models constructed for each one)

Stage 2: Underway

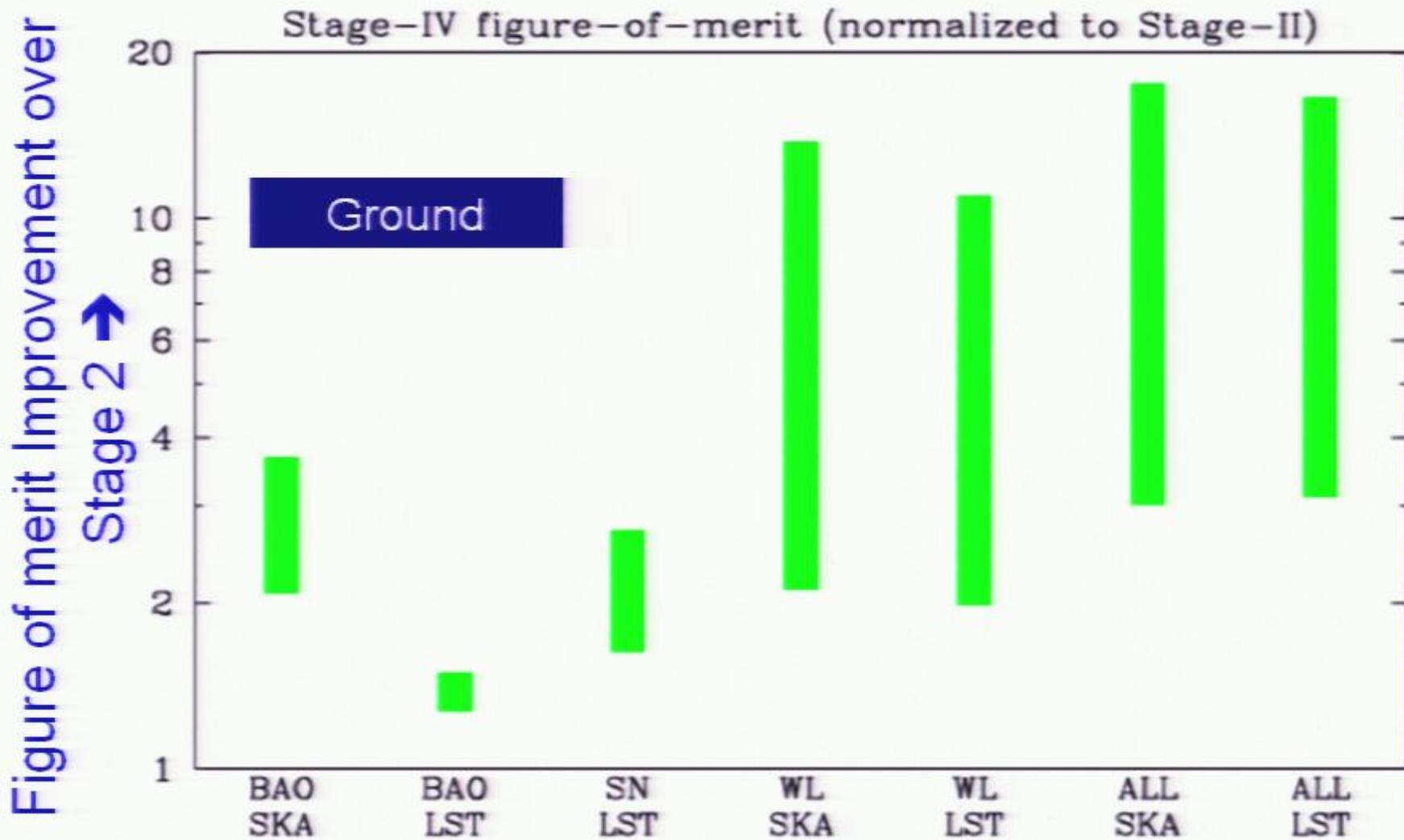
Stage 3: Medium size/term projects

Stage 4: Large longer term projects (ie JDEM, LST)

DETF modeled

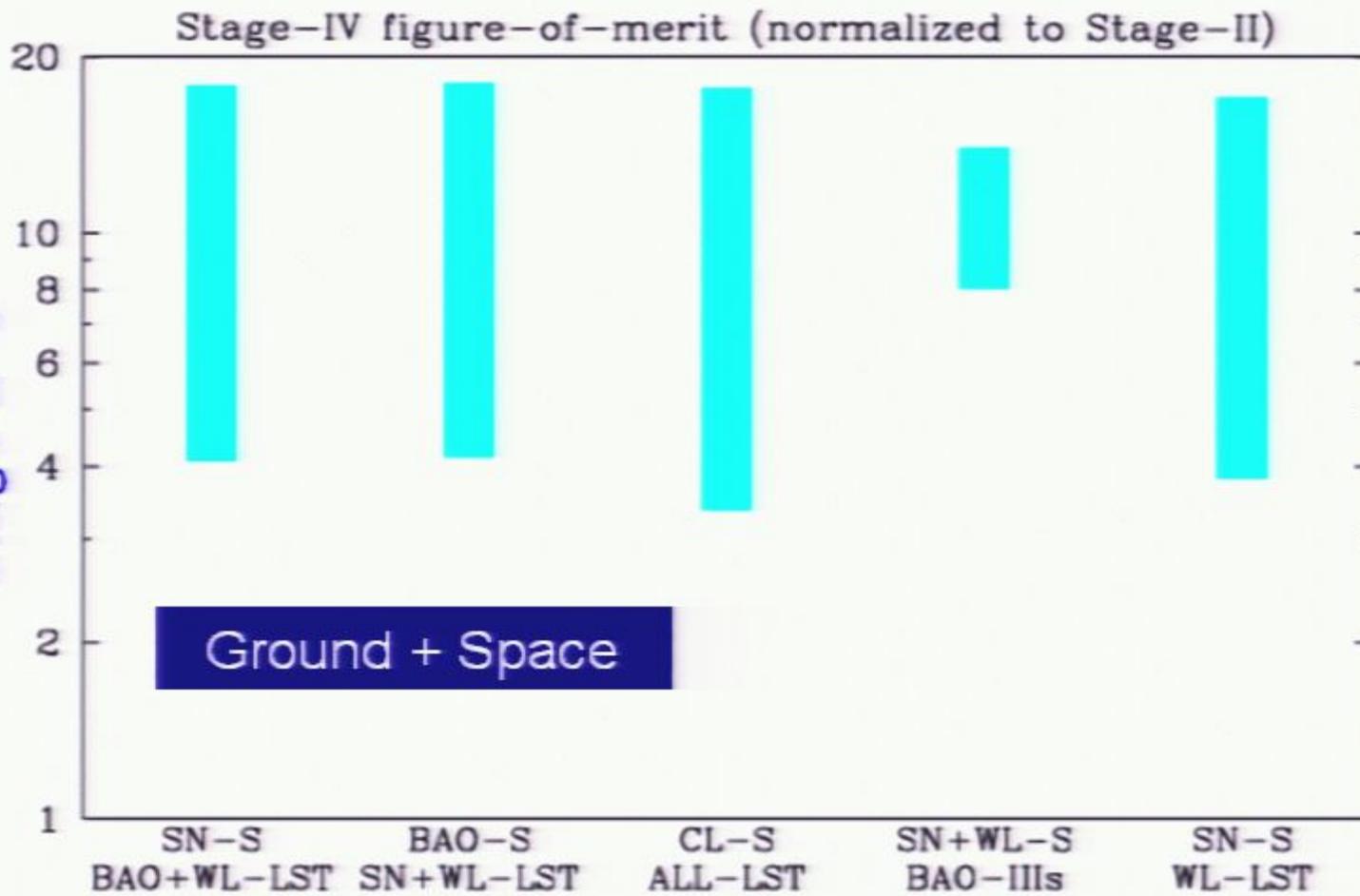
- SN
- Weak Lensing
- Baryon Oscillation
- Cluster data

DETF Projections

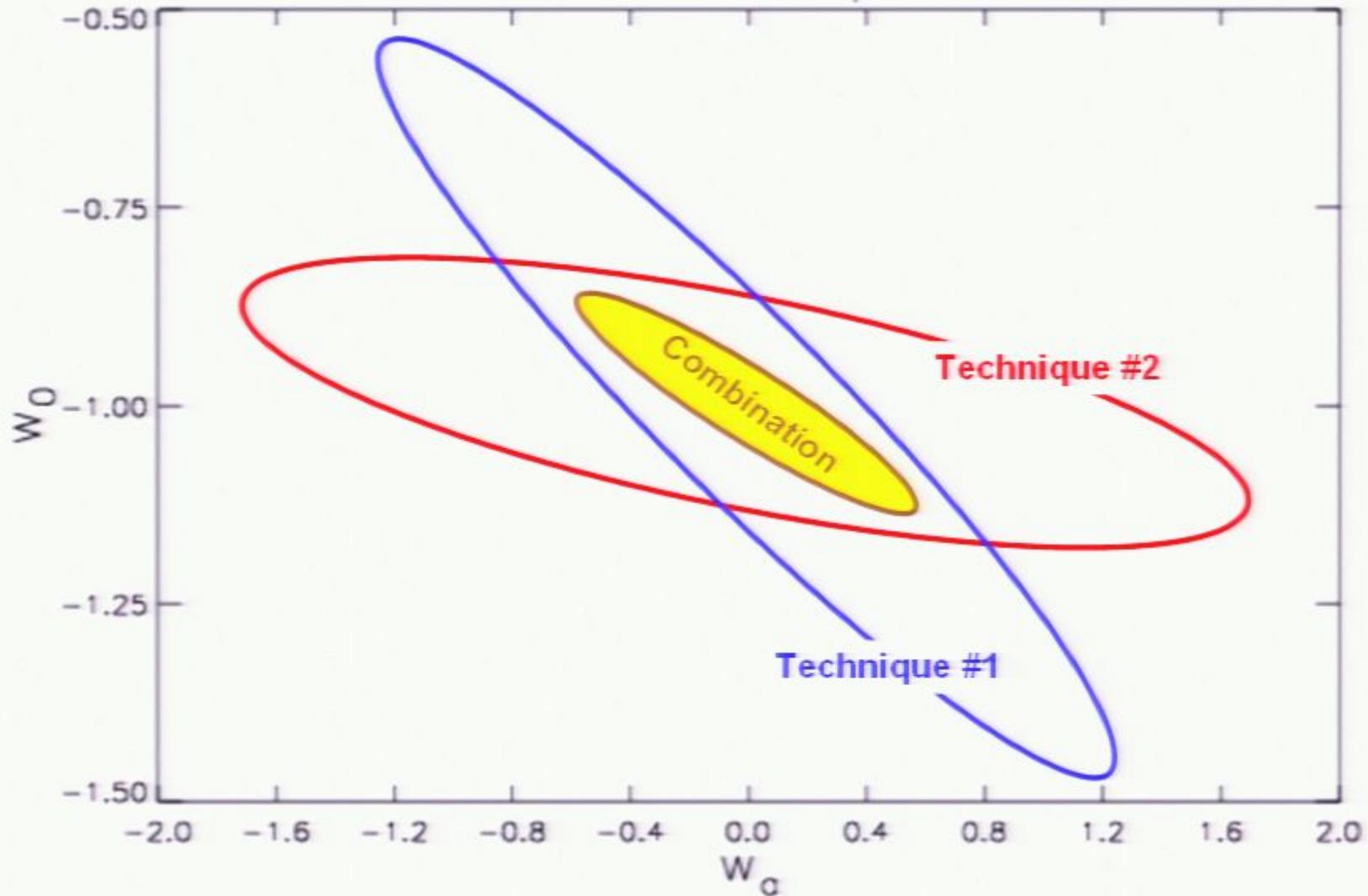


DETF Projections

Figure of merit Improvement over Stage 2 →



A technical point: The role of correlations



From the DETF Executive Summary

One of our main findings is that **no single technique** can answer the outstanding questions about dark energy: combinations of **at least two** of these techniques must be used to fully realize the promise of future observations.

Already there are proposals for **major, long-term (Stage IV) projects** incorporating these techniques that have the promise of increasing our figure of merit by a factor of ten beyond the level it will reach with the conclusion of current experiments. **What is urgently needed is a commitment to fund a program comprised of a selection of these projects.** The selection should be made on the basis of critical evaluations of their costs, benefits, and risks.

Followup questions:

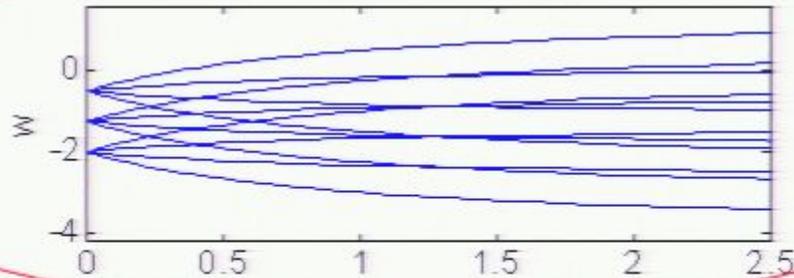
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How good is the $w(a)$ ansatz?

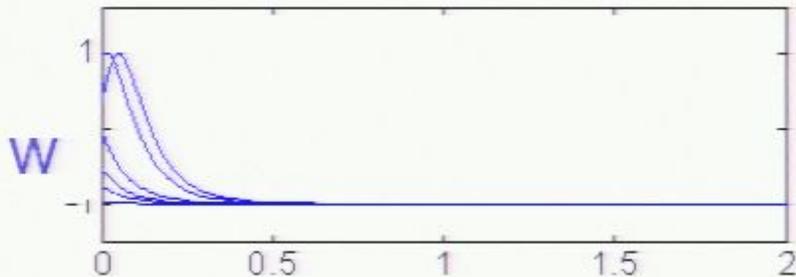
Sample $w(z)$ curves in w_0 - w_a space



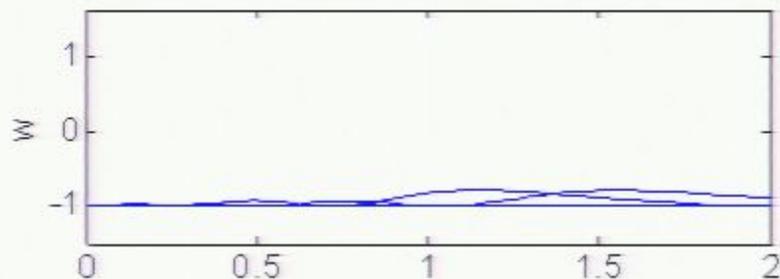
$$w(a) = w_0 + w_a (1 - a)$$

w_0 - w_a can only do these

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



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

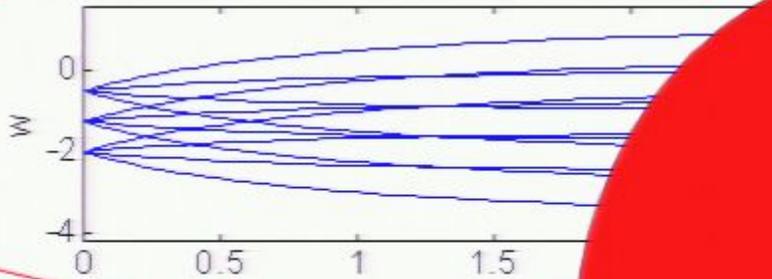


DE models can do this
(and much more)

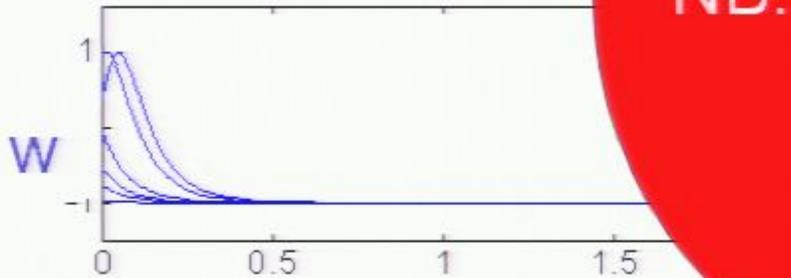
Z

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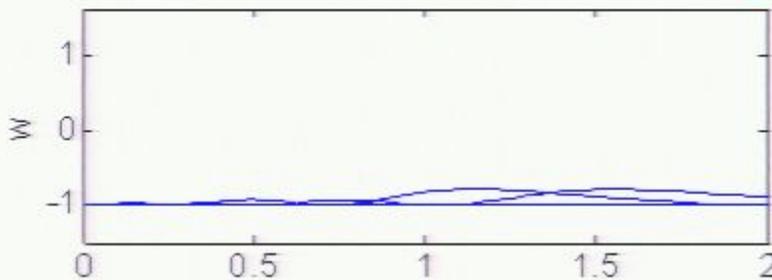
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Sample $w(z)$ curves for the PN



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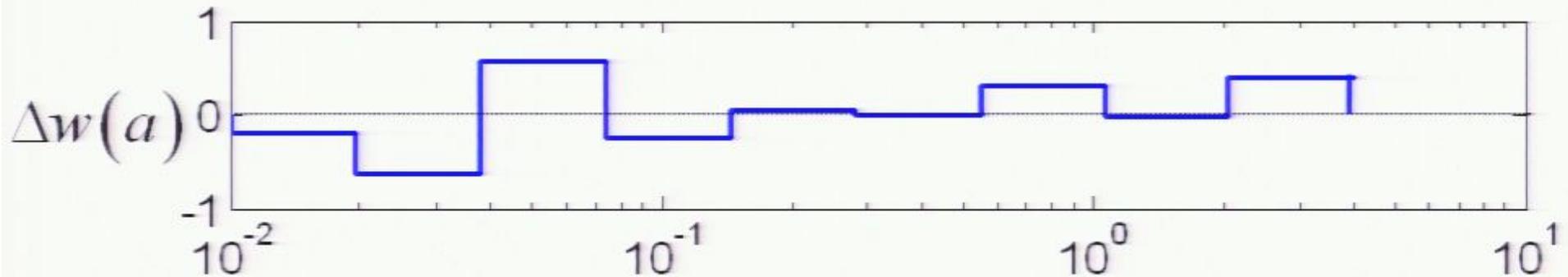
NB: Better than

$$w(a) = w_0$$

& flat

DE models can do this
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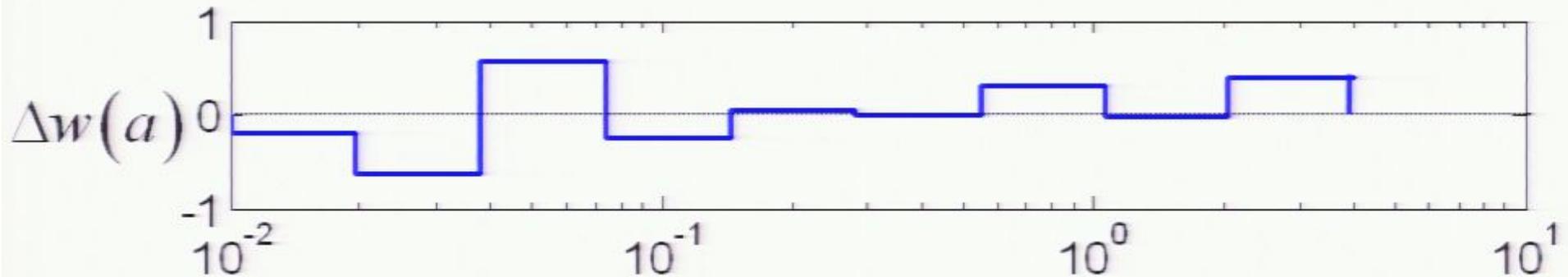
Try (9)D stepwise constant $w(a)$



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

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

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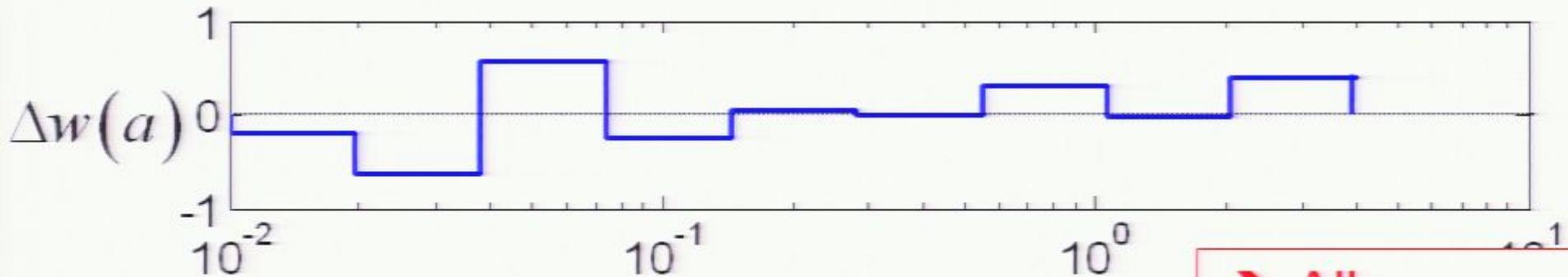
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Used by

Huterer & Turner;
Huterer & Starkman;
Knox et al;
Crittenden & Pogosian
Linder; Reiss et al;
Krauss et al
de Putter & Linder;
Sullivan et al

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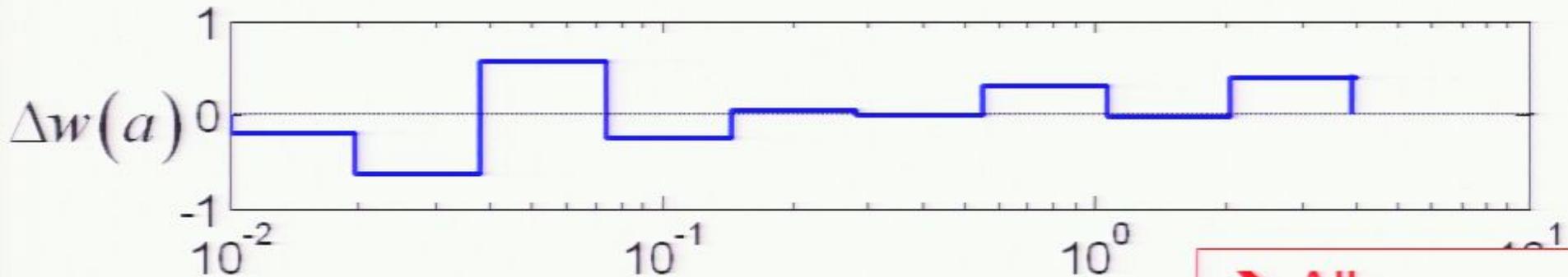


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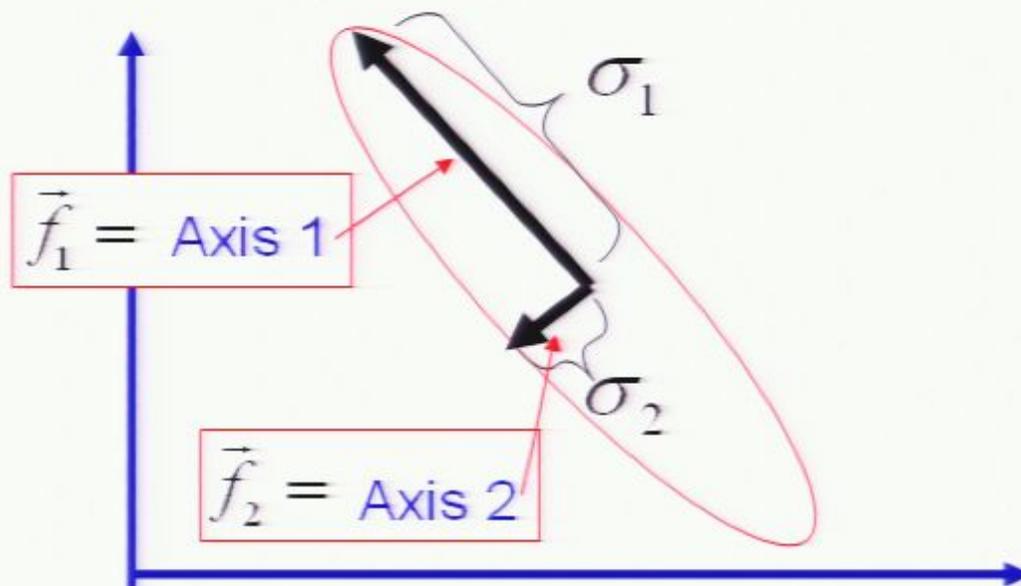
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“Convergence”

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:

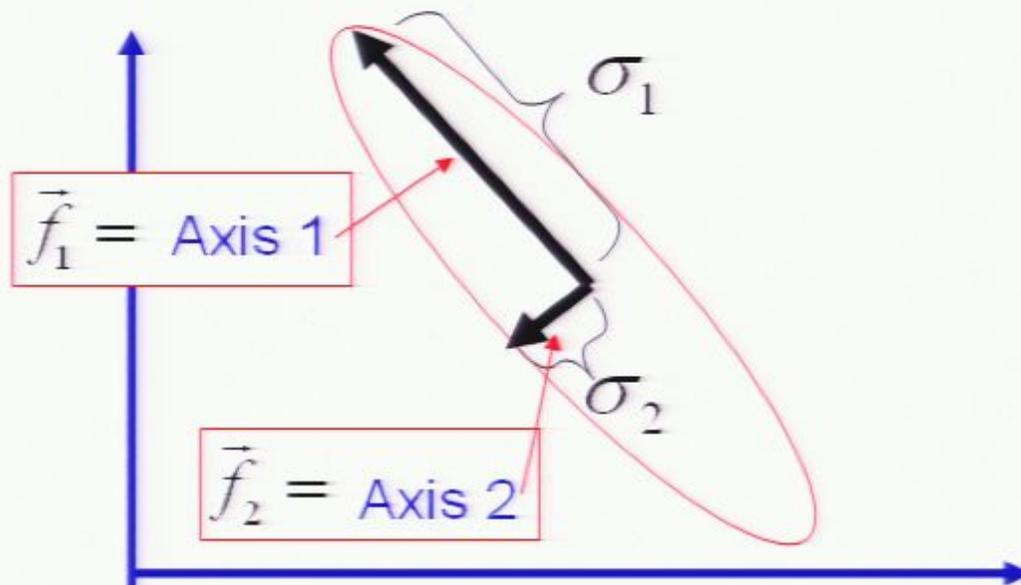


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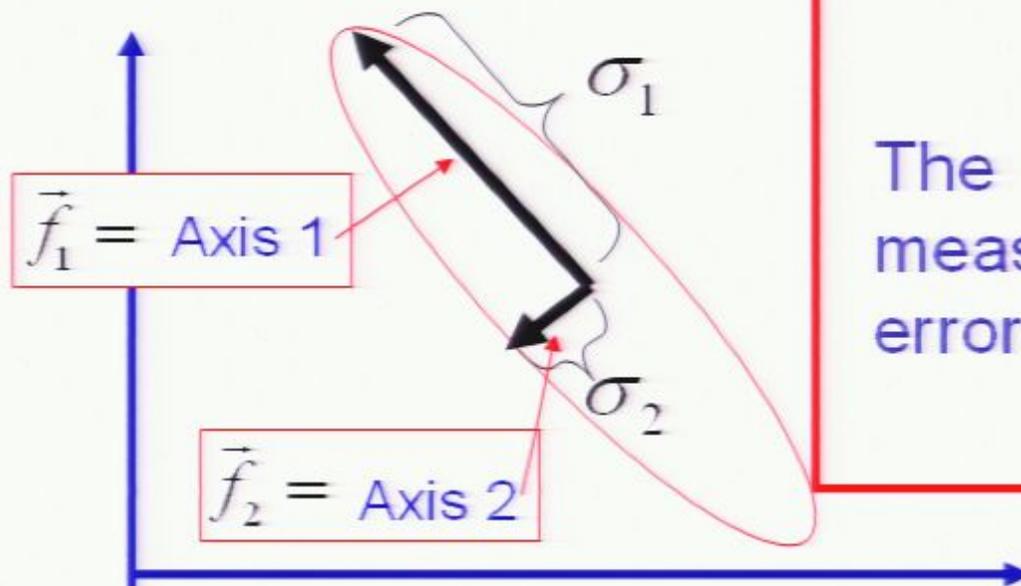
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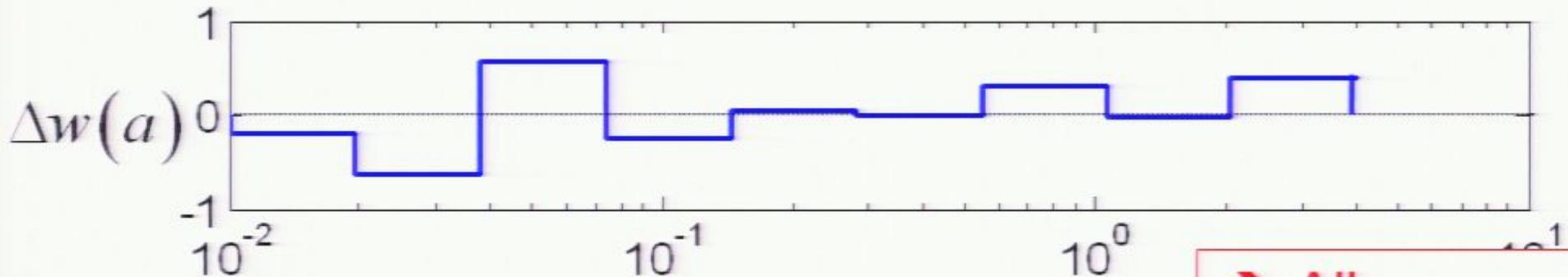


NB: in general the \vec{f}_i s form a complete basis:

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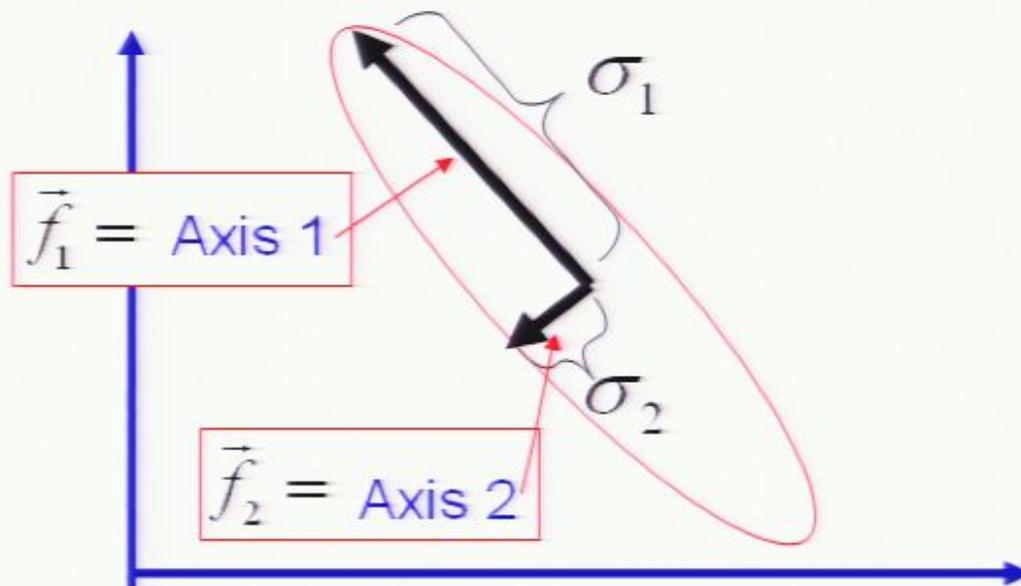
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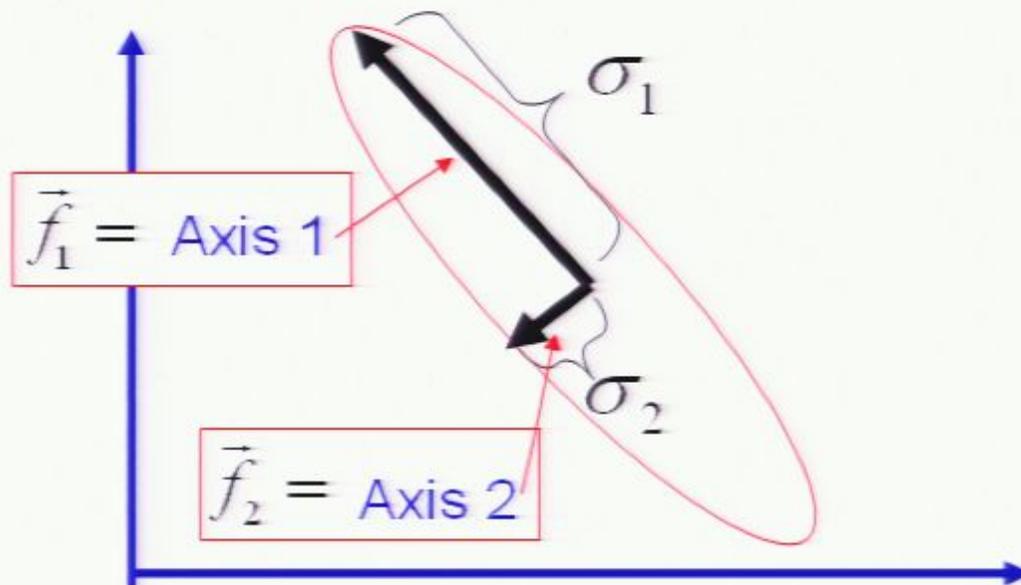


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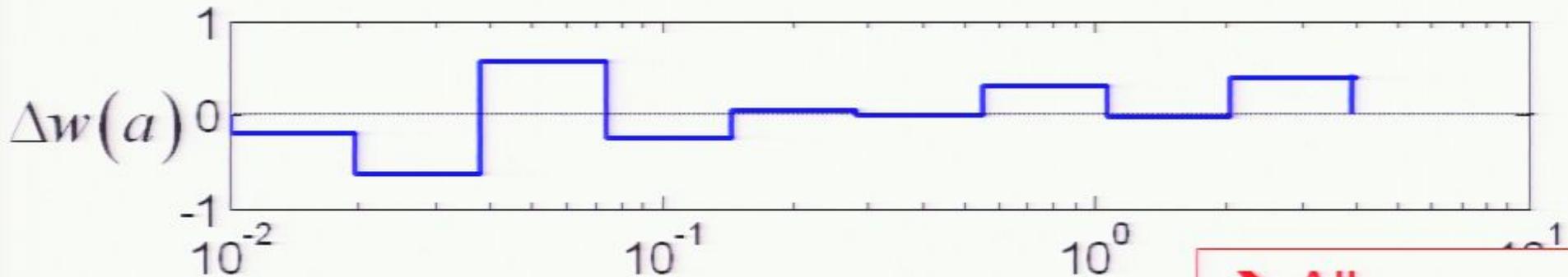
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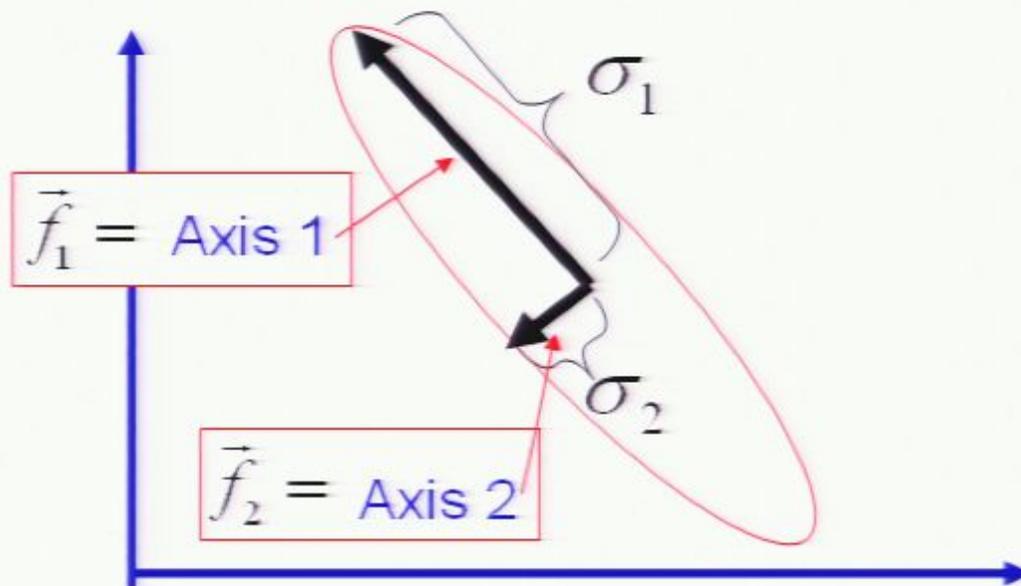
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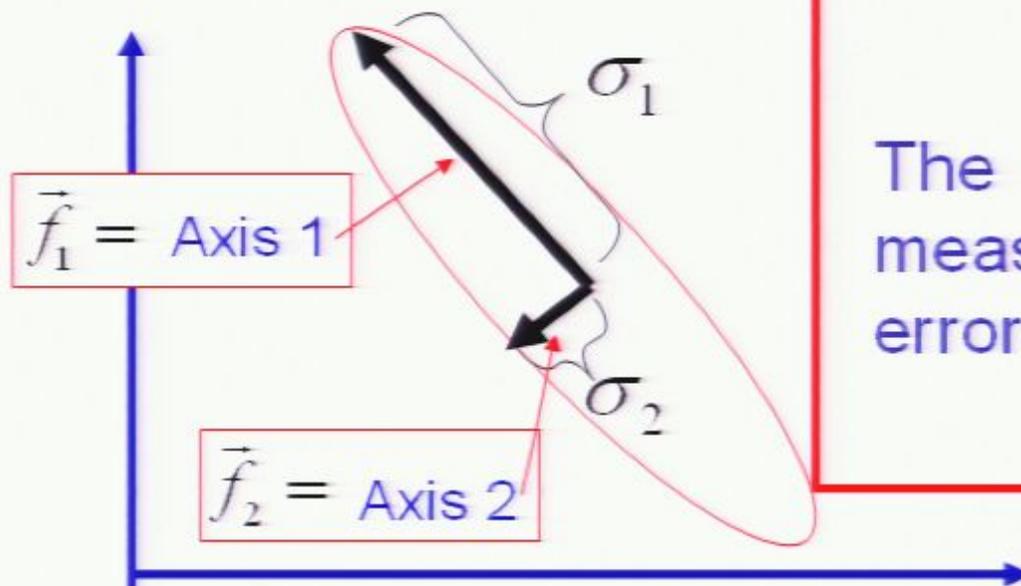
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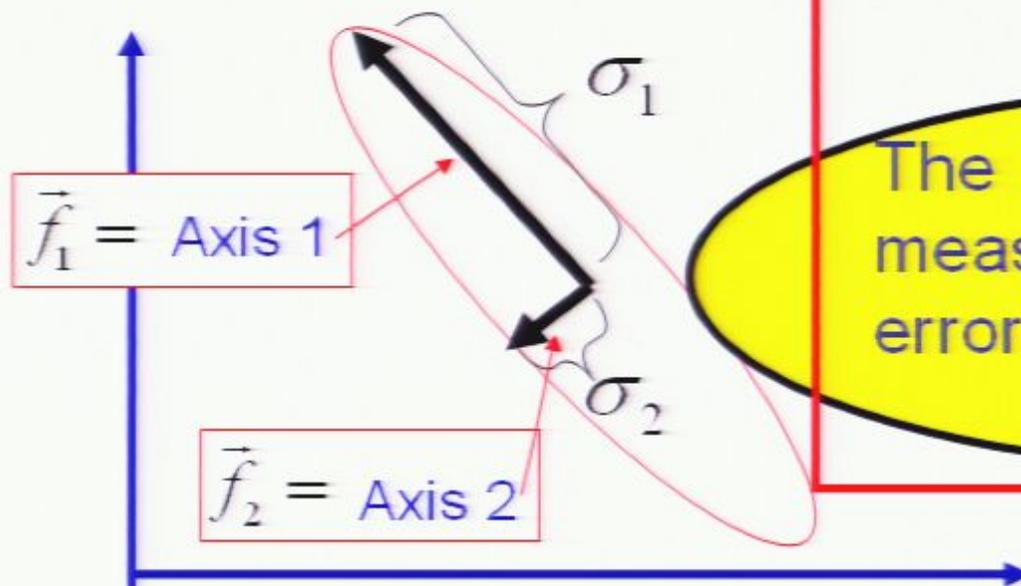
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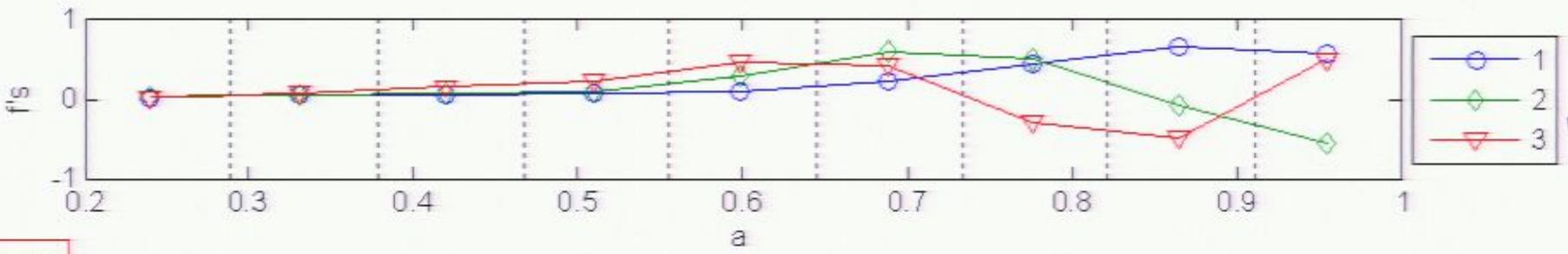
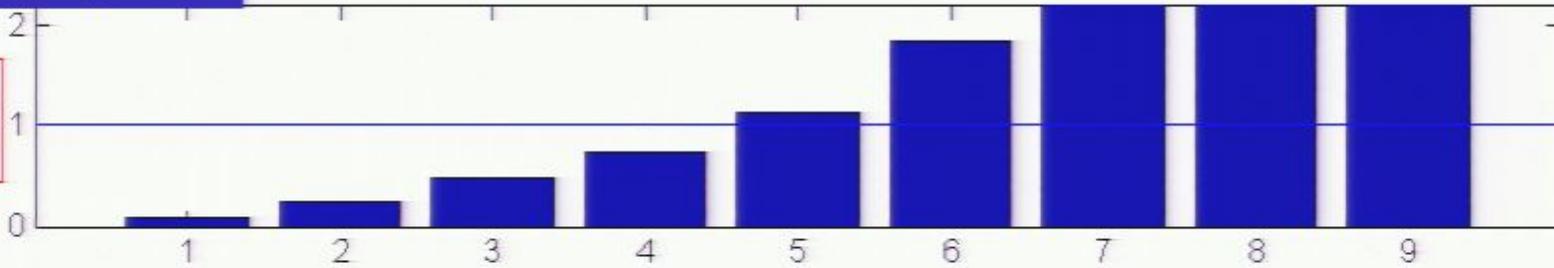
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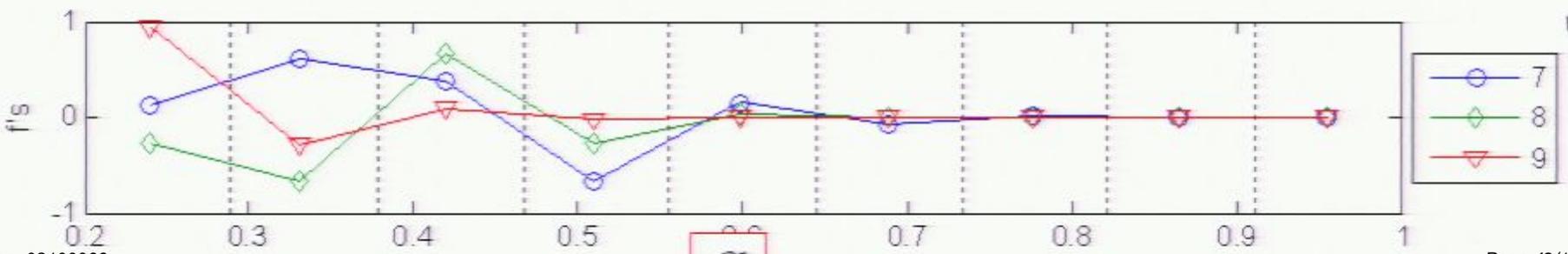
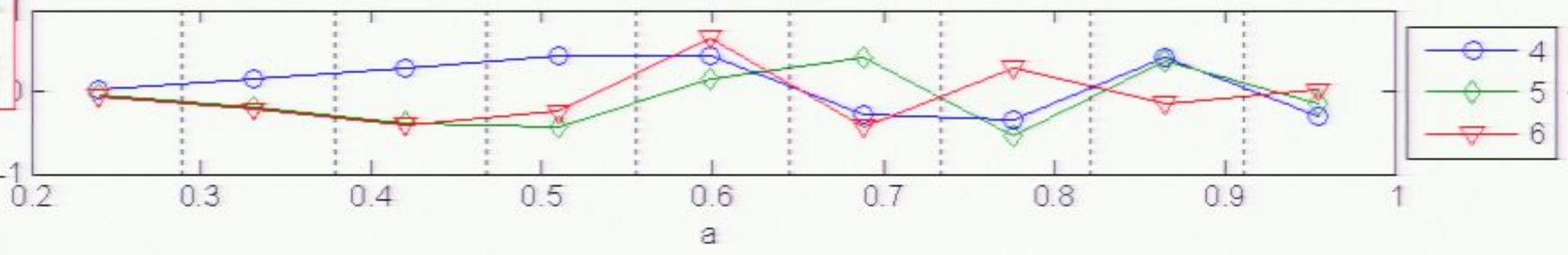
Characterizing 9D ellipses by principle axes and corresponding errors

DETF stage 2

σ_i



f_i



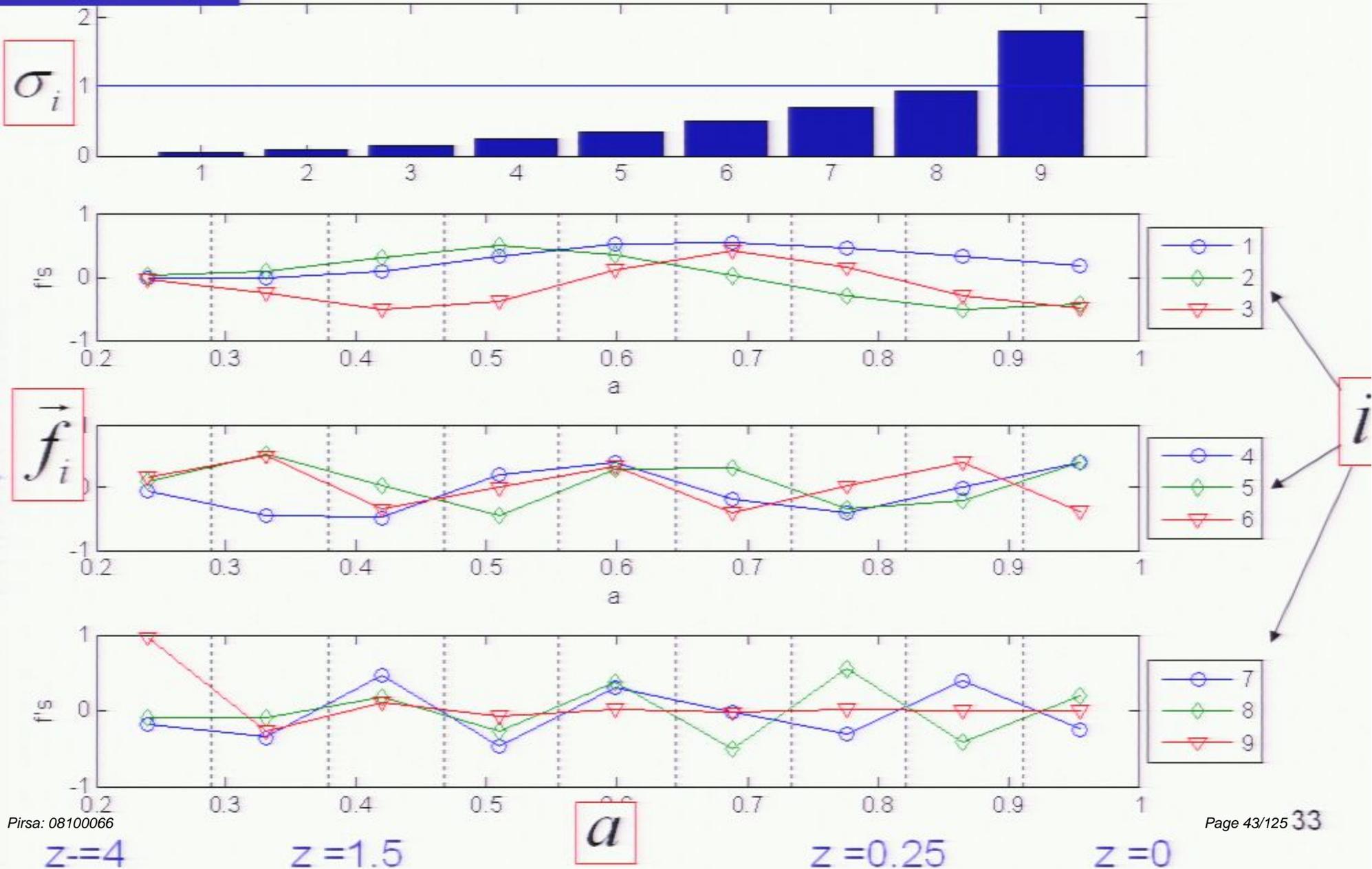
i

Principle axes

$z=-4$ $z=1.5$ a $z=0.25$ $z=0$

Characterizing 9D ellipses by principle axes and corresponding errors

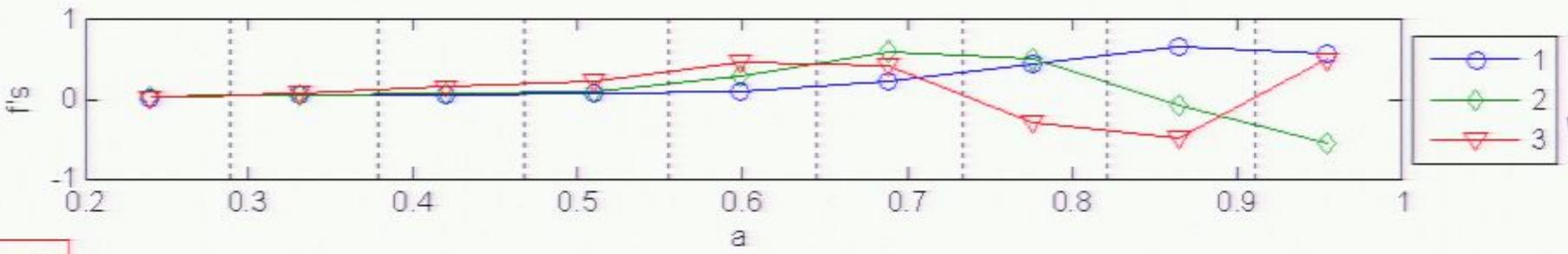
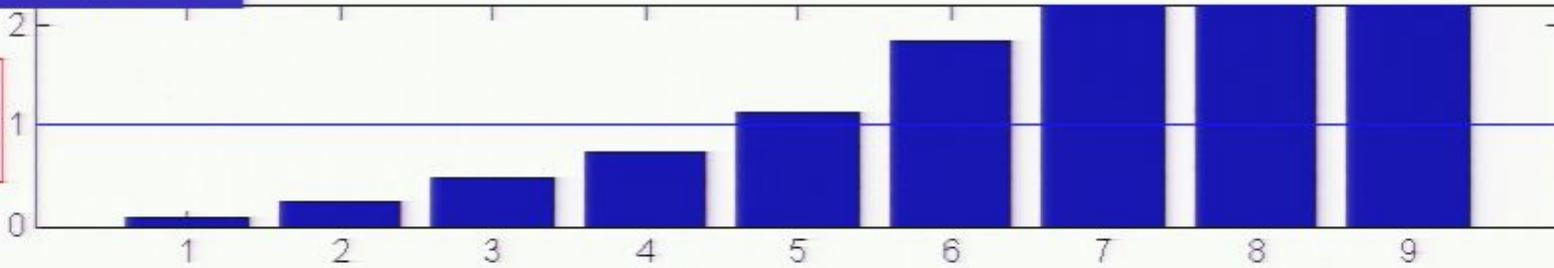
Stage 4 Opt



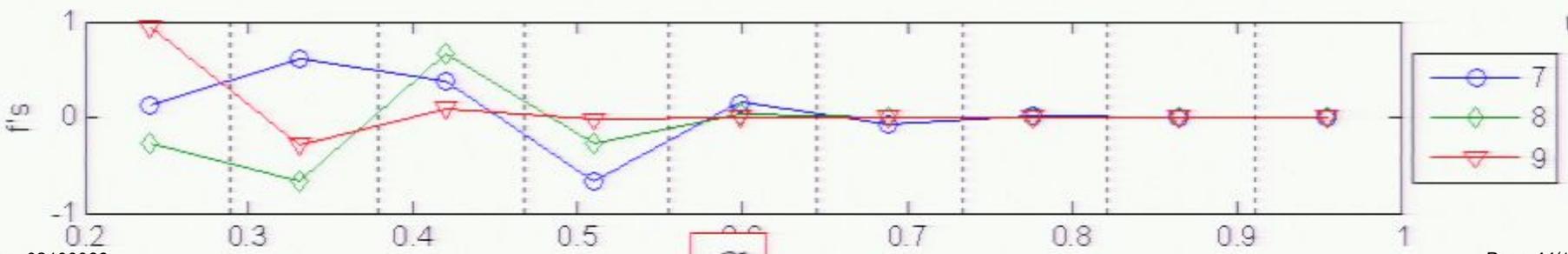
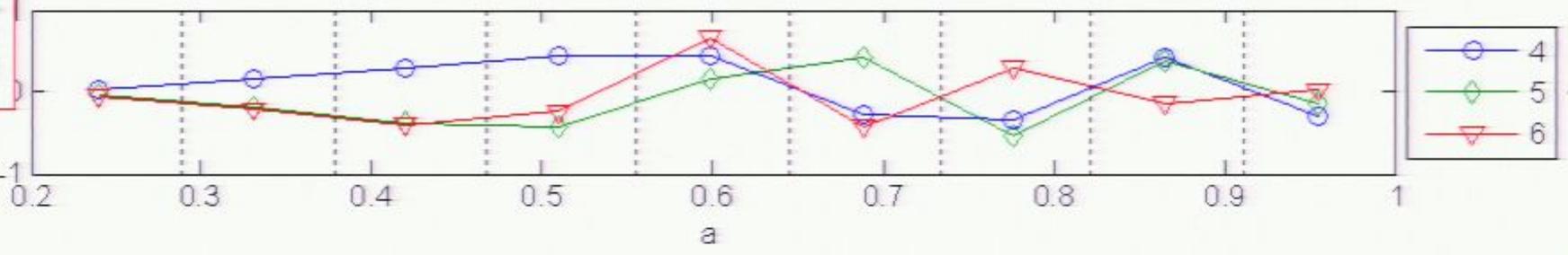
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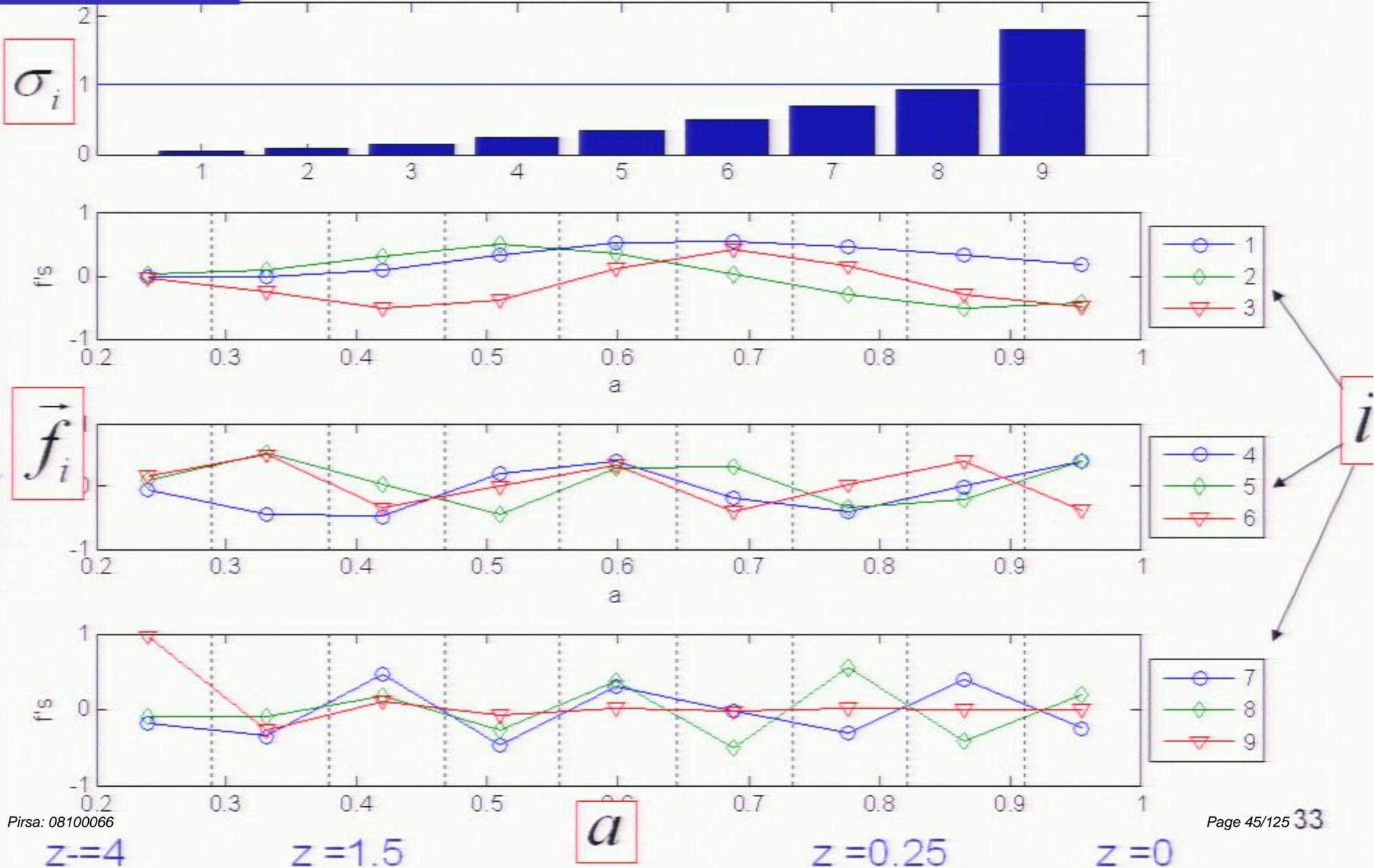
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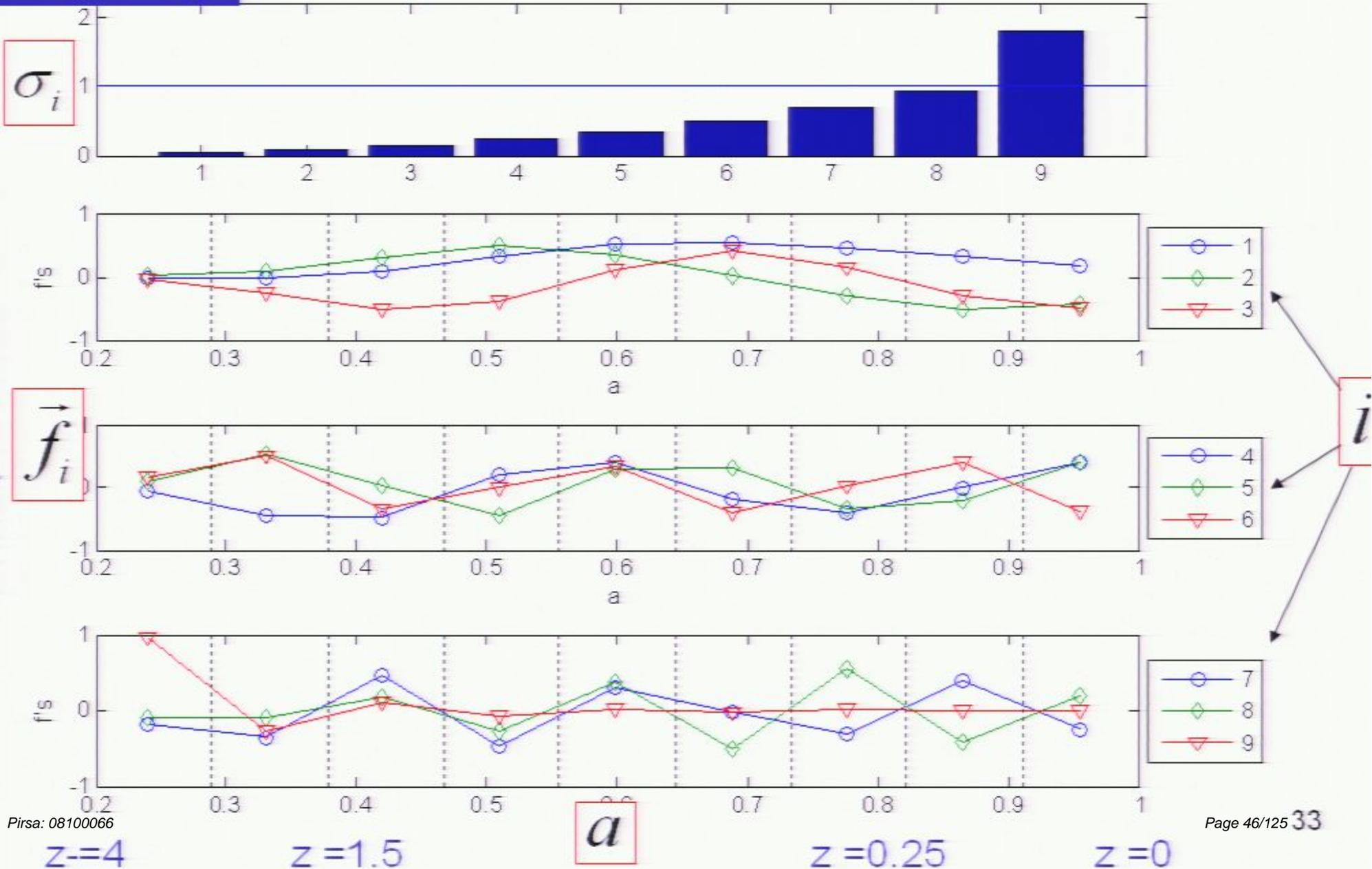
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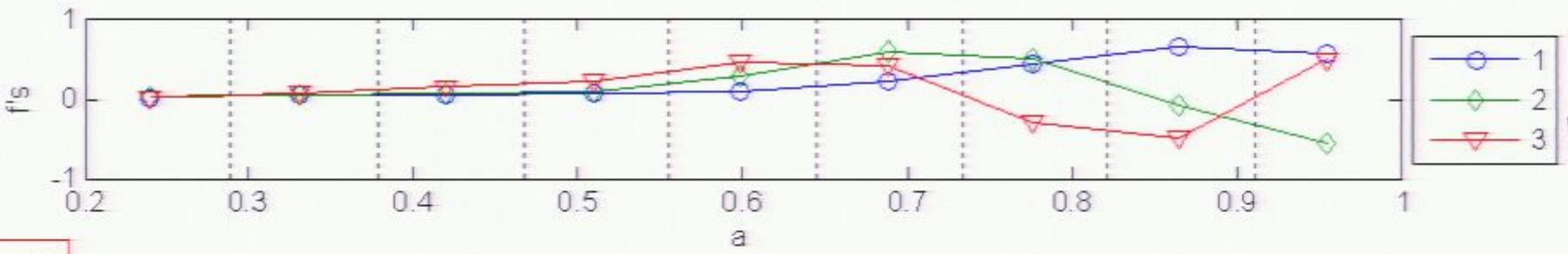
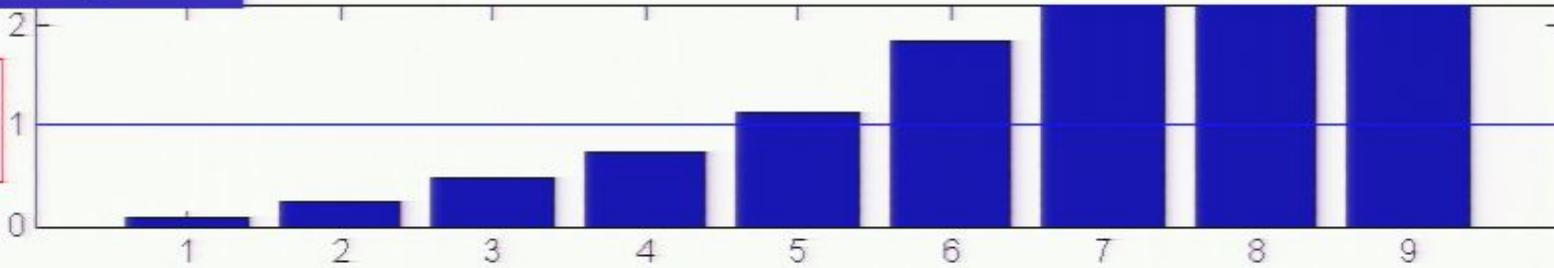
VL Stage 4 Opt



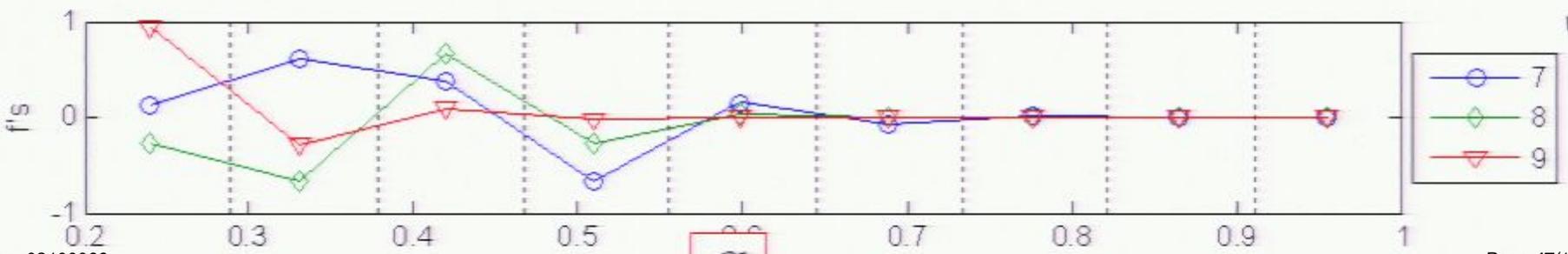
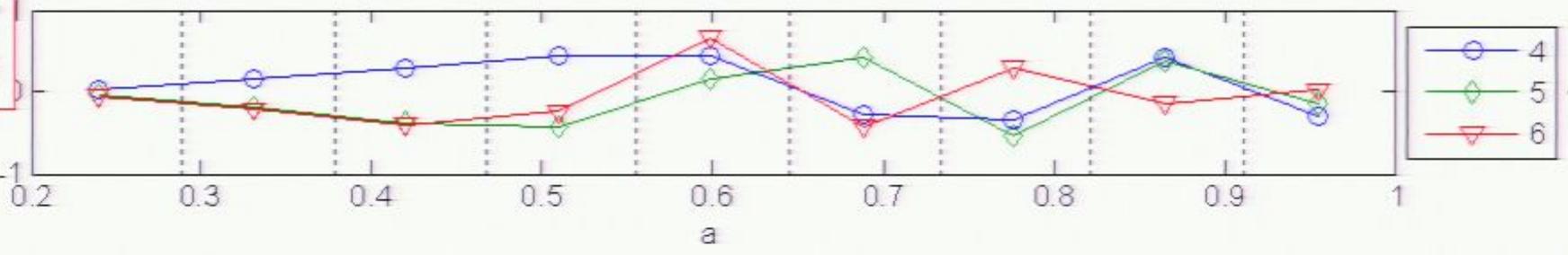
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DETF stage 2

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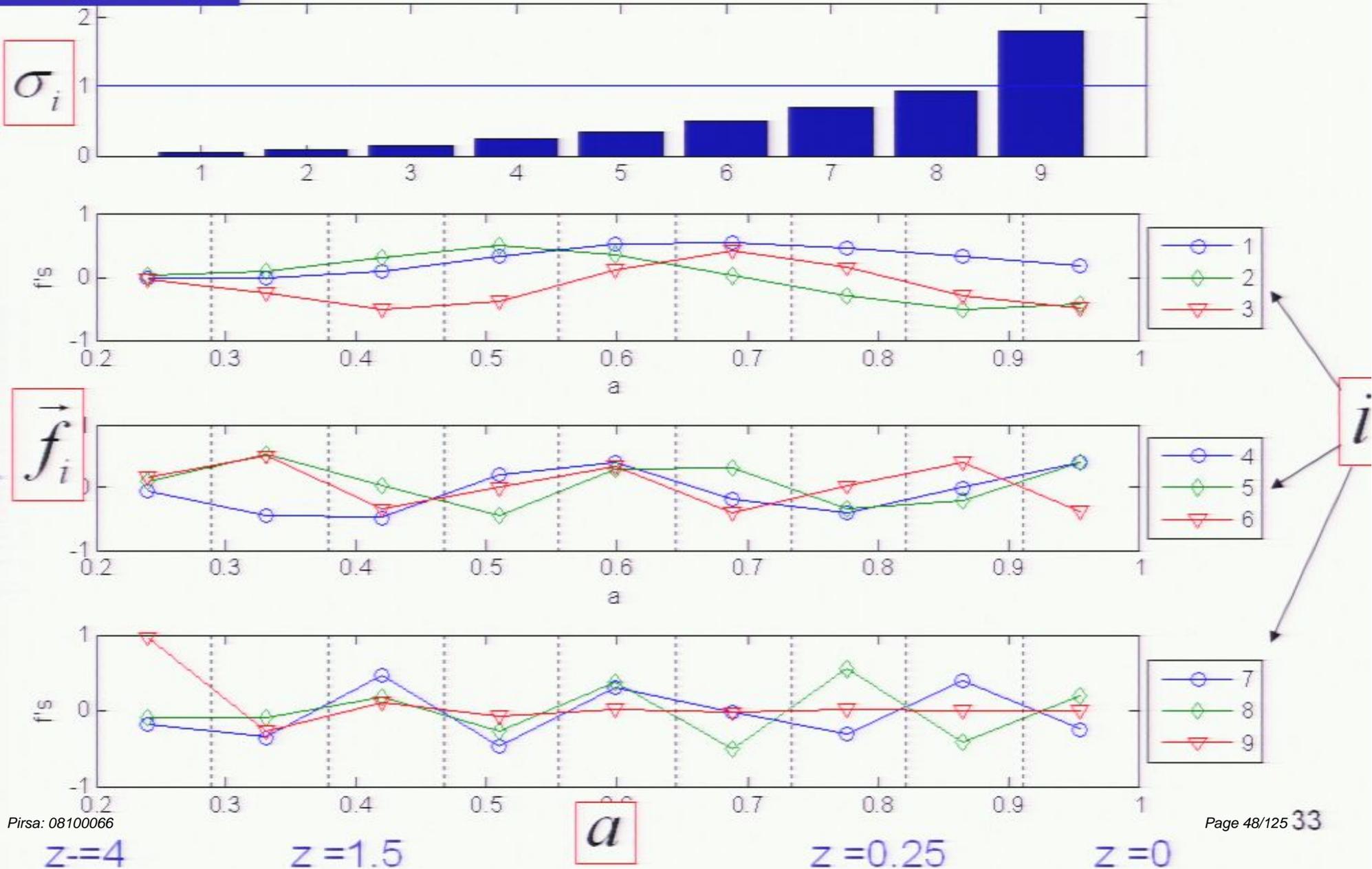
a

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Characterizing 9D ellipses by principle axes and corresponding errors

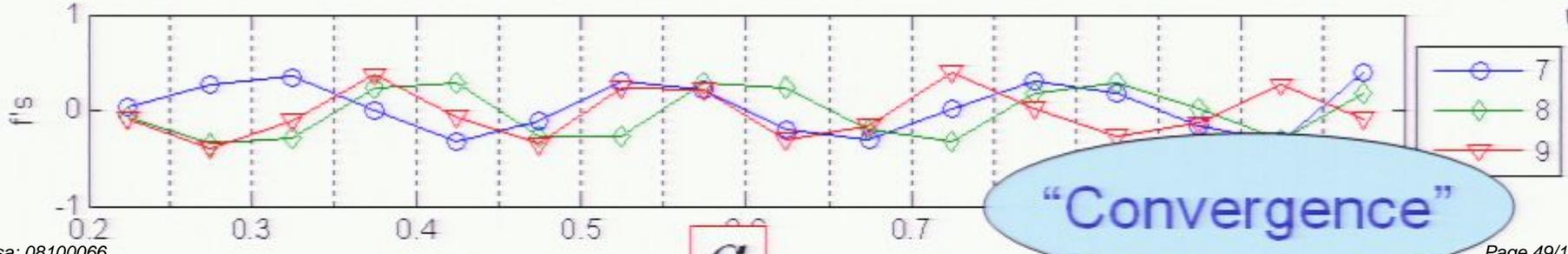
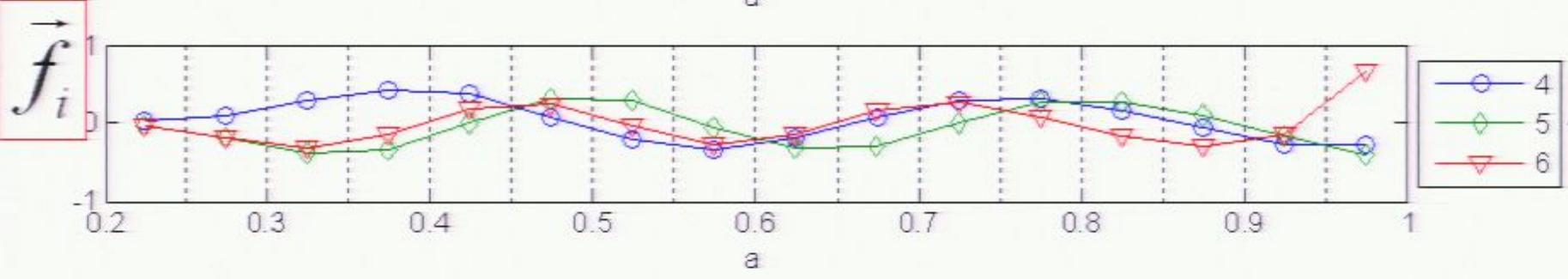
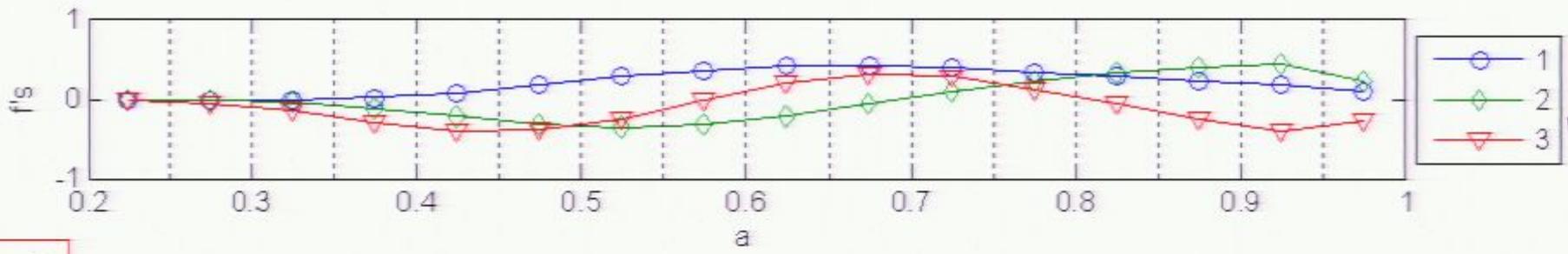
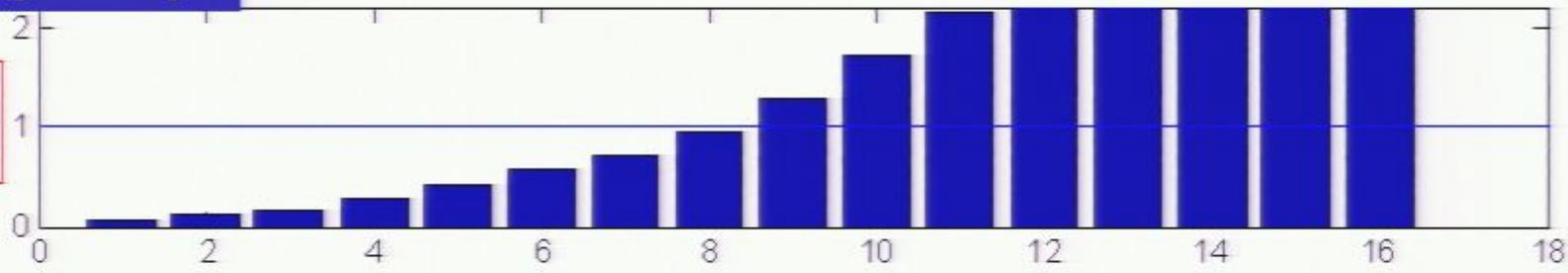
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Stage 4 Opt

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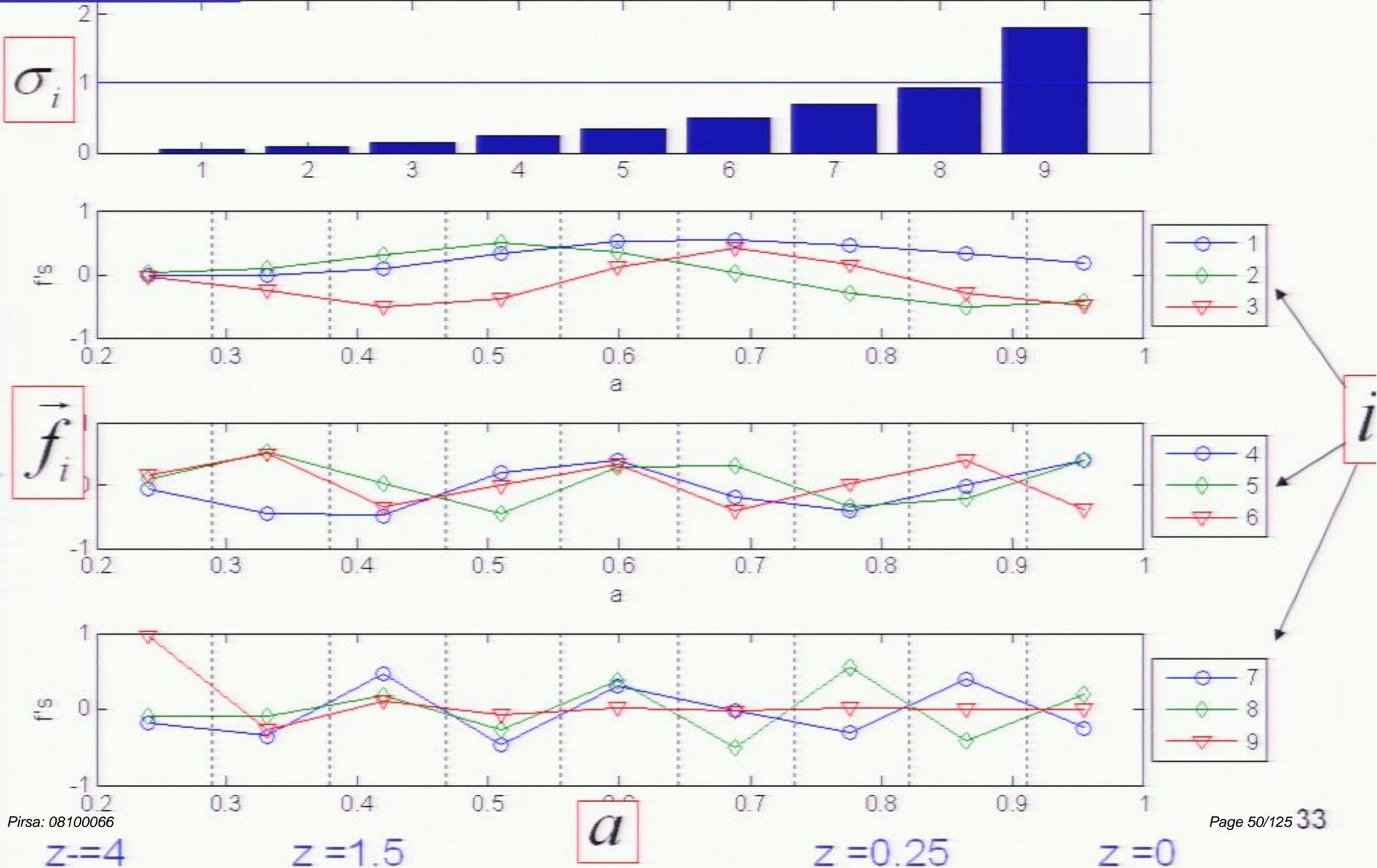
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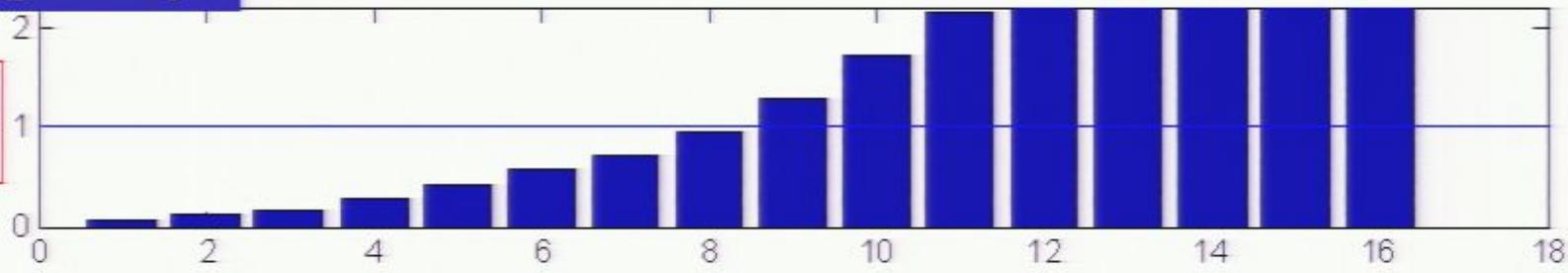
VL Stage 4 Opt



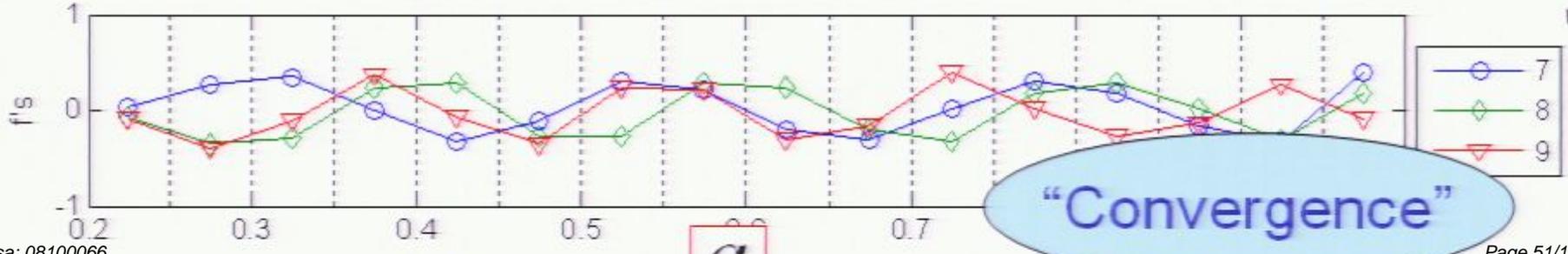
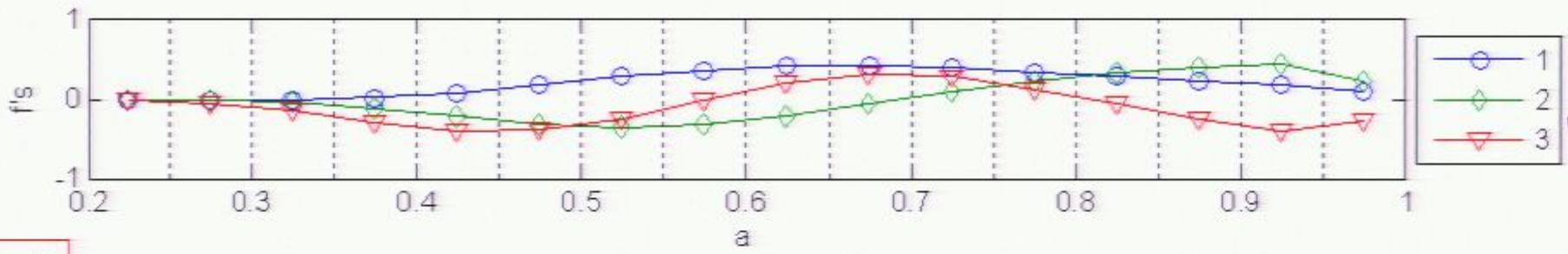
Characterizing 9D ellipses by principle axes and corresponding errors

Stage 4 Opt

σ_i



Principle axes



i

$z=-4$

$z=1.5$

a

$z=0.25$

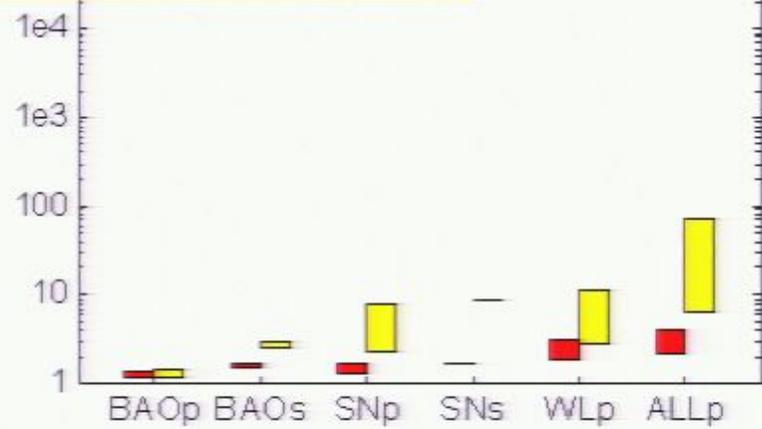
$z=0$

DETF(-CL)

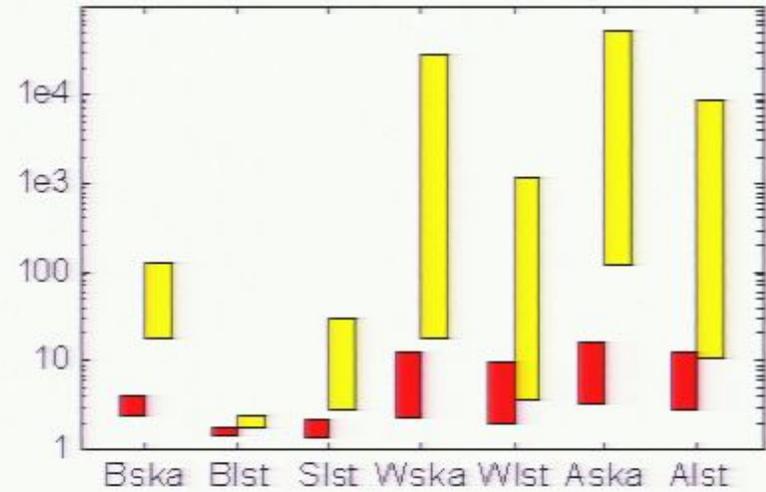
$\mathcal{F}_{\text{DETF/9D}}$

Stage 3

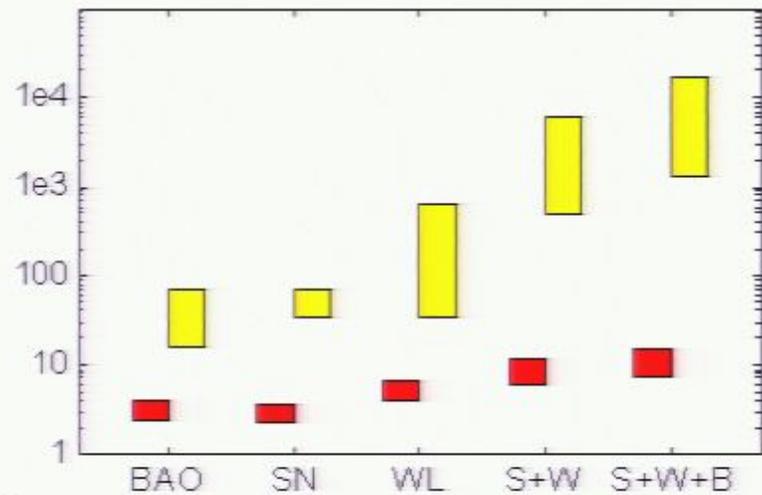
9D (-CL)



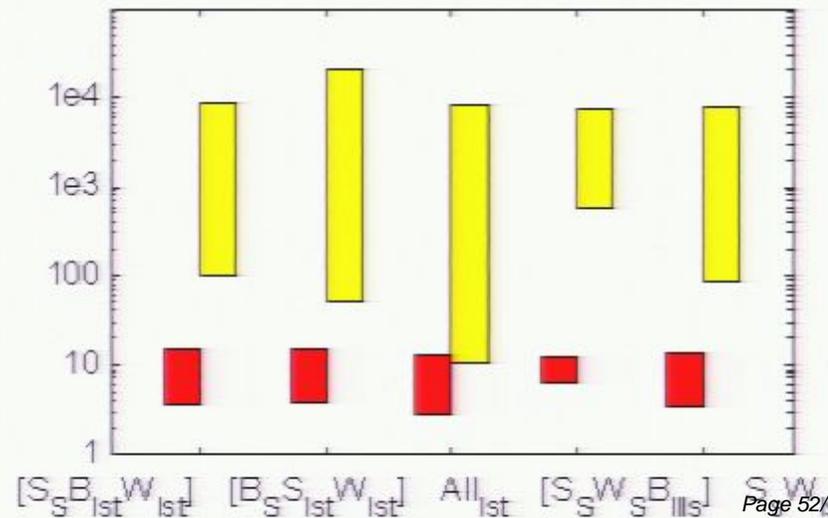
Stage 4 Ground



Stage 4 Space

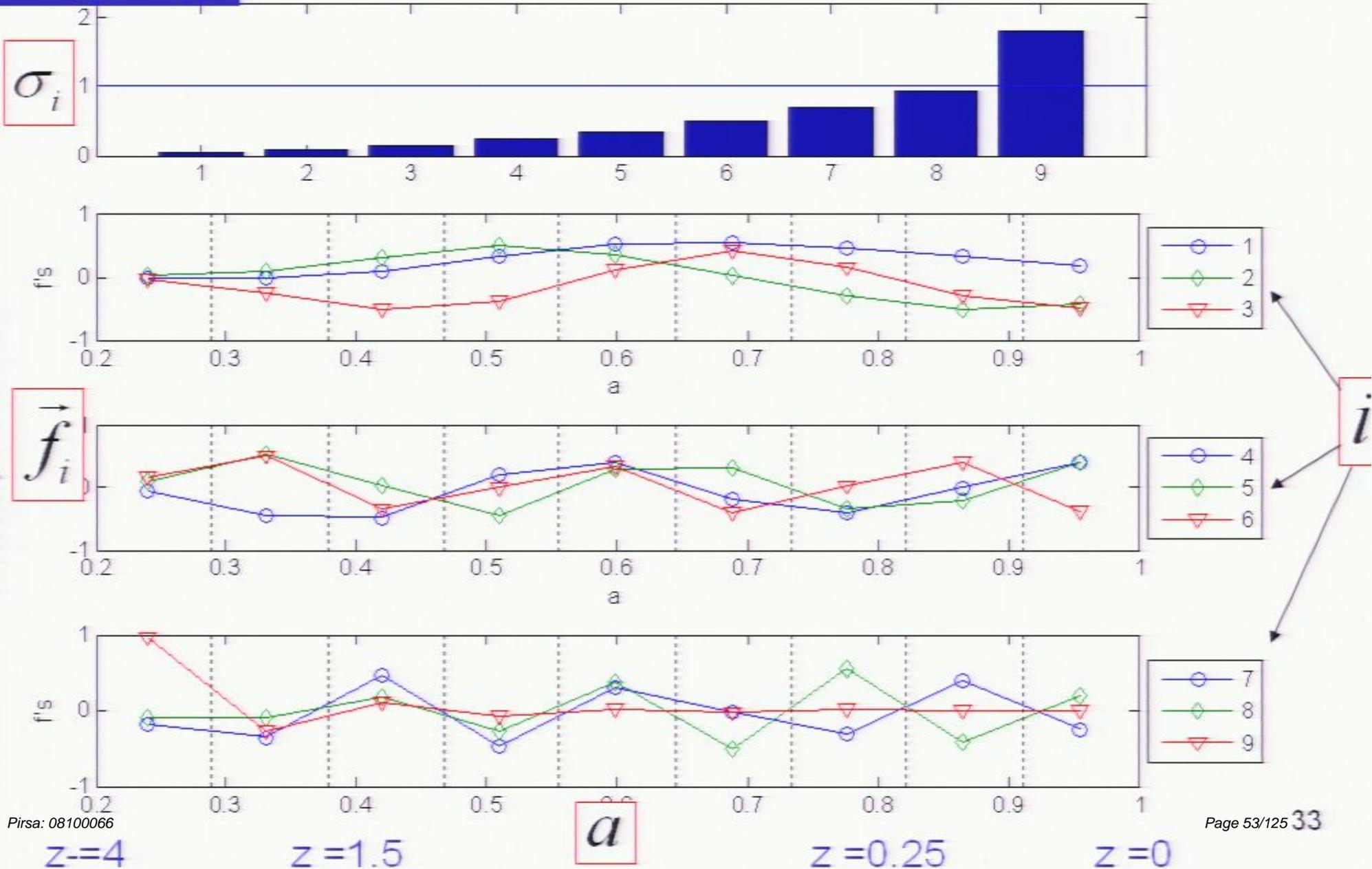


Stage 4 Ground+Space



Characterizing 9D ellipses by principle axes and corresponding errors

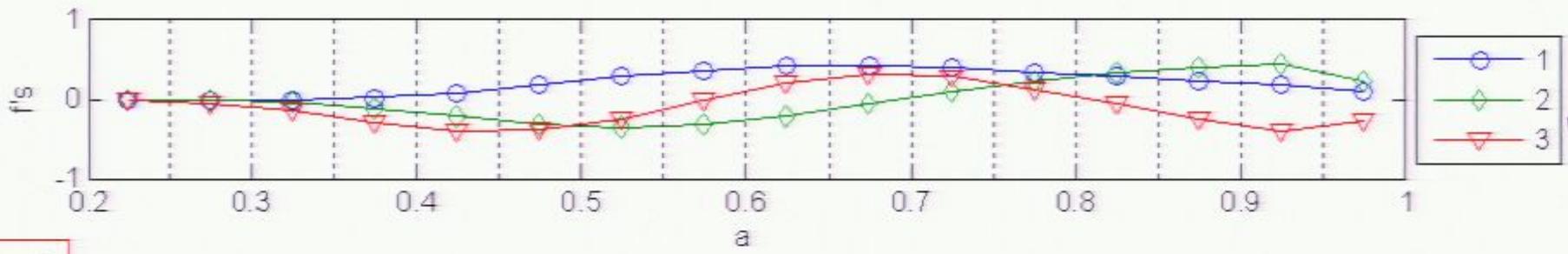
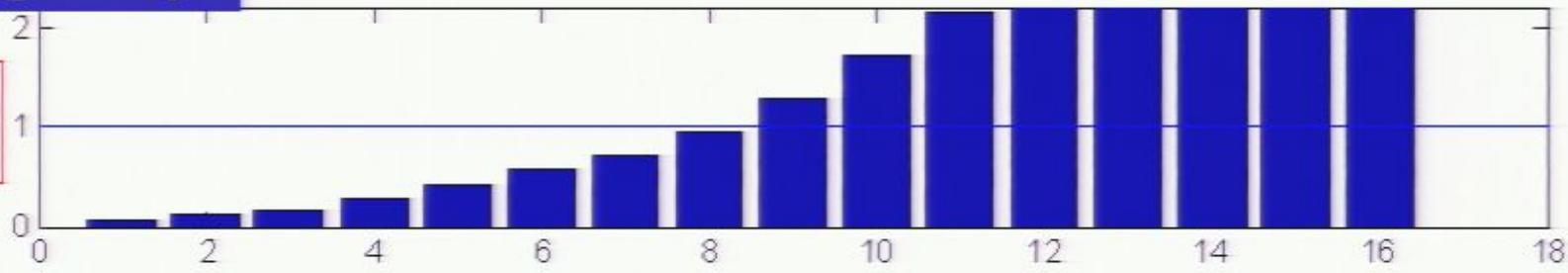
Stage 4 Opt



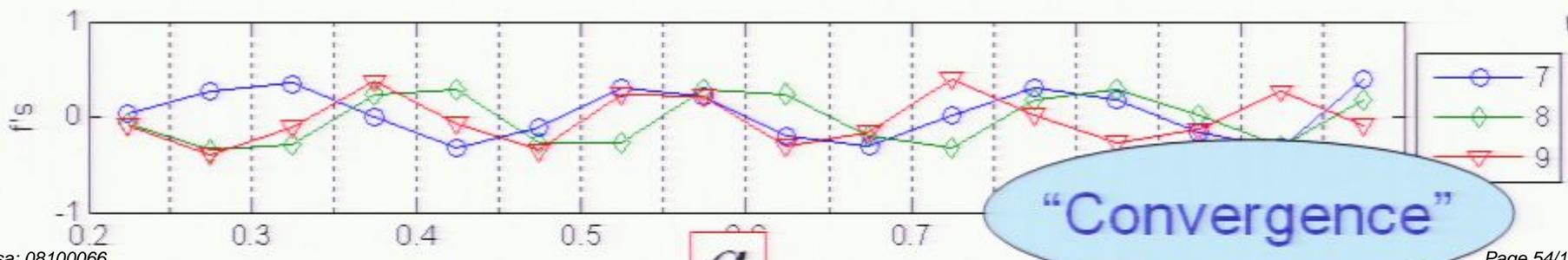
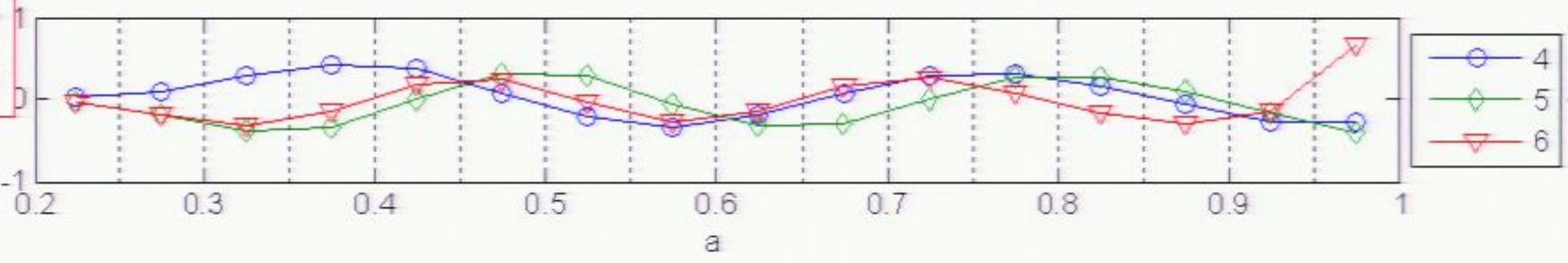
Characterizing 9D ellipses by principle axes and corresponding errors

Stage 4 Opt

σ_i



f_i



i

Multiple axes

$z=-4$

$z=1.5$

a

$z=0.25$

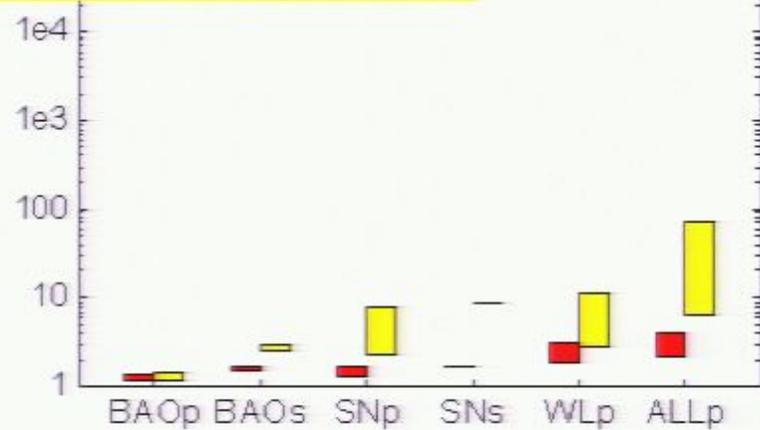
$z=0$

DETF(-CL)

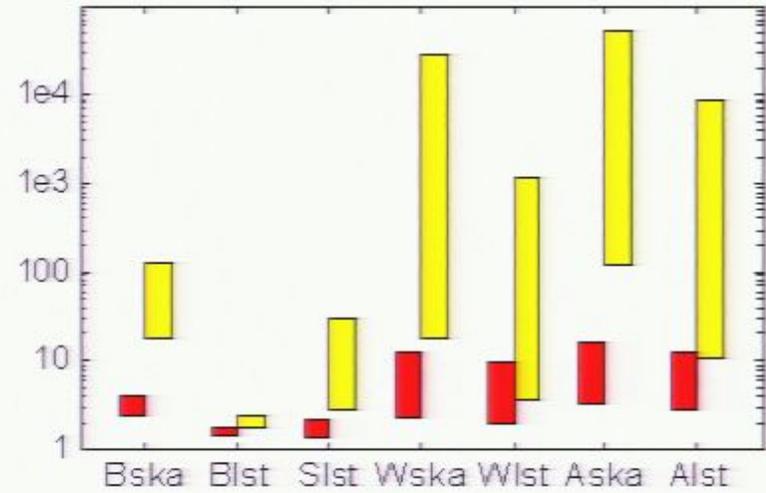
$\mathcal{F}_{\text{DETF/9D}}$

Stage 3

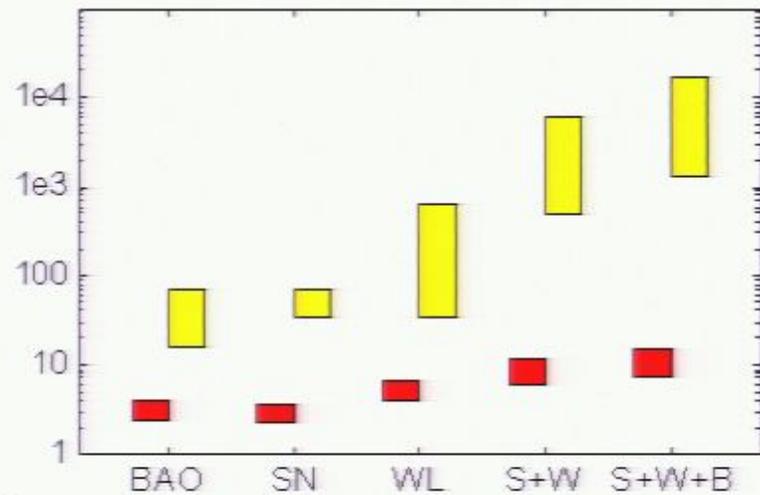
9D (-CL)



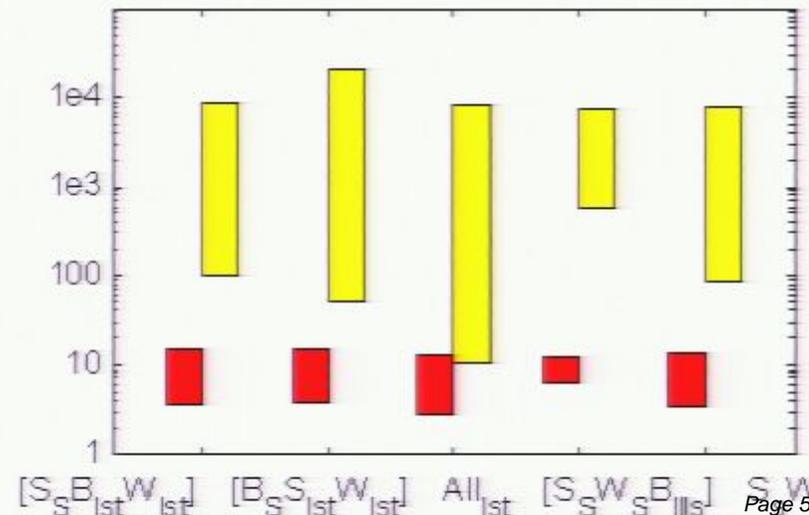
Stage 4 Ground



Stage 4 Space



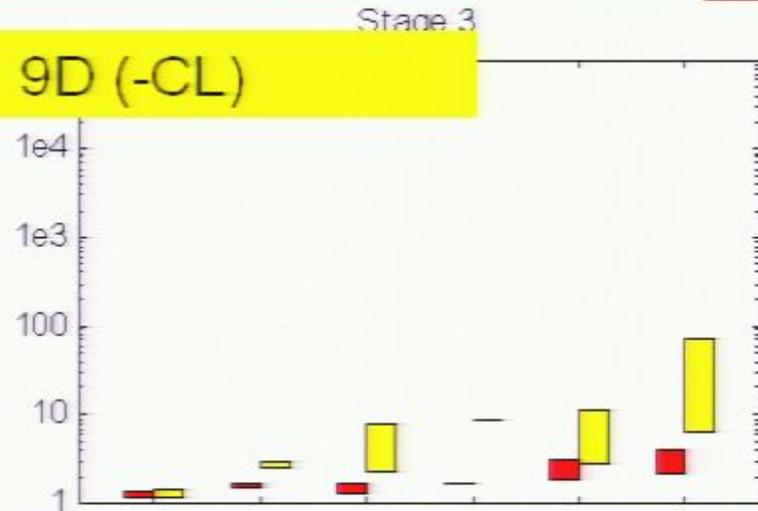
Stage 4 Ground+Space



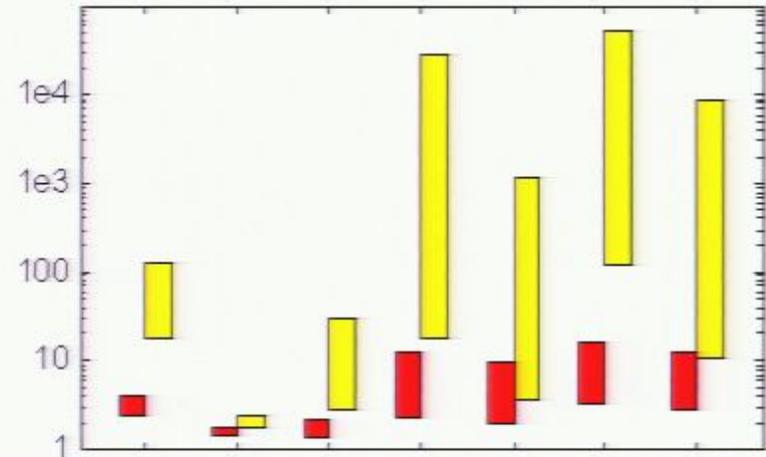
DETF(-CL)

$F_{\text{DETF/9D}}$

9D (-CL)

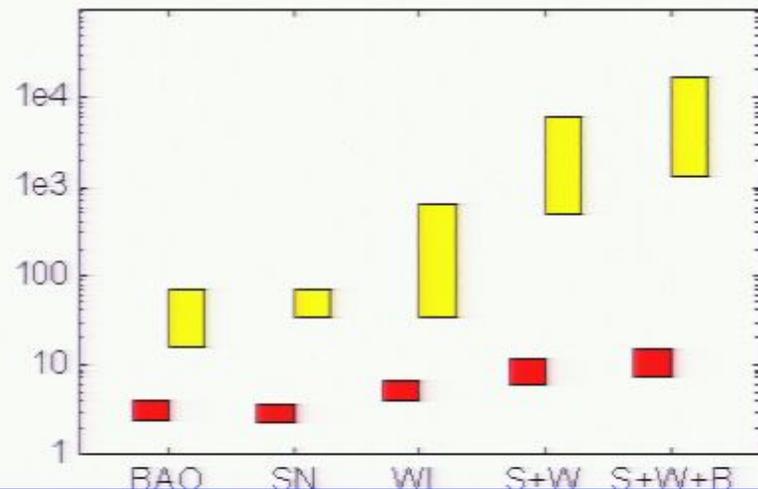


Stage 4 Ground

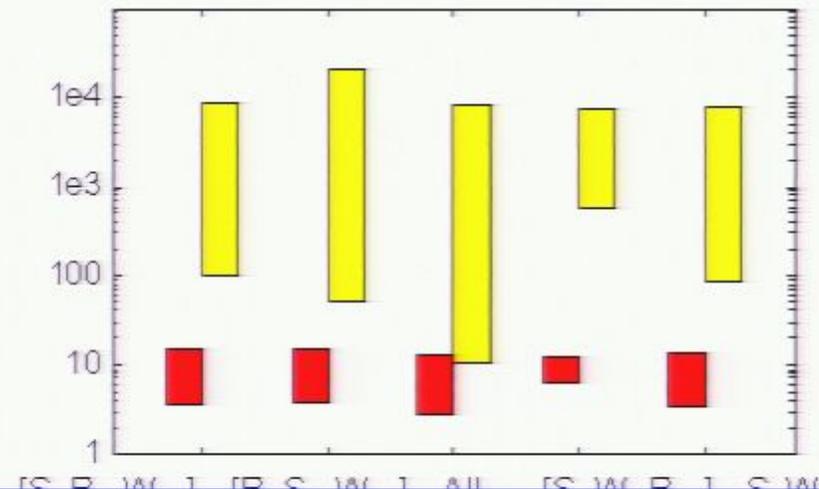


Stage 2 → Stage 3 = 1 order of magnitude (vs 0.5 for DETF)

Stage 4 Space



Stage 4 Ground+Space



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

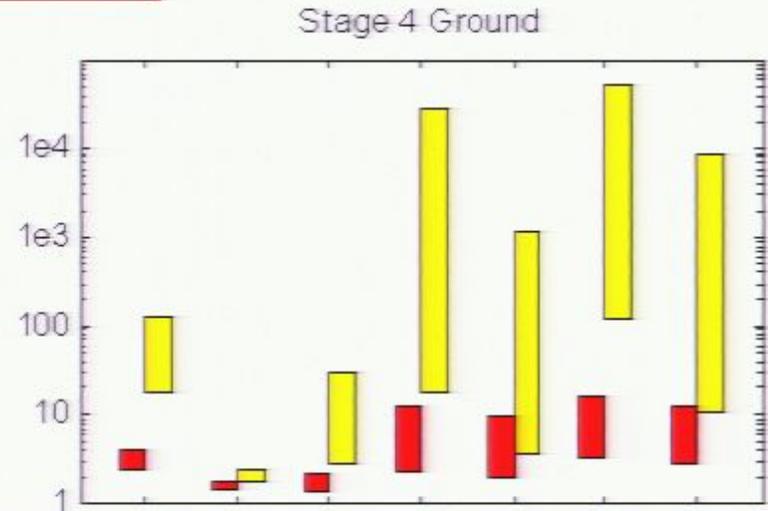
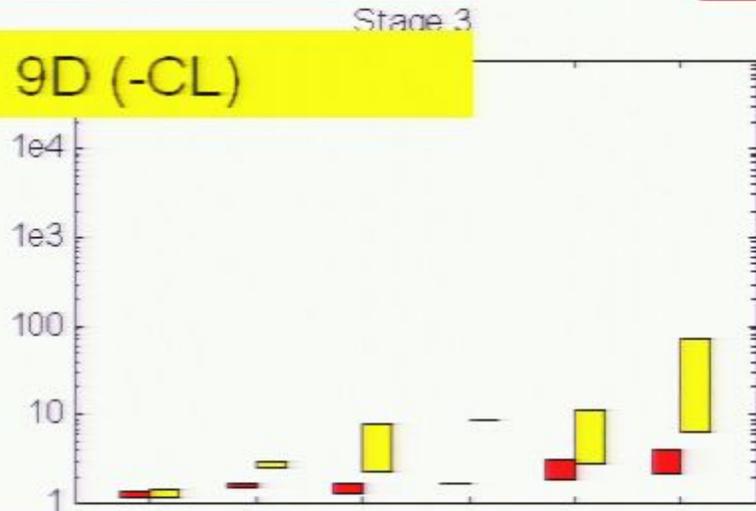
Upshot of 9D FoM:

- 1) DETF underestimates impact of expts
- 2) DETF underestimates relative value of Stage 4 vs Stage 3
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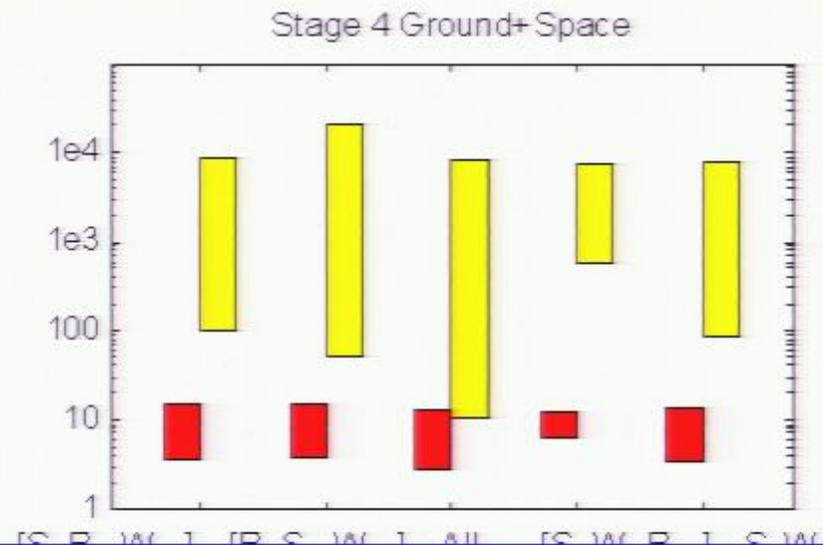
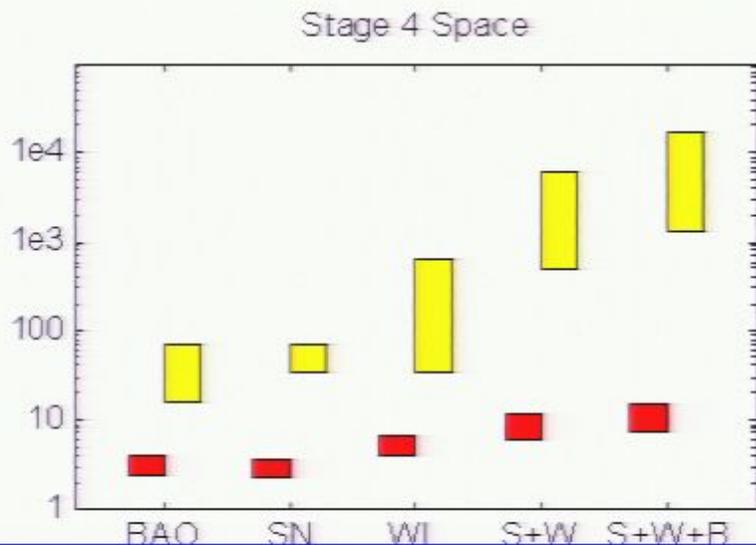
DETF(-CL)

$F_{\text{DETF/9D}}$

9D (-CL)



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



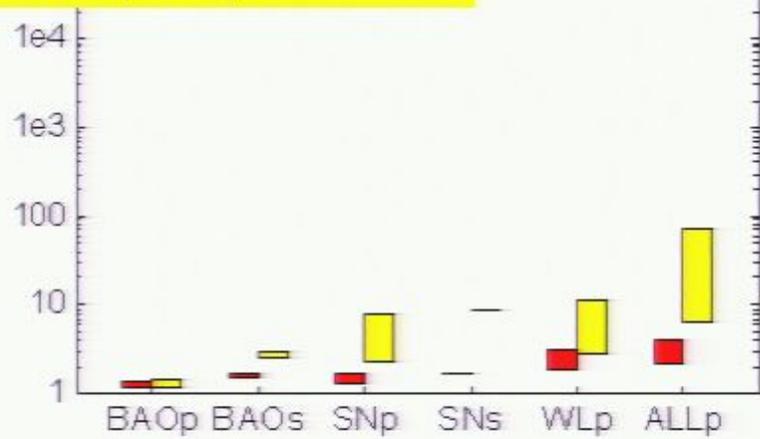
Stage 2 → Stage 4 = 3 orders of magnitude (vs 1 for DETF)

DETF(-CL)

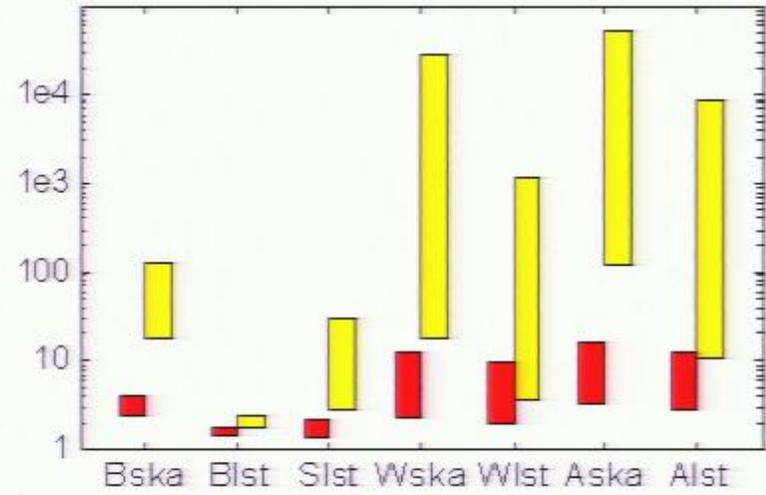
$\mathcal{F}_{\text{DETF/9D}}$

Stage 3

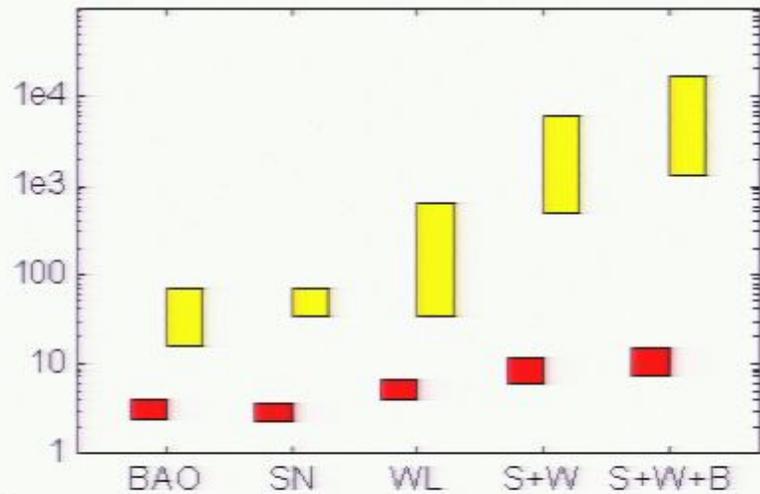
9D (-CL)



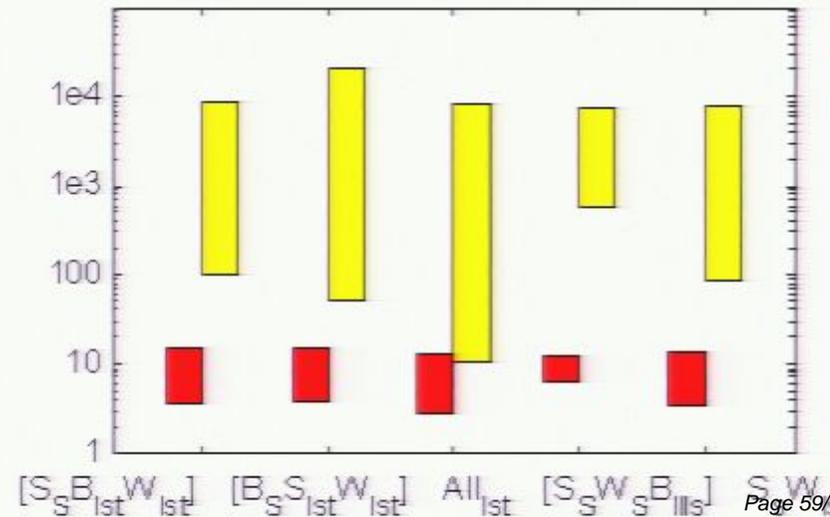
Stage 4 Ground



Stage 4 Space



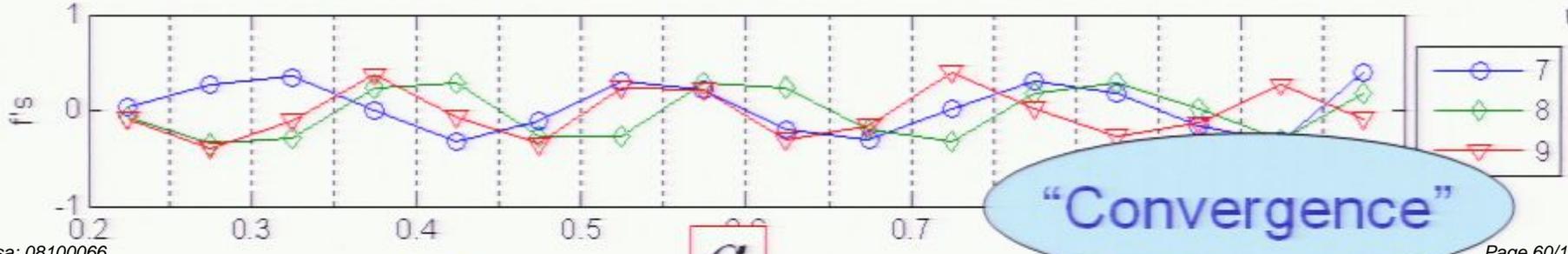
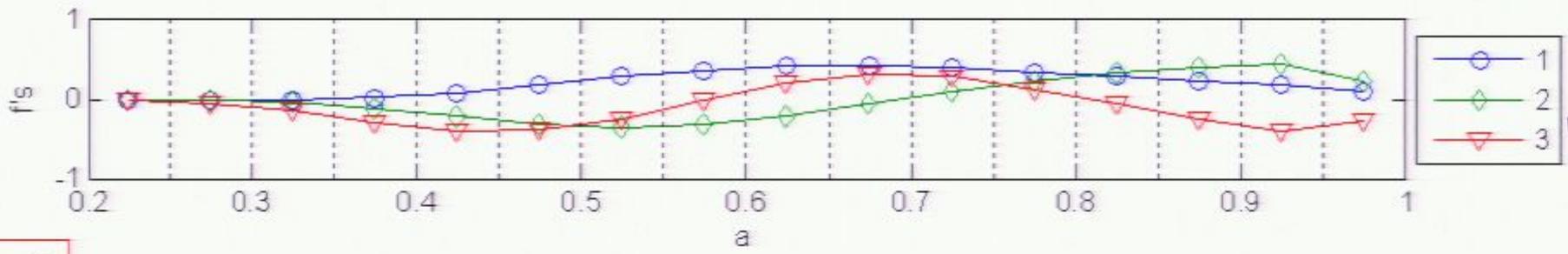
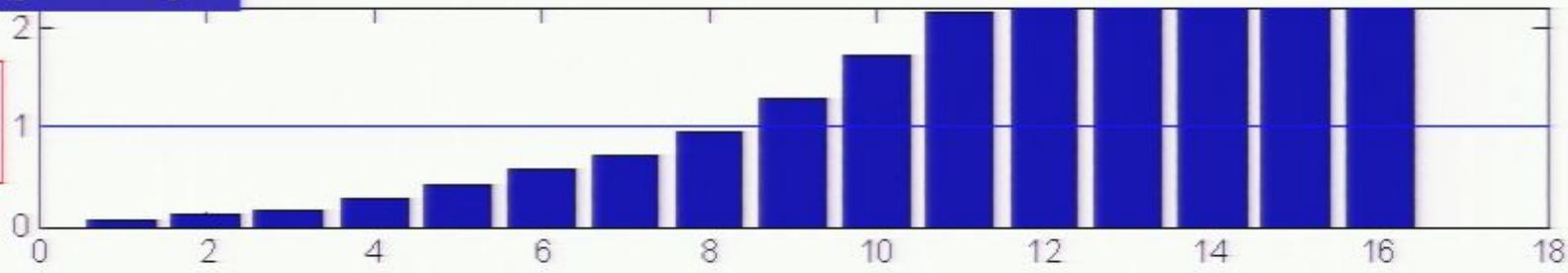
Stage 4 Ground+Space



Characterizing 9D ellipses by principle axes and corresponding errors

Stage 4 Opt

σ_i



Principle axes

i

a

“Convergence”

$z=-4$

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Inverts
cost/FoM
Estimates
S3 vs S4

Upshot of 9D FoM:

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 - 3) The above can be understood approximately in terms of a simple rescaling
 - 4) DETF FoM is fine for most purposes (ranking, value of combinations etc).
- A nice way to gain insights into data (real or imagined)

Followup questions:

- In what ways might the choice of DE parameters have skewed the DETF results?
- What impact can these data sets have on specific DE models (vs abstract parameters)?
- To what extent can these data sets deliver discriminating power between specific DE models?
- How is the DoE/ESA/NASA Science Working Group looking at these questions?

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A: Only by an overall (possibly important) rescaling

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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

Mark Yashar

+AA

+

DETF
Simulated data

+

Quintessence
potentials

+

MCMC

See also Dutta & Sorbo 2006, Huterer and Turner 1999
& especially *Huterer and Peiris 2006*

The potentials

Exponential (Wetterich, Peebles & Ratra)

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

PNGB aka Axion (Frieman et al)

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

Exponential with prefactor (AA & Skordis)

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

Inverse Power Law (Ratra & Peebles, Steinhardt et al)

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Stronger than
average
motivations &
interest

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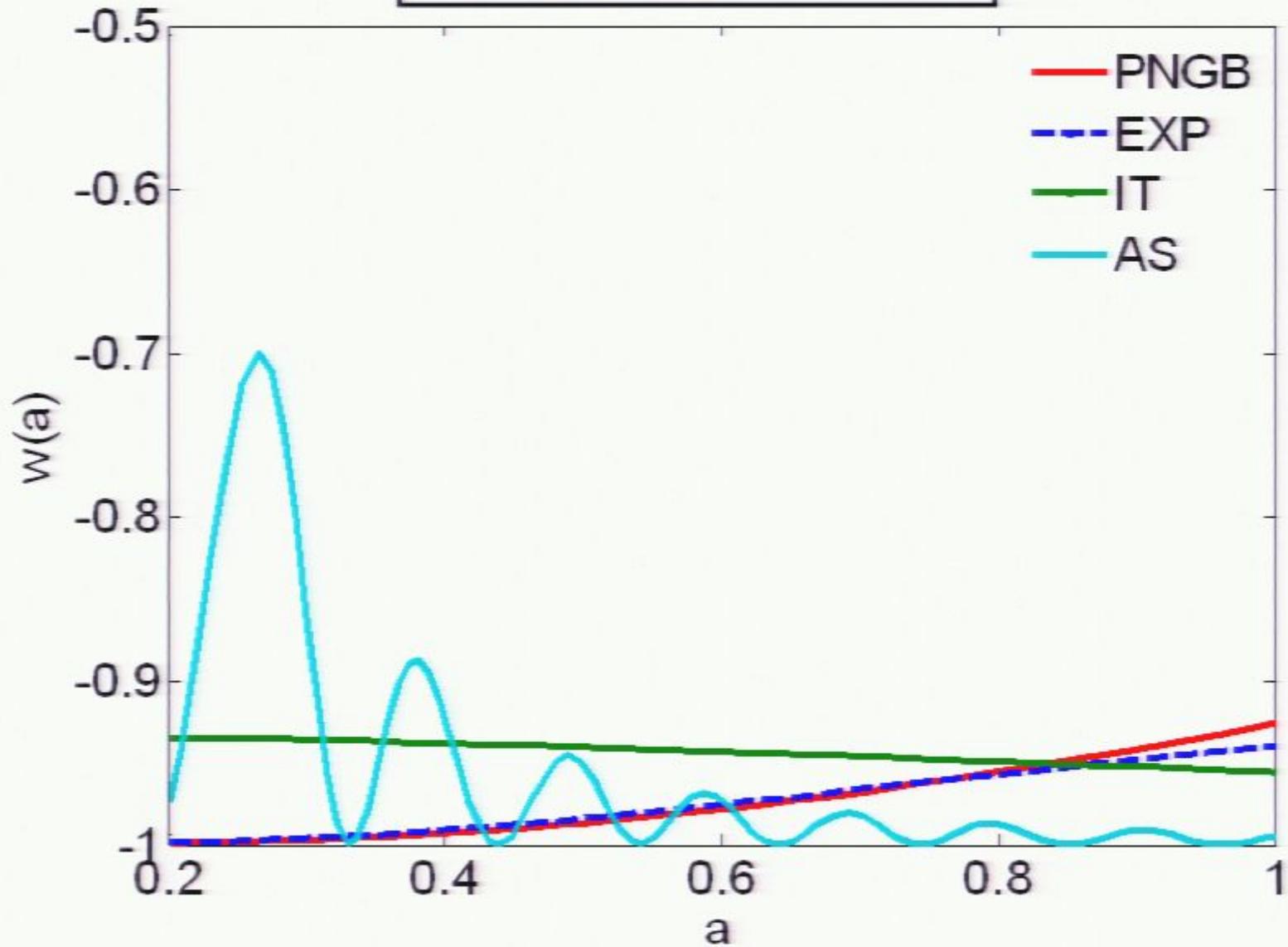
ArXiv Dec 08,
PRD in press

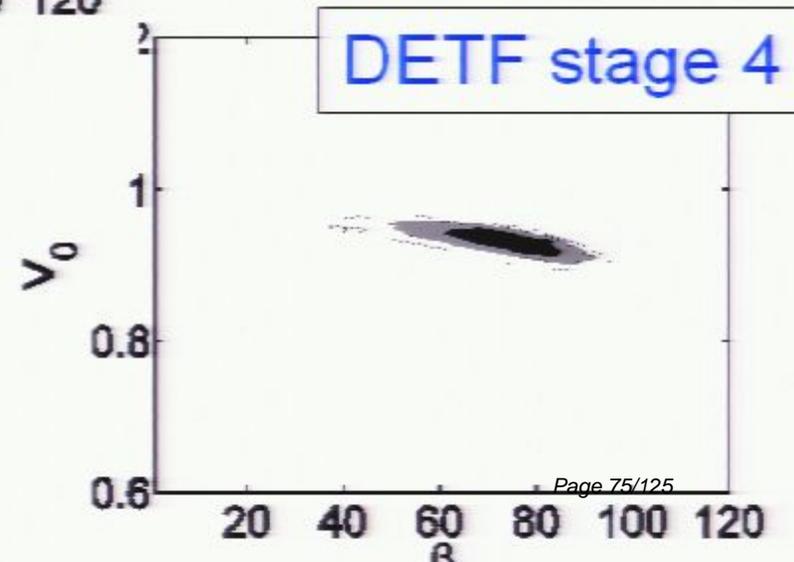
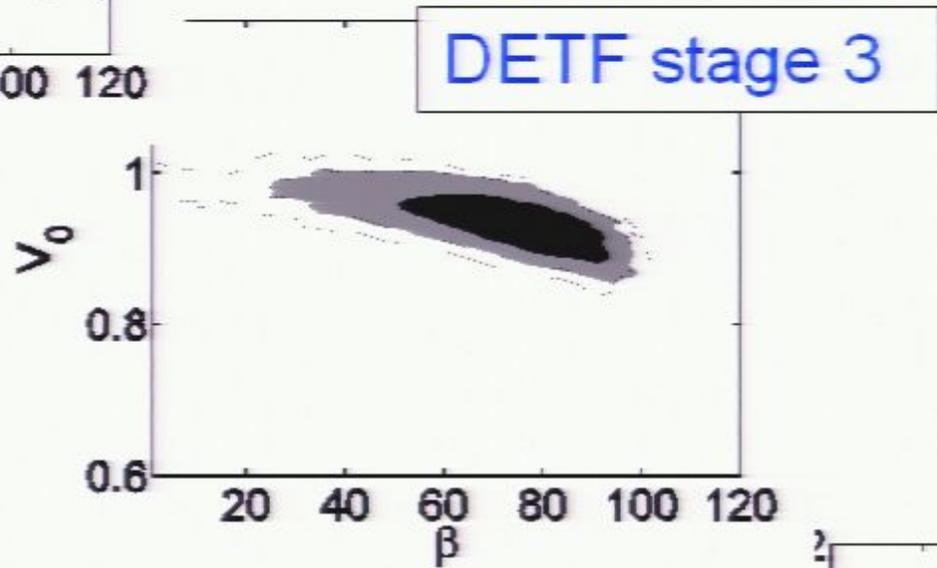
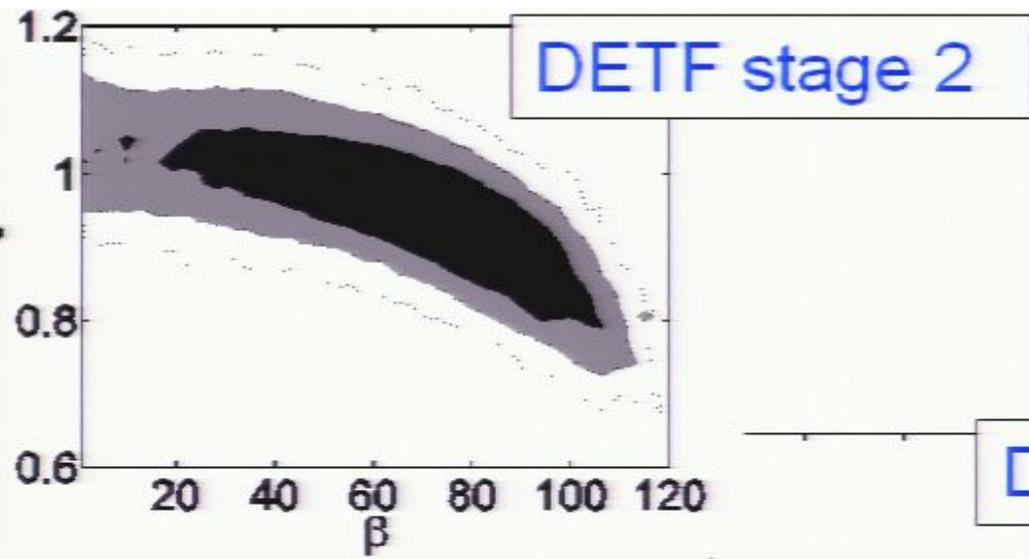
Inverse Tracker (Ratra & Peebles, Steinhardt et al)

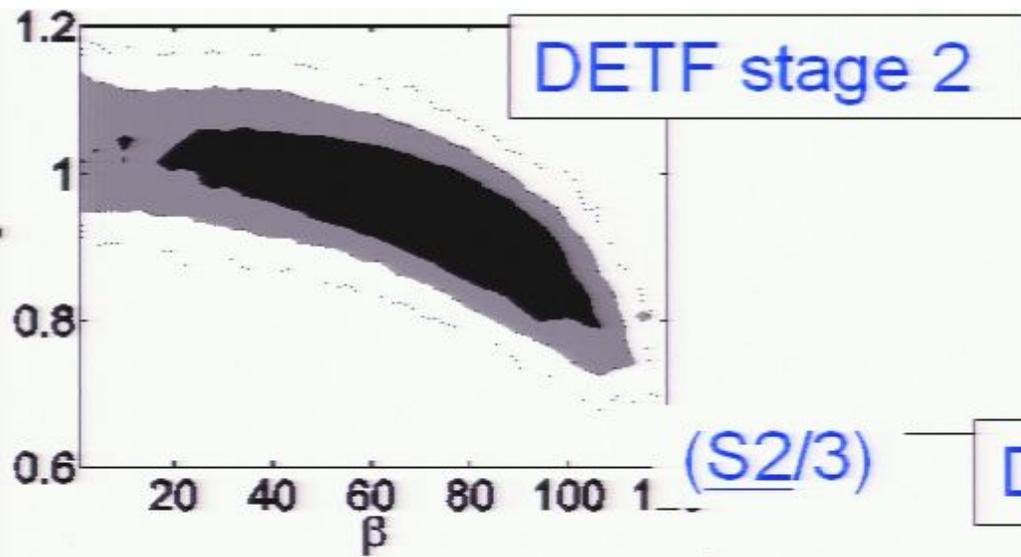
$$V(\varphi) = V_0 \left(\frac{m}{\varphi} \right)^\alpha$$

← In prep.

...they cover a variety of behavior.





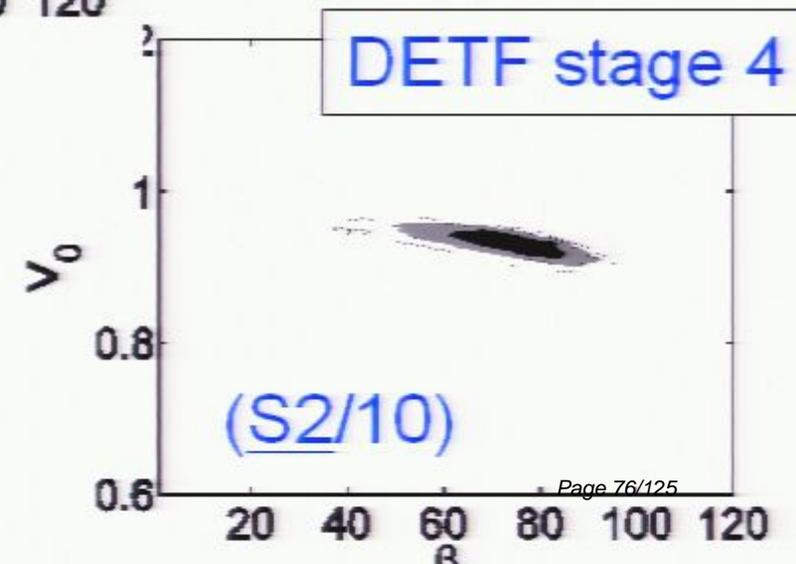


(S2/3)



Upshot:

Story in scalar field parameter space very similar to DETF story in w_0 - w_a space.



(S2/10)

Followup questions:

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A: Very similar to DETF results in w_0 - w_a space

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Michael Barnard *et al* [arXiv:0804.0413](https://arxiv.org/abs/0804.0413)

Problem:

Each scalar field model is defined in its own parameter space. How should one quantify discriminating power among models?

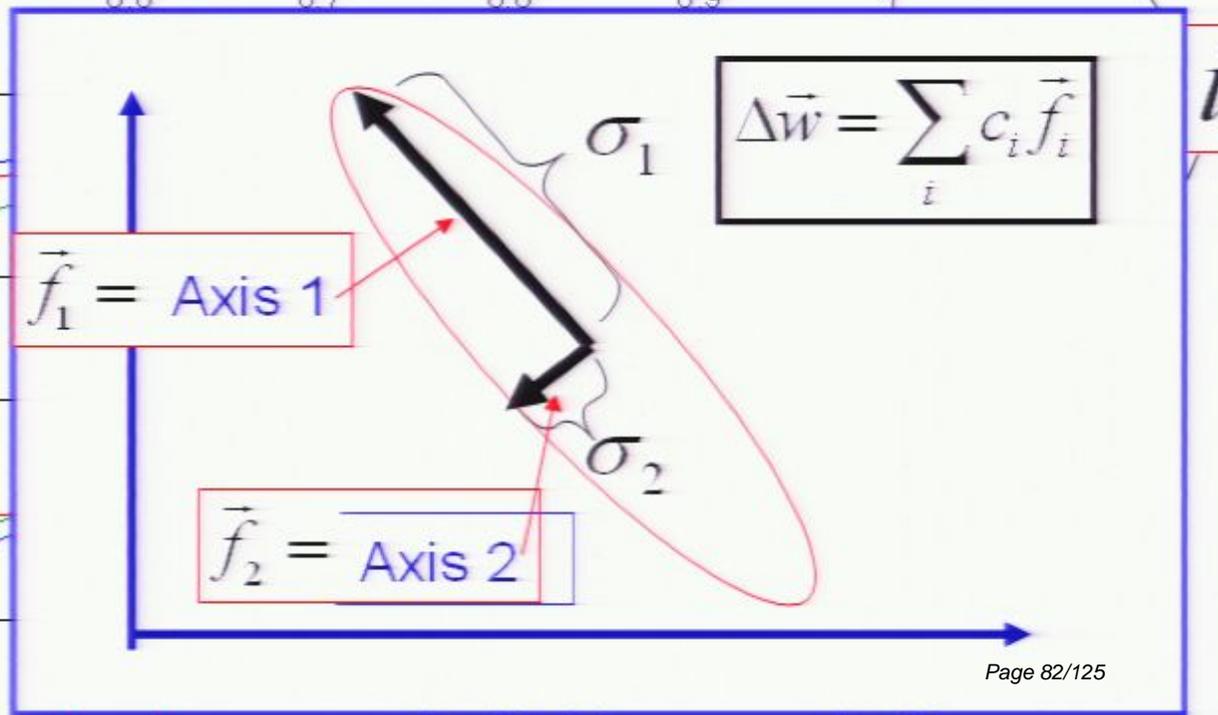
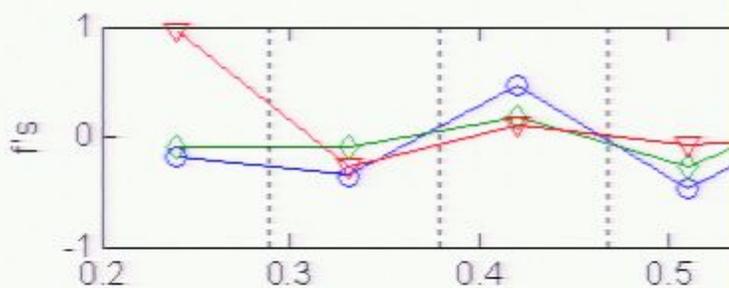
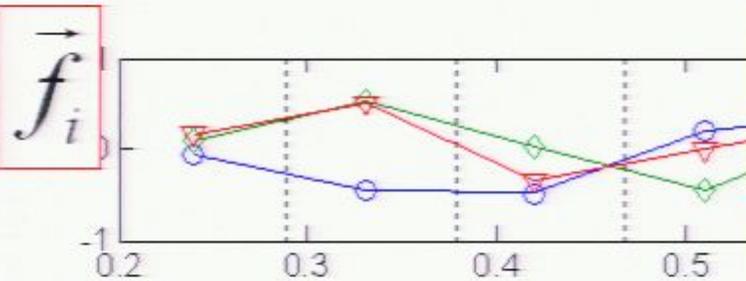
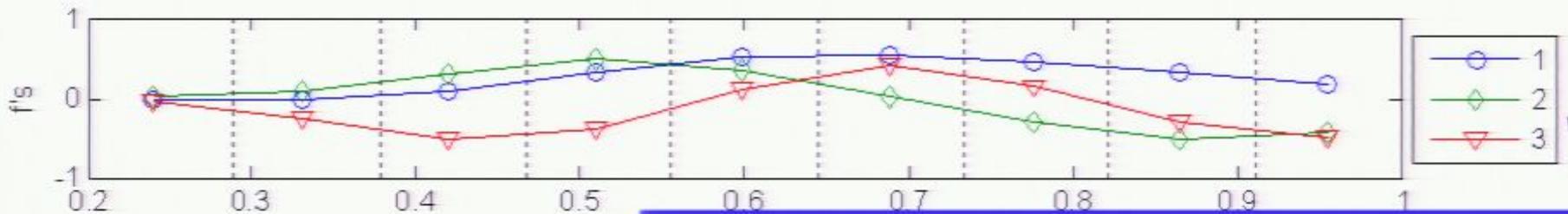
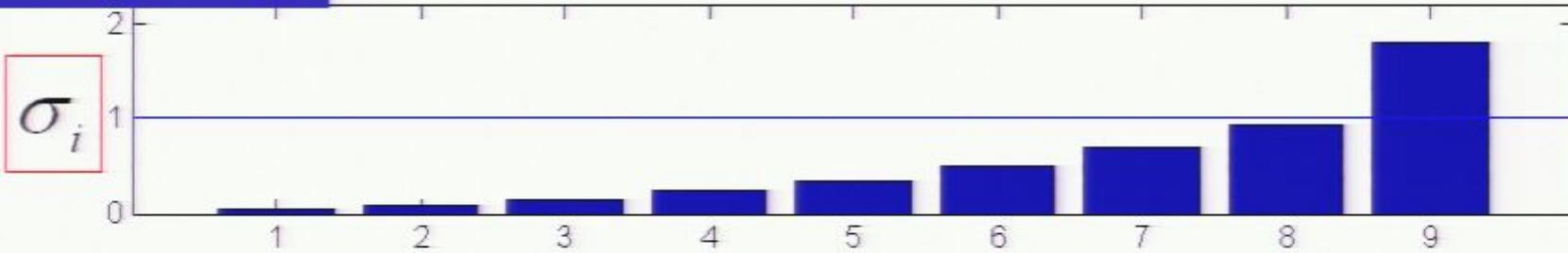
Our answer:

→ Form each set of scalar field model parameter values, map the solution into $w(a)$ eigenmode space, the space of uncorrelated observables.

→ Make the comparison in the space of uncorrelated observables.

Characterizing 9D ellipses by principle axes and corresponding errors

Stage 4 Opt



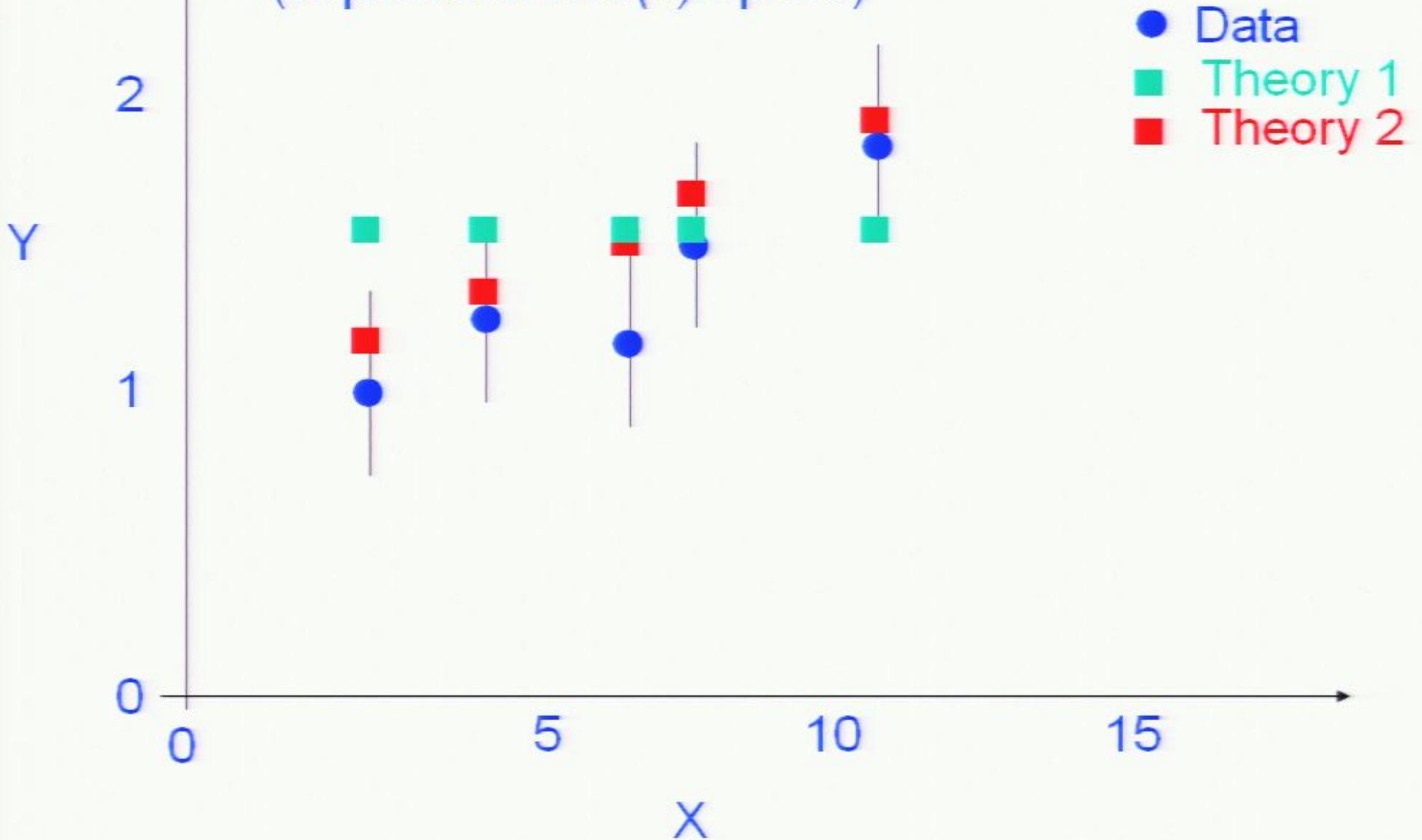
$z = -4$

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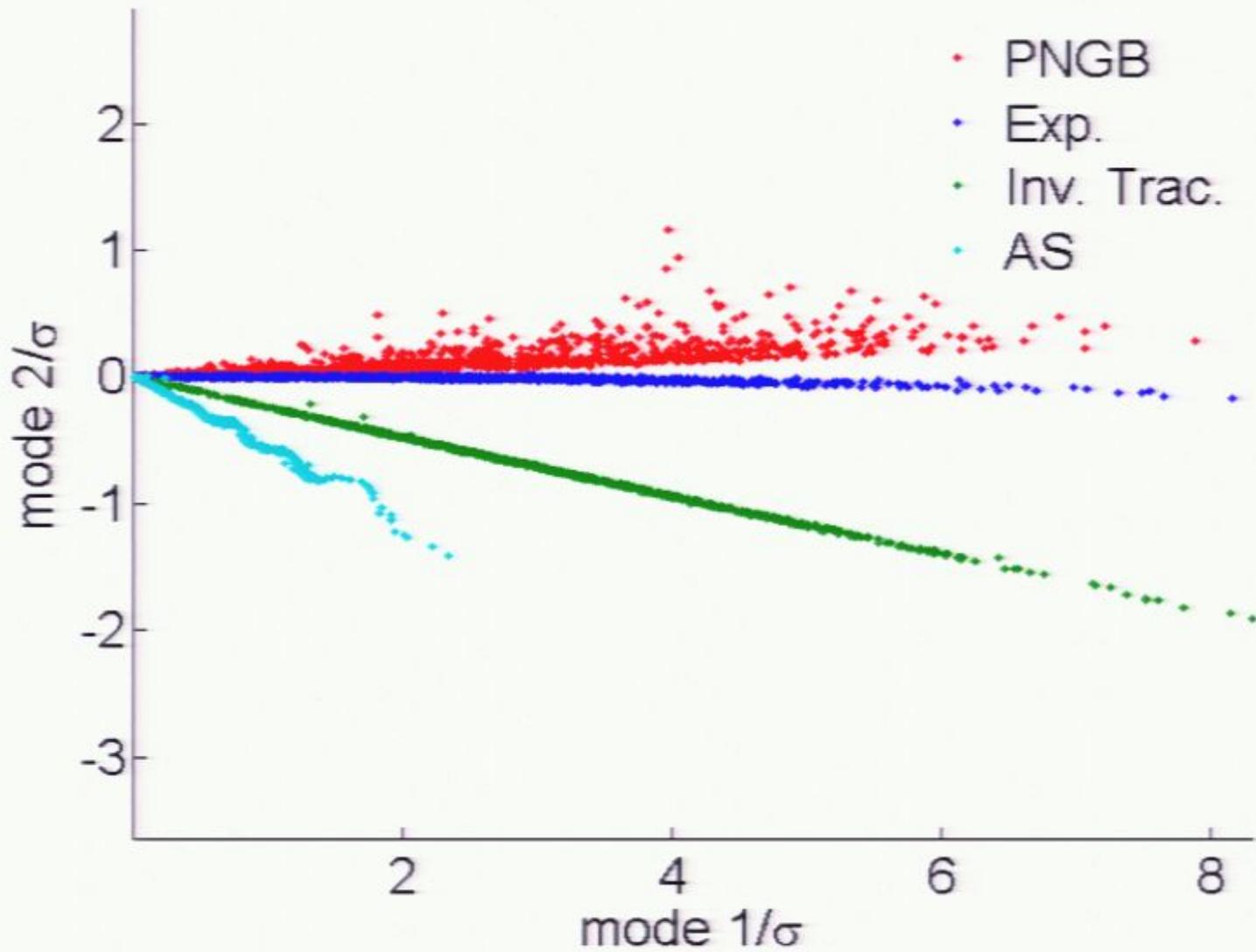
$z = 0$

Concept: Uncorrelated data points
(expressed in $w(a)$ space)



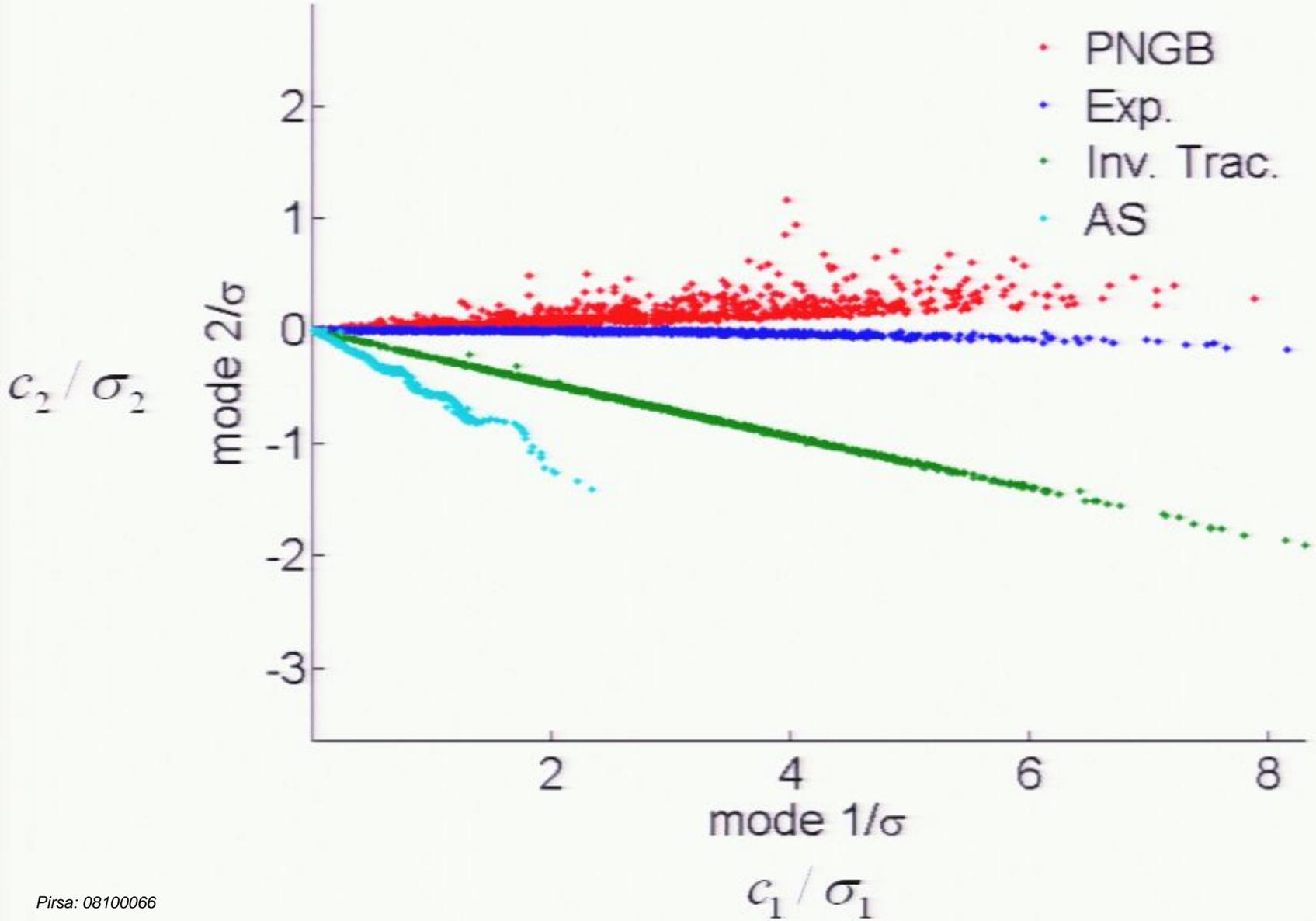
Starting point: MCMC chains giving distributions for each model at Stage 2.

DETF Stage 3 photo [Opt]



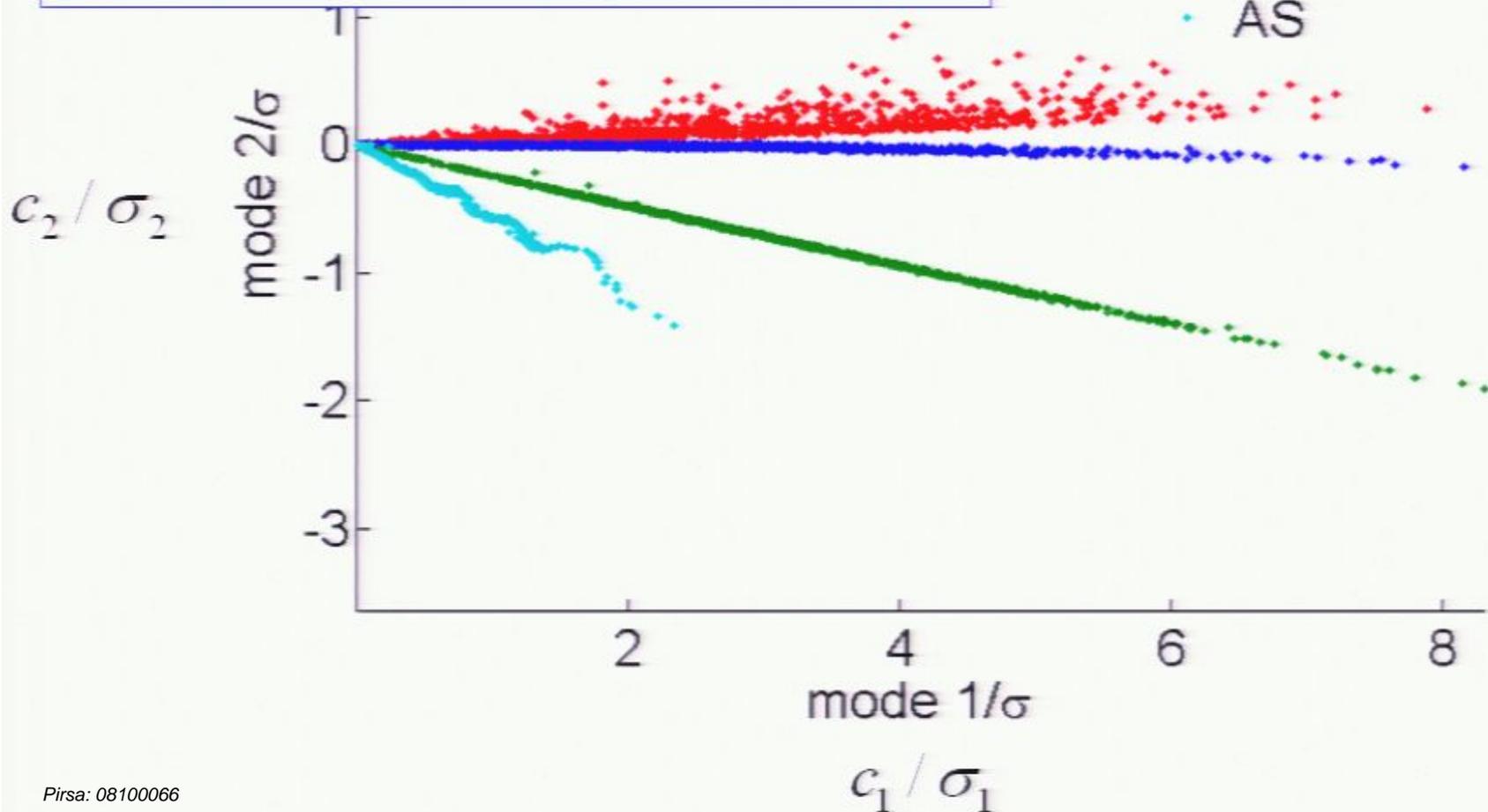
$$\Delta \vec{w} = \sum_i c_i \vec{f}_i$$

DETF Stage 3 photo [Opt]

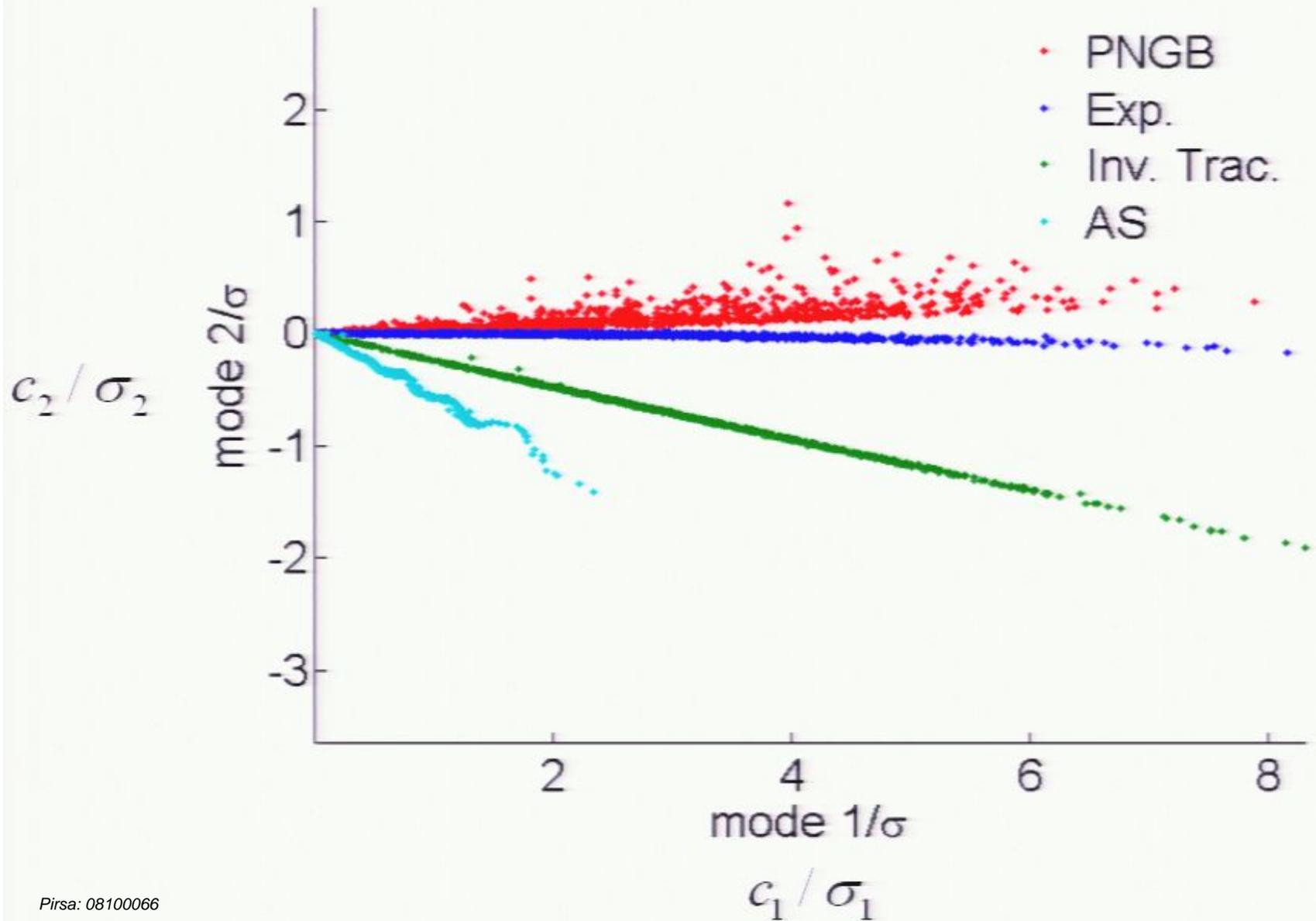


- Distinct model locations
- mode amplitude/ σ_i “physical”
- Modes (and σ_i 's) reflect specific expts.

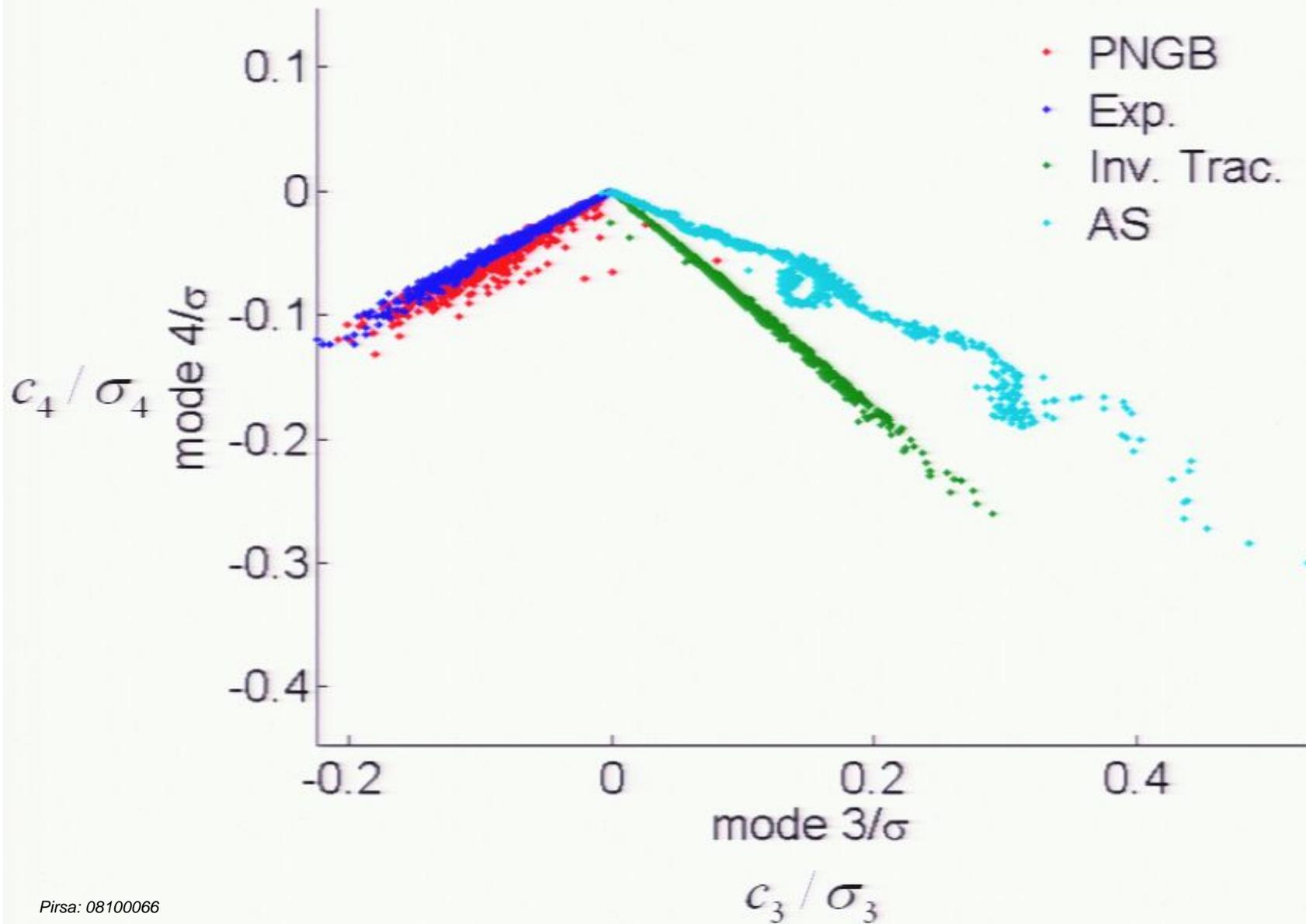
- PNGB
- Exp.
- Inv. Trac.
- AS



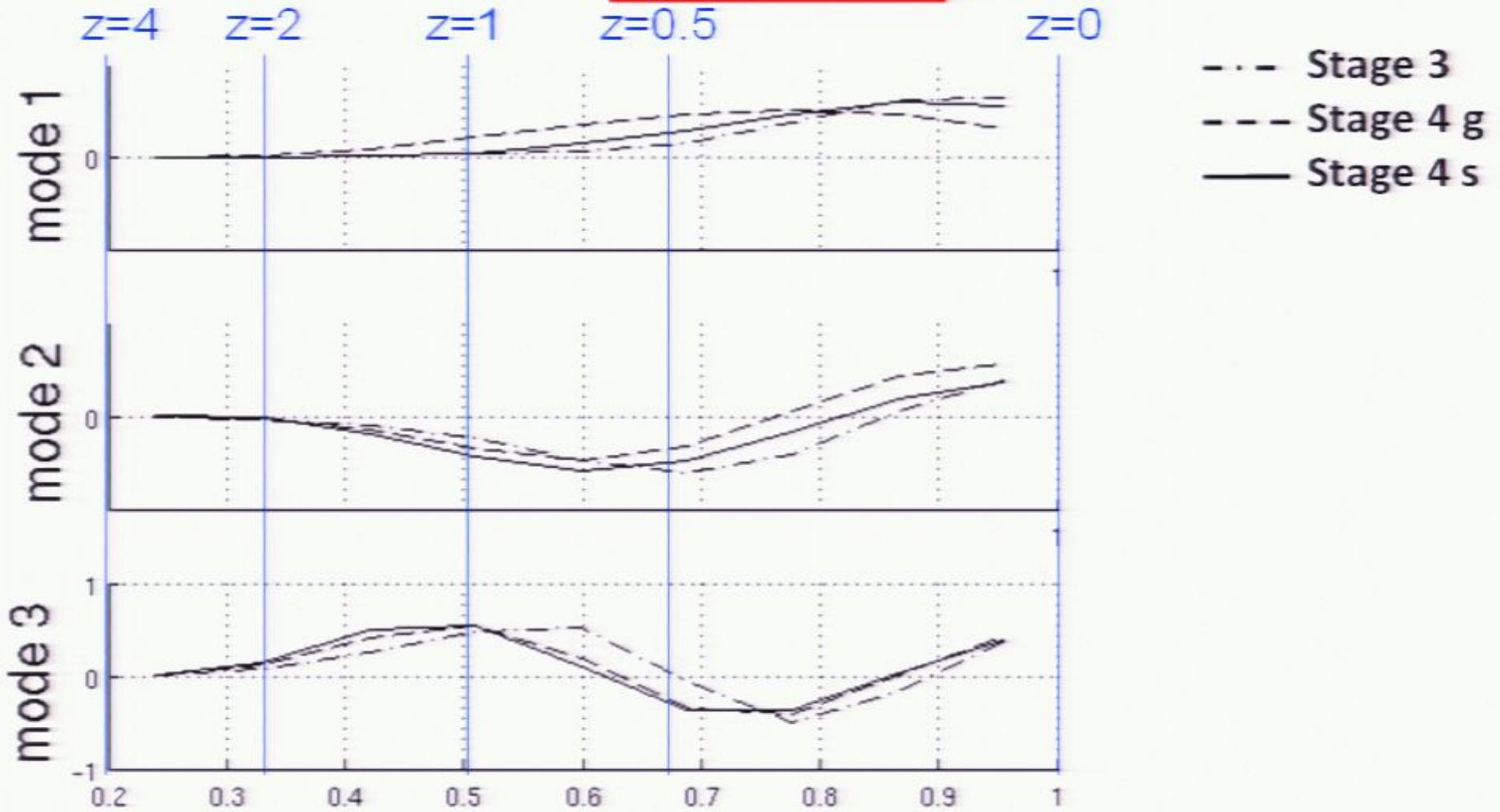
DETF Stage 3 photo [Opt]



DETF Stage 3 photo [Opt]

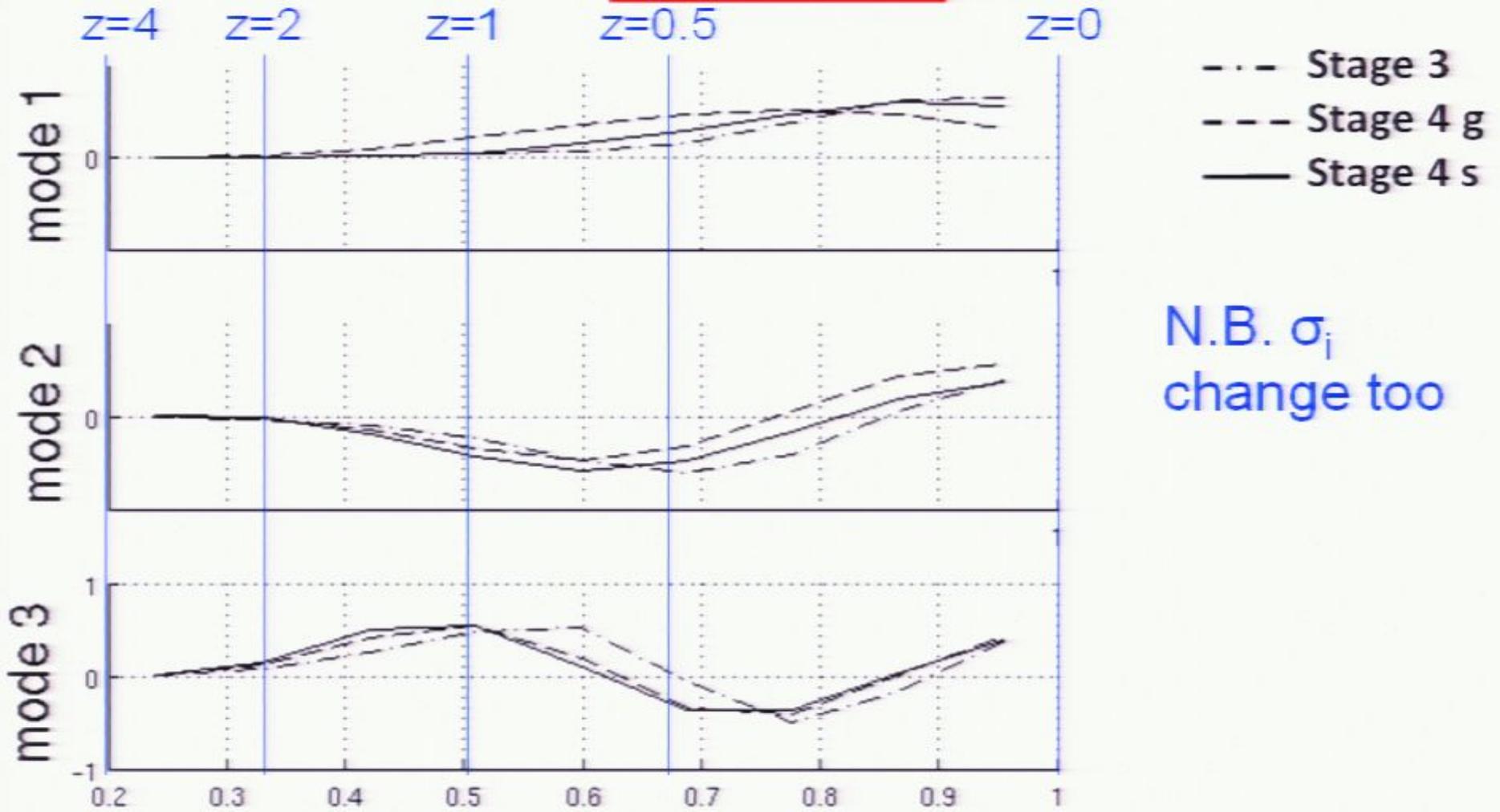


Eigenmodes:



a

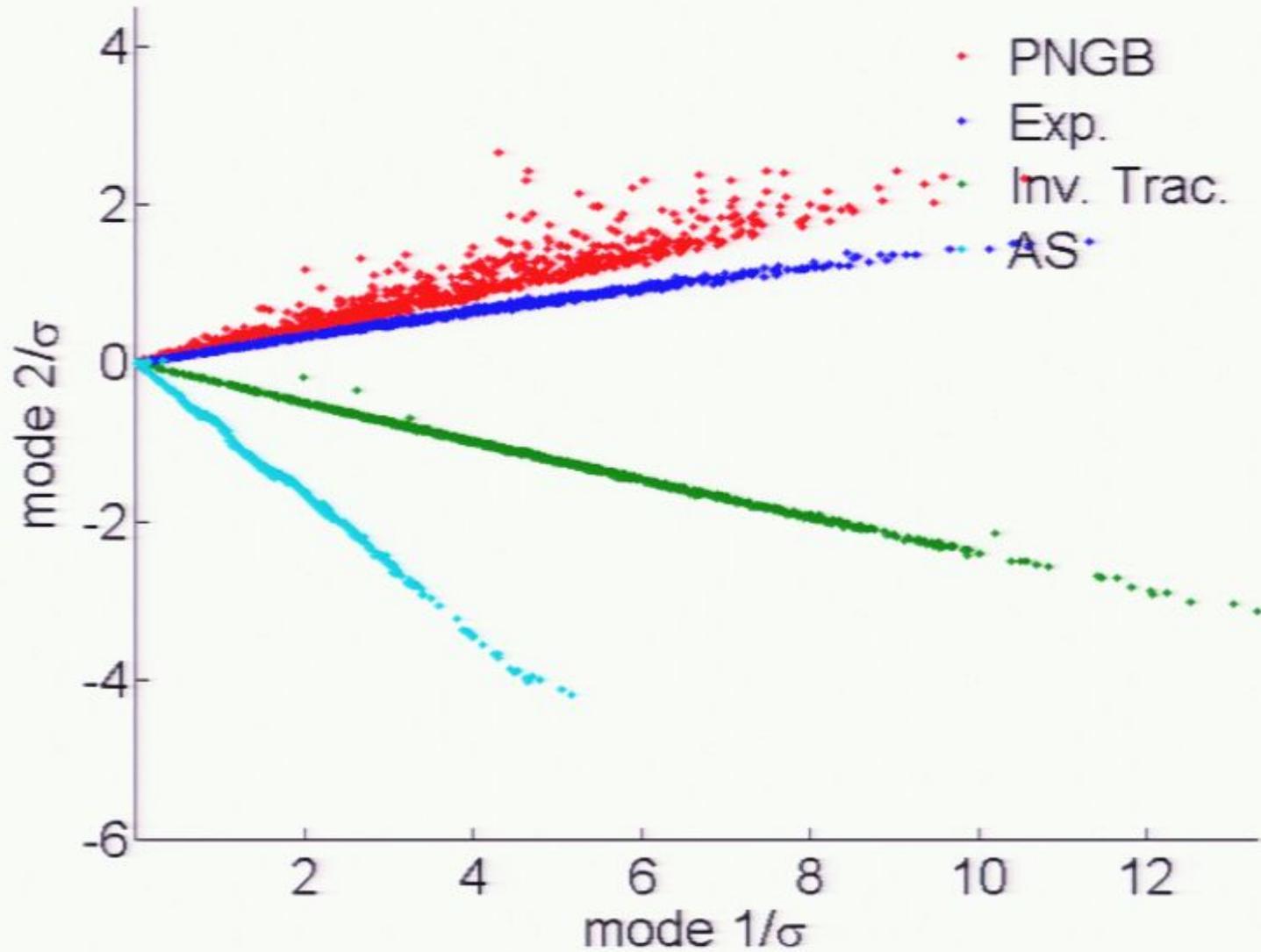
Eigenmodes:



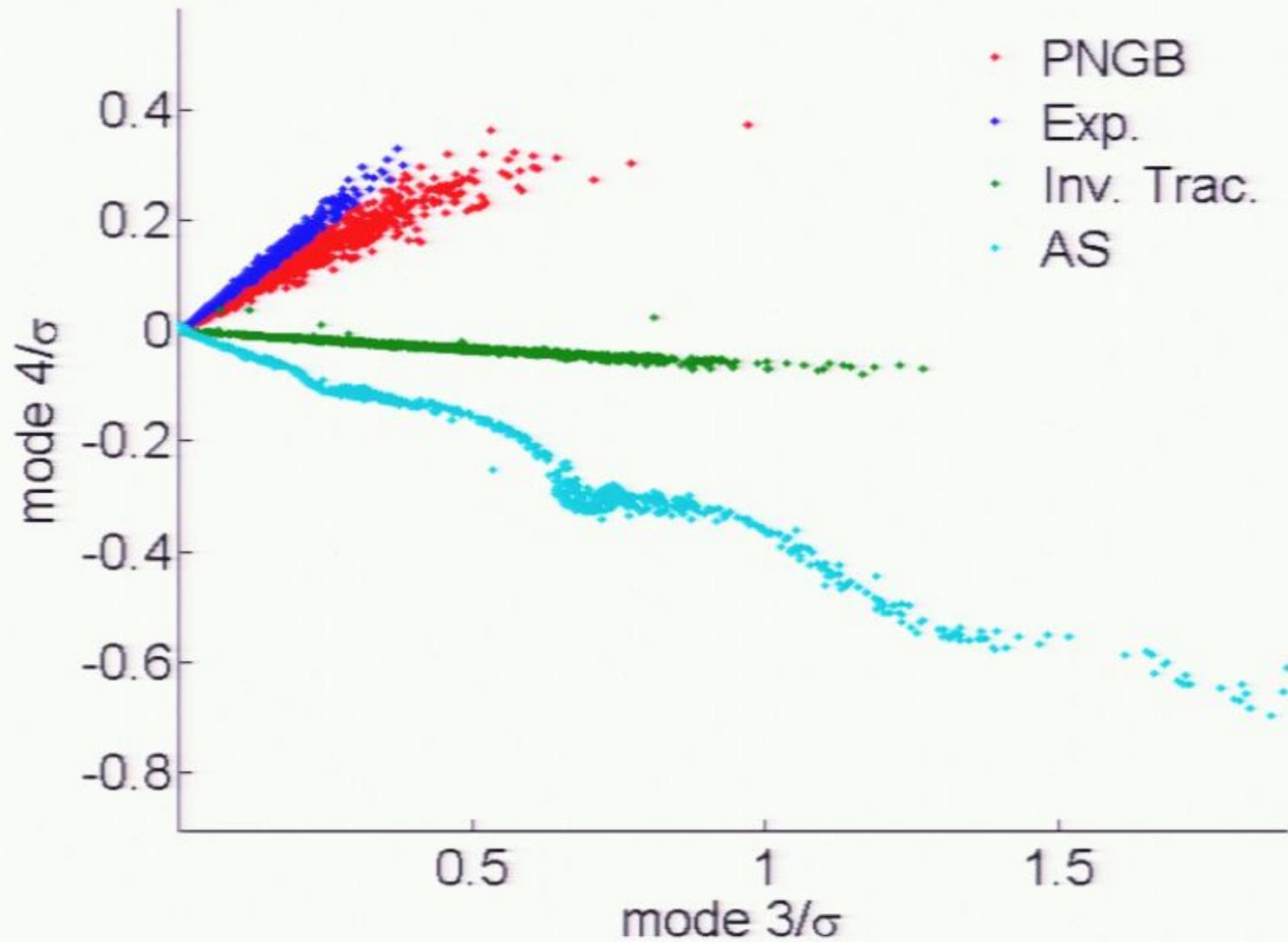
N.B. σ_i
change too

a

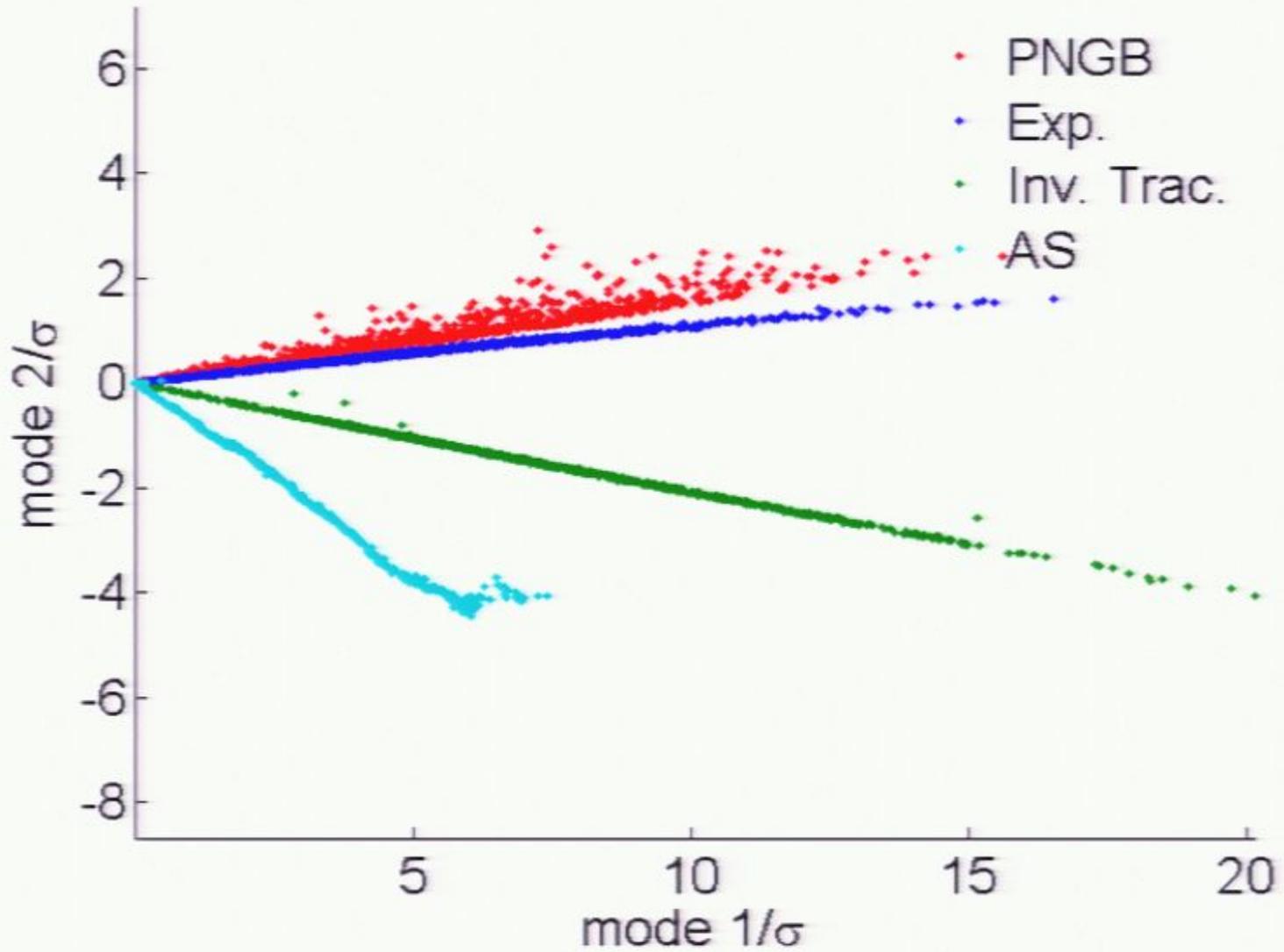
DETF Stage 4 ground [Opt]



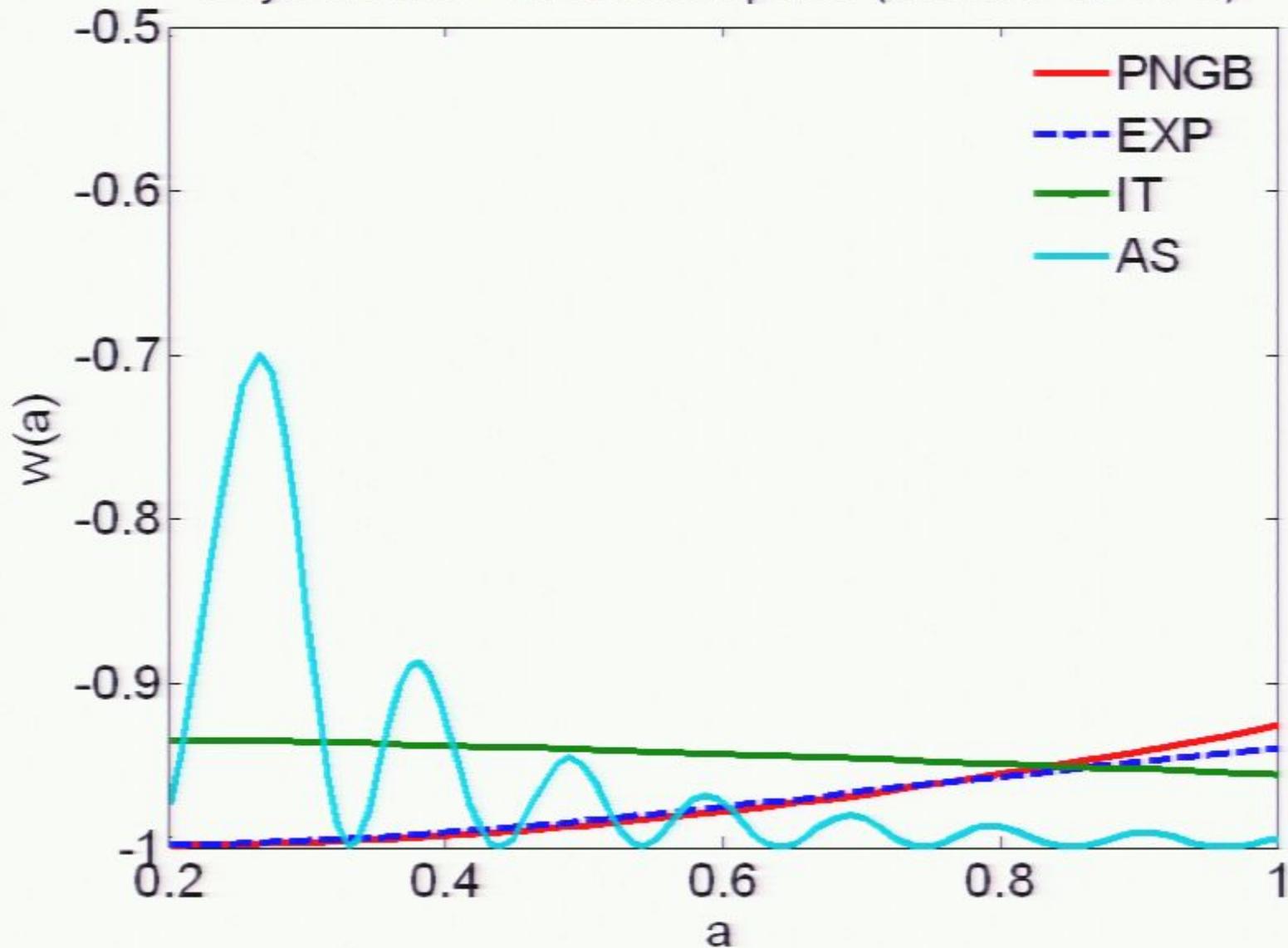
DETF Stage 4 ground [Opt]



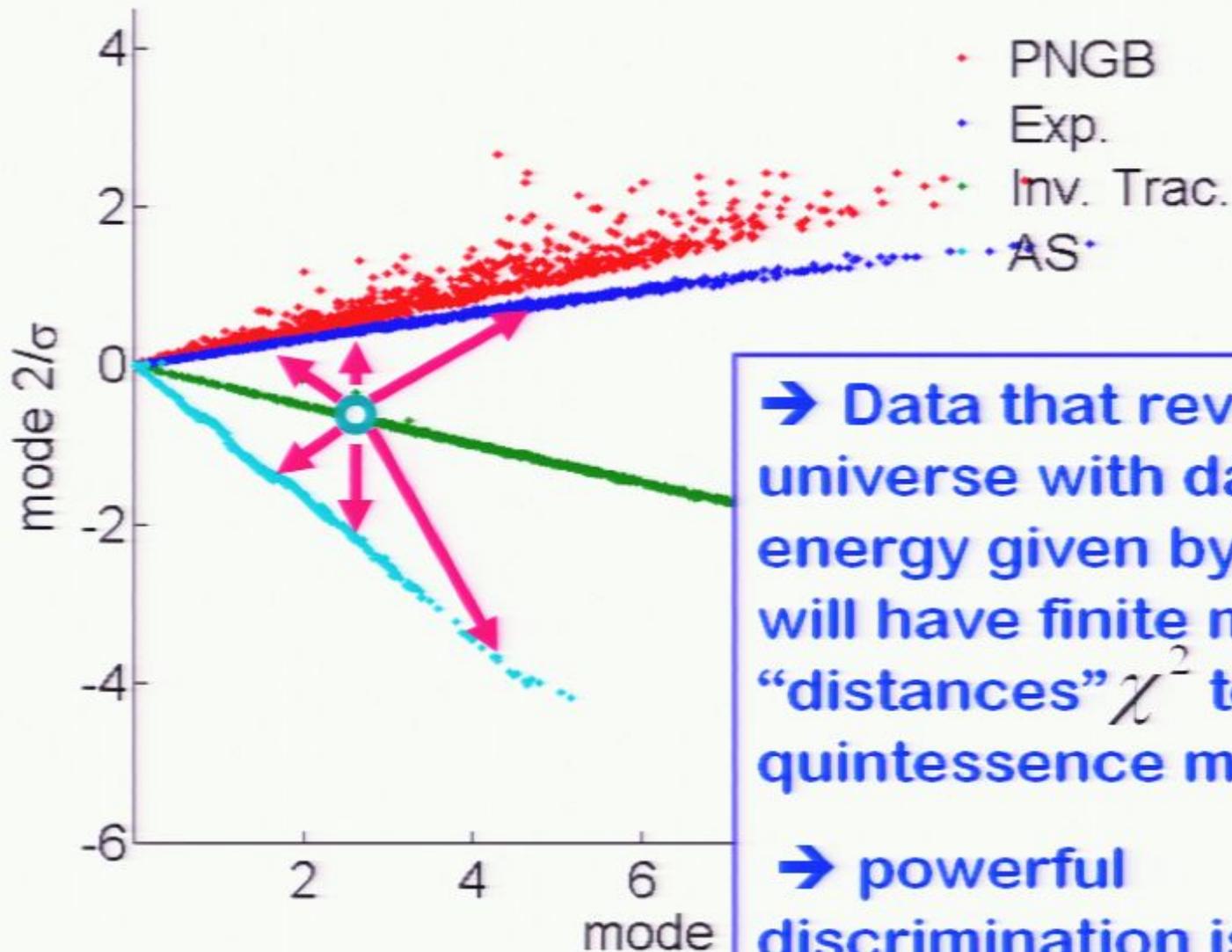
DETF Stage 4 space [Opt]



The different kinds of curves correspond to different “trajectories” in mode space (similar to FT’s)



DETF Stage 4 ground

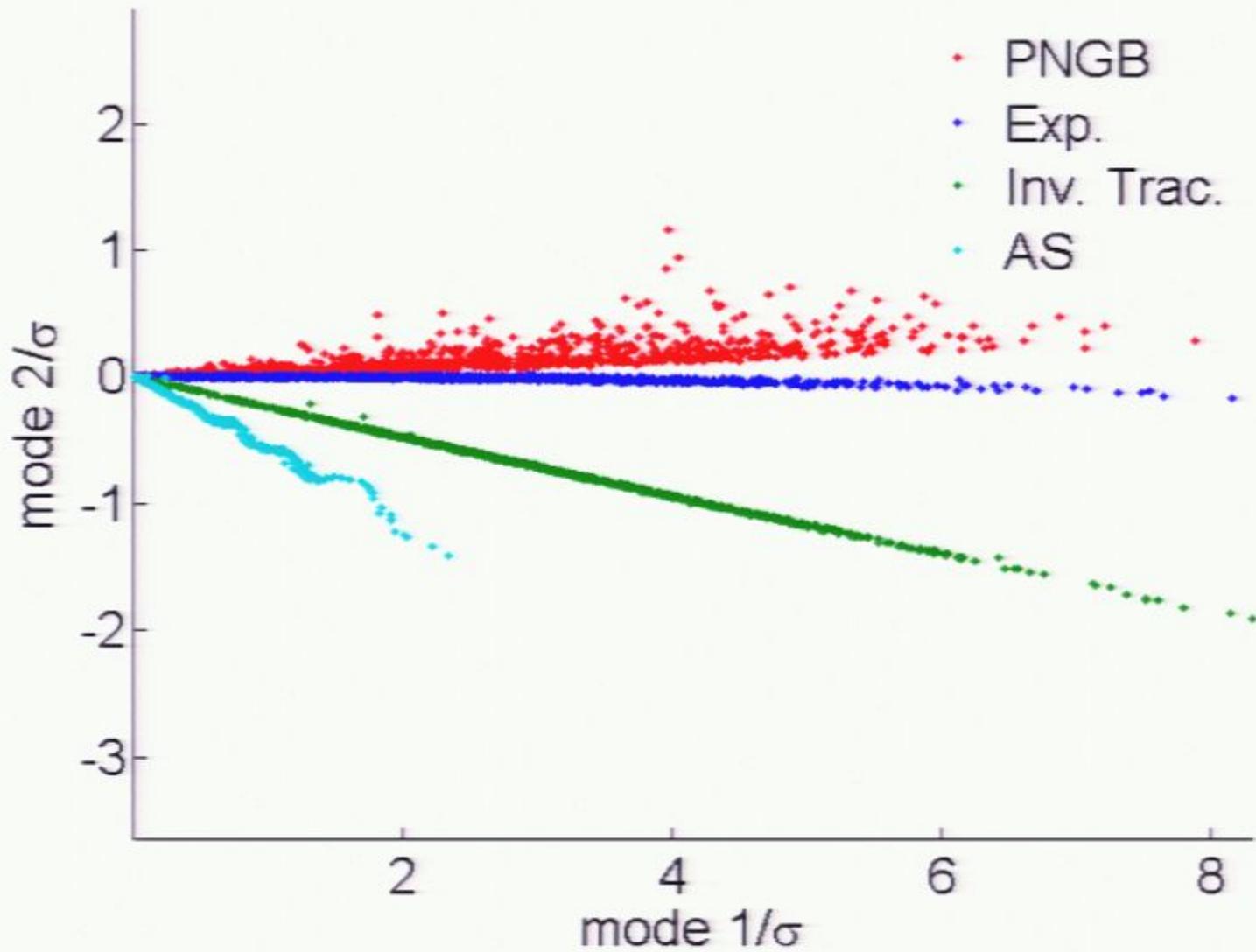


→ Data that reveals a universe with dark energy given by “○” will have finite minimum “distances” χ^2 to other quintessence models

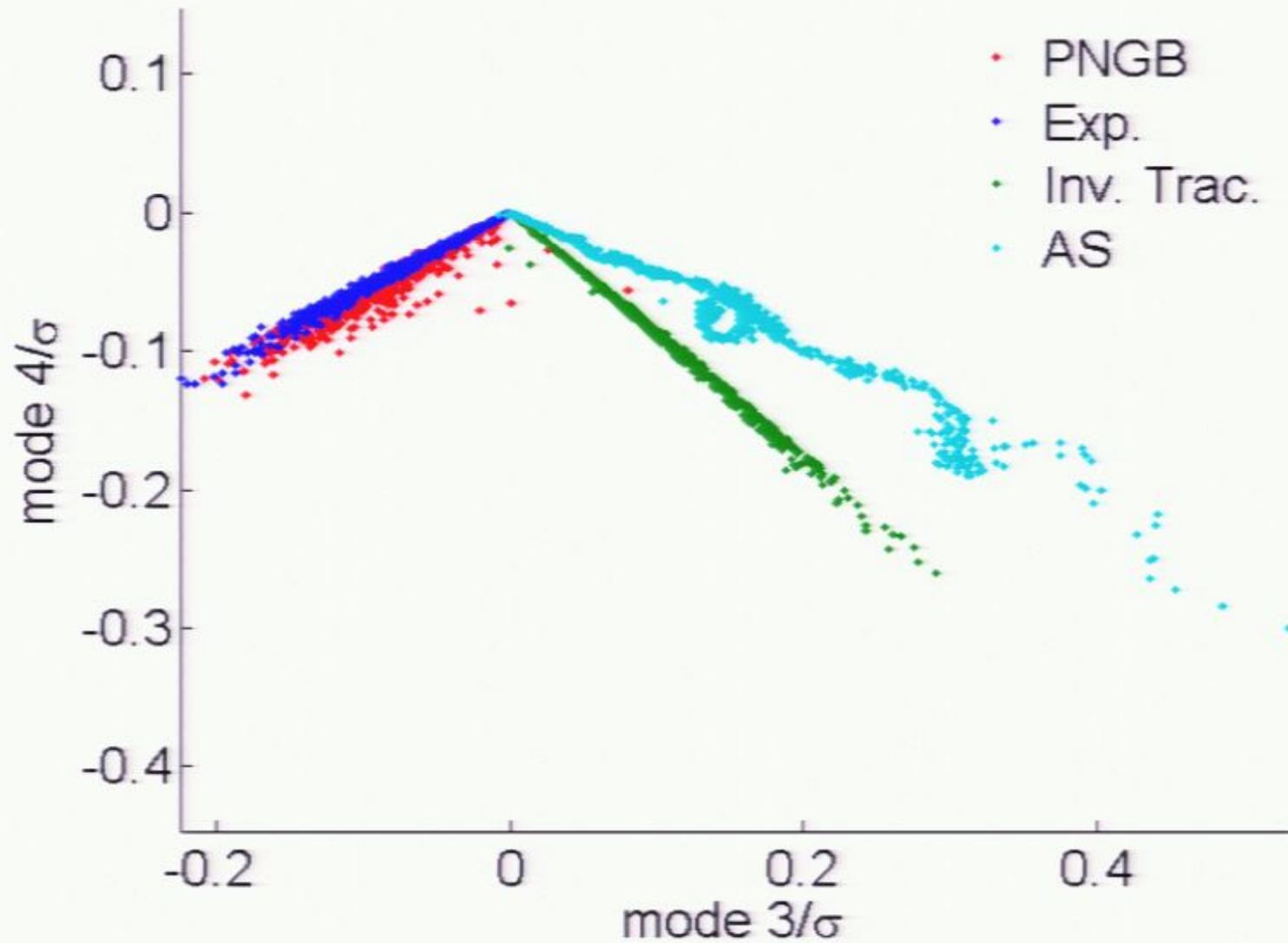
→ powerful discrimination is possible.

Consider discriminating power
of each experiment (→ look at
units on axes)

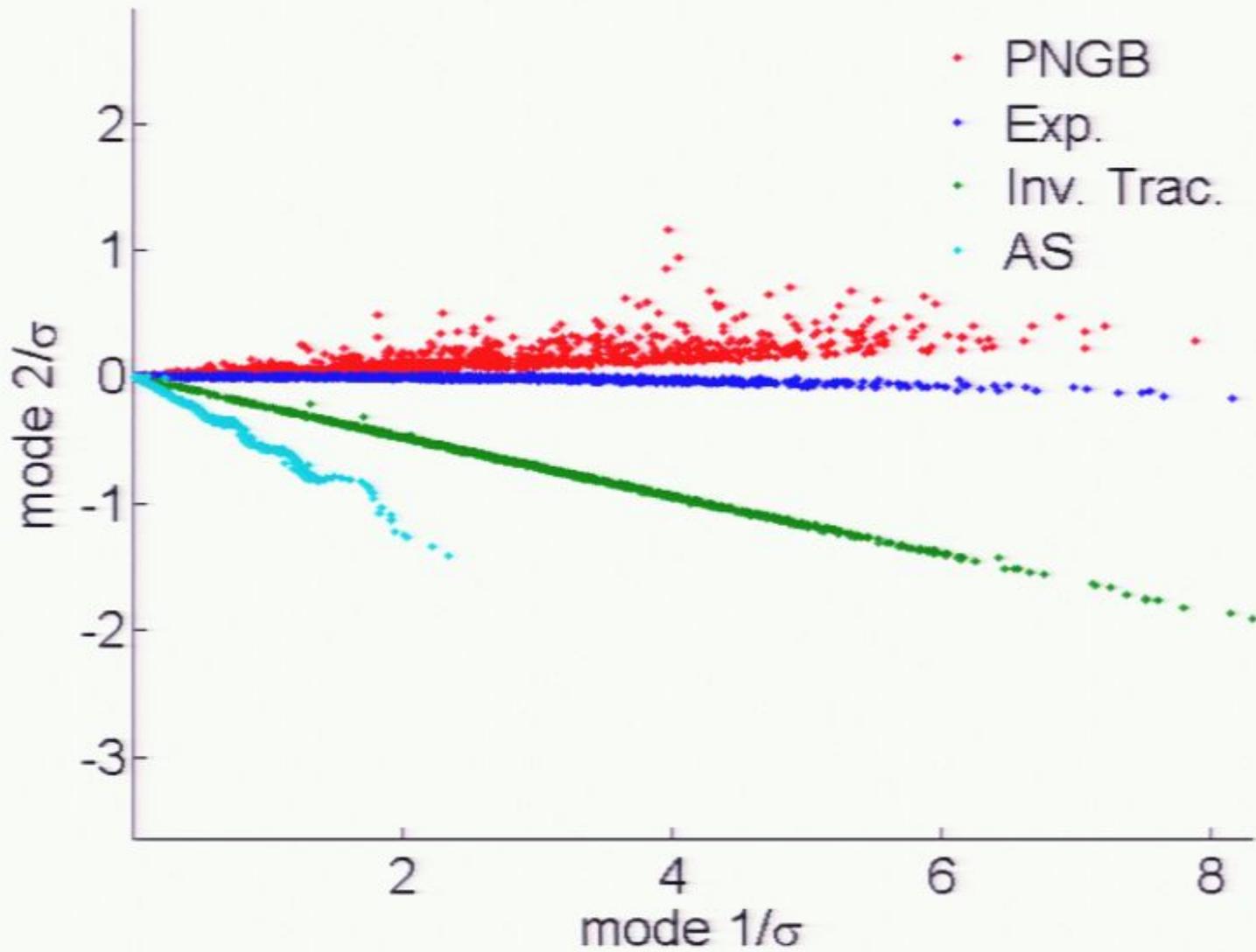
DETF Stage 3 photo [Opt]



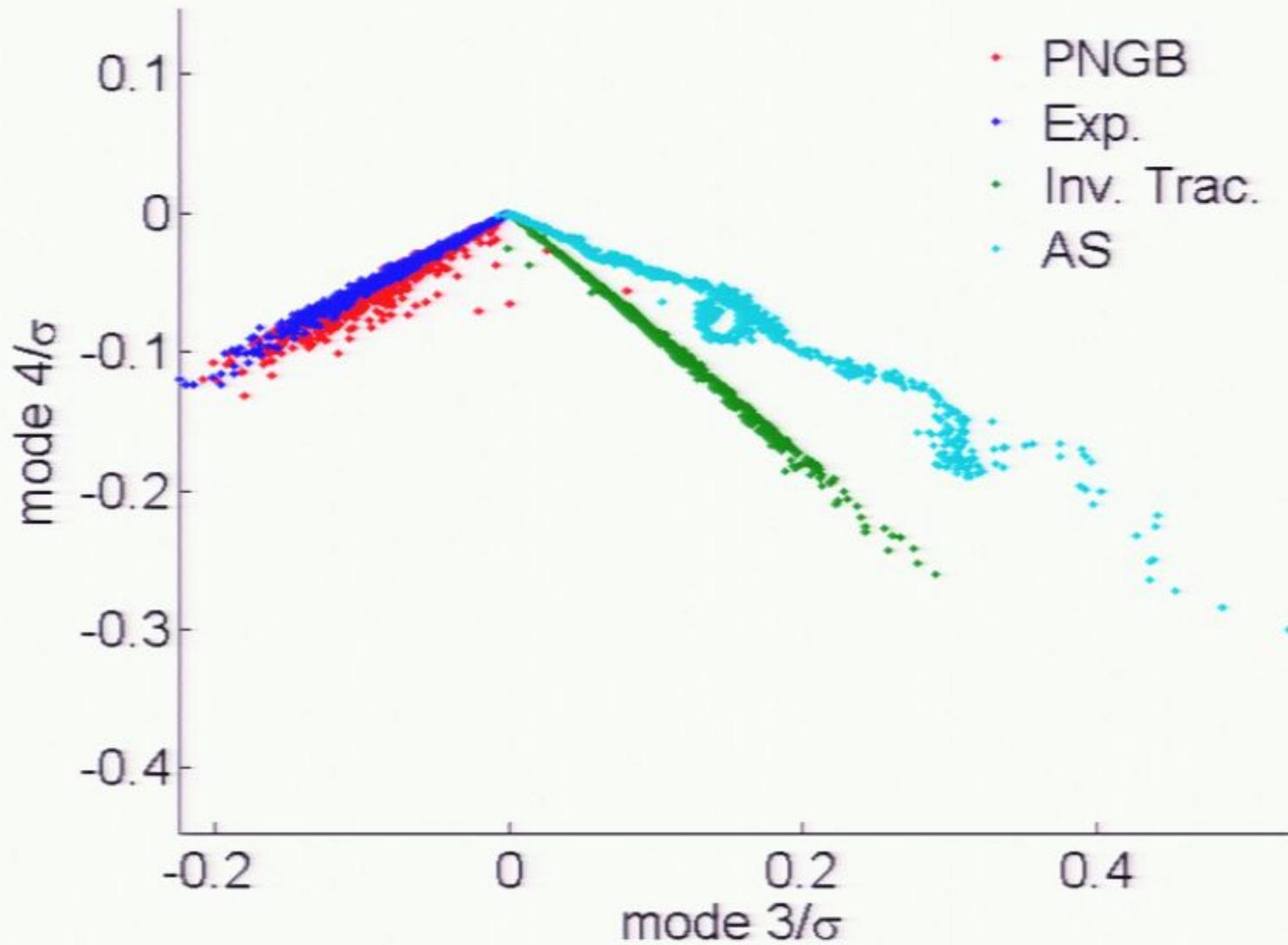
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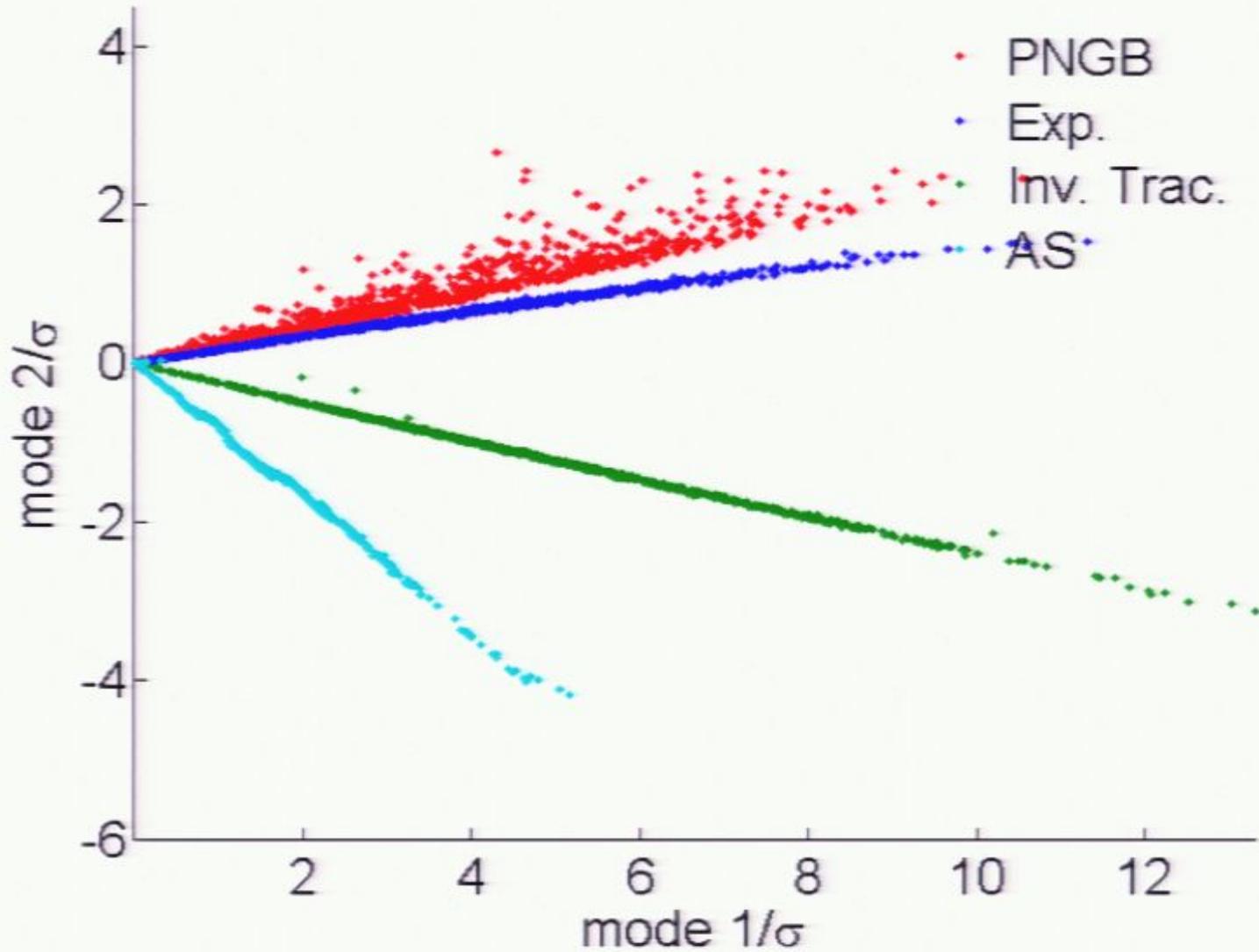
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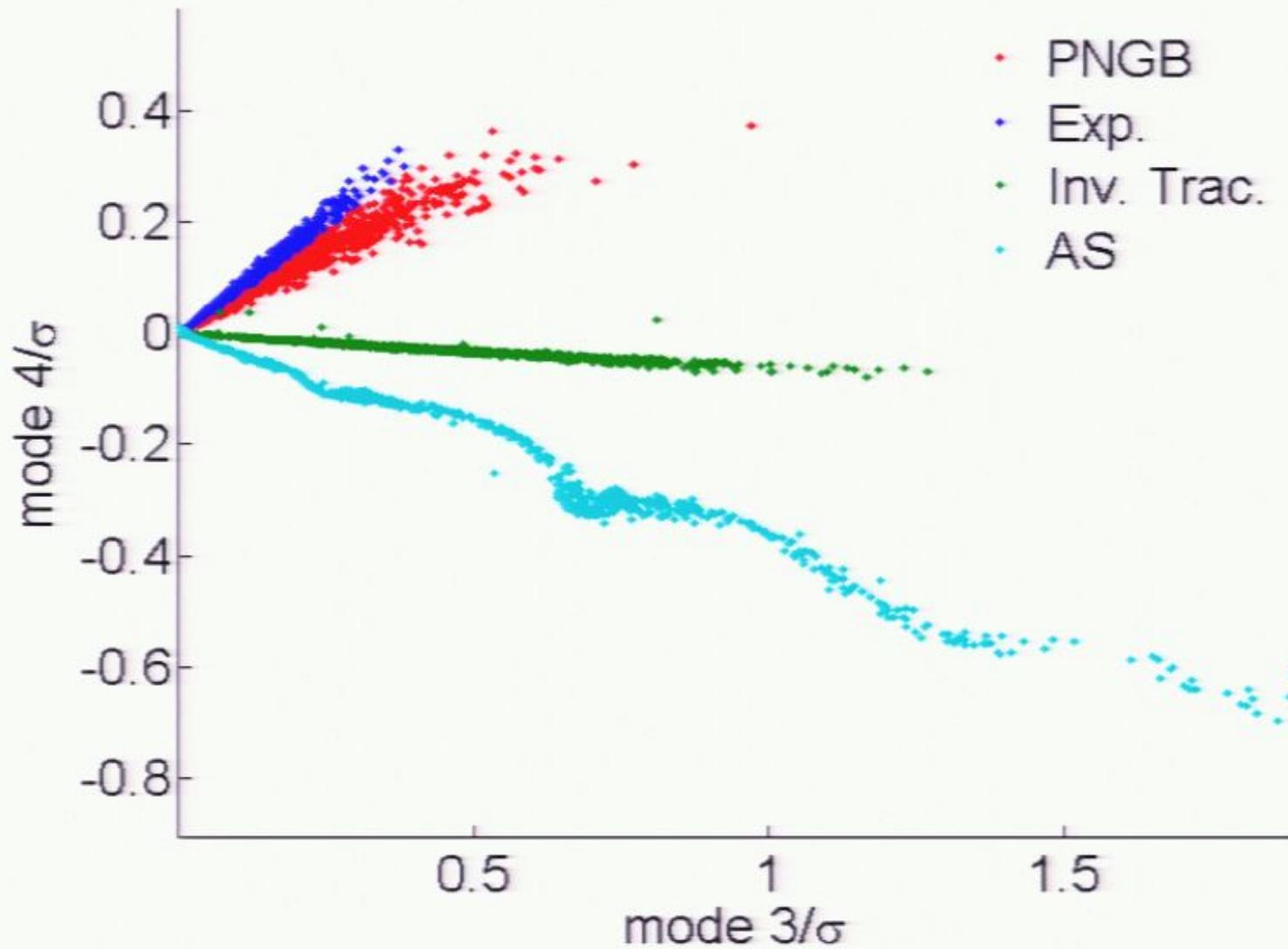
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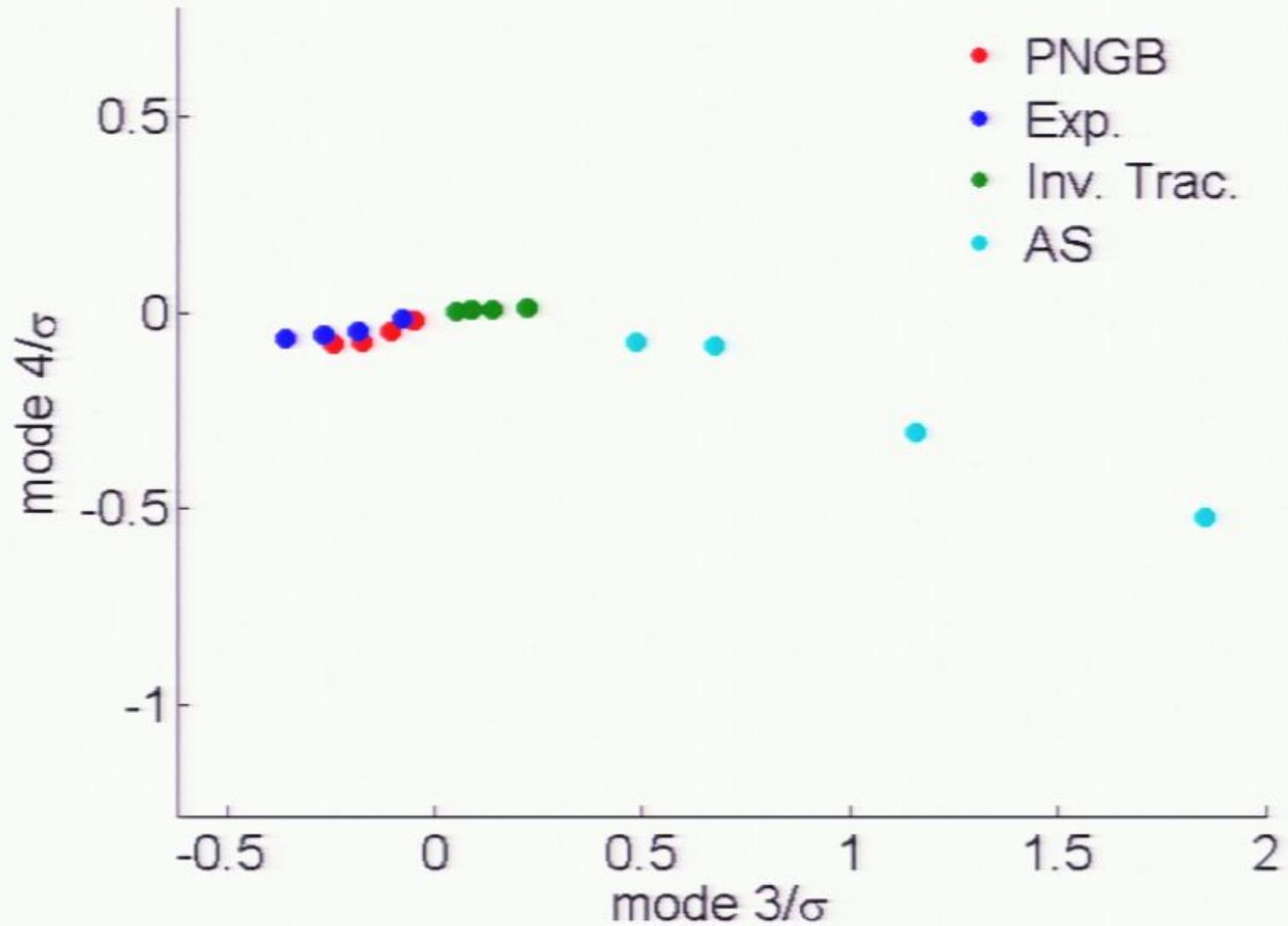
DETF Stage 4 ground [Opt]



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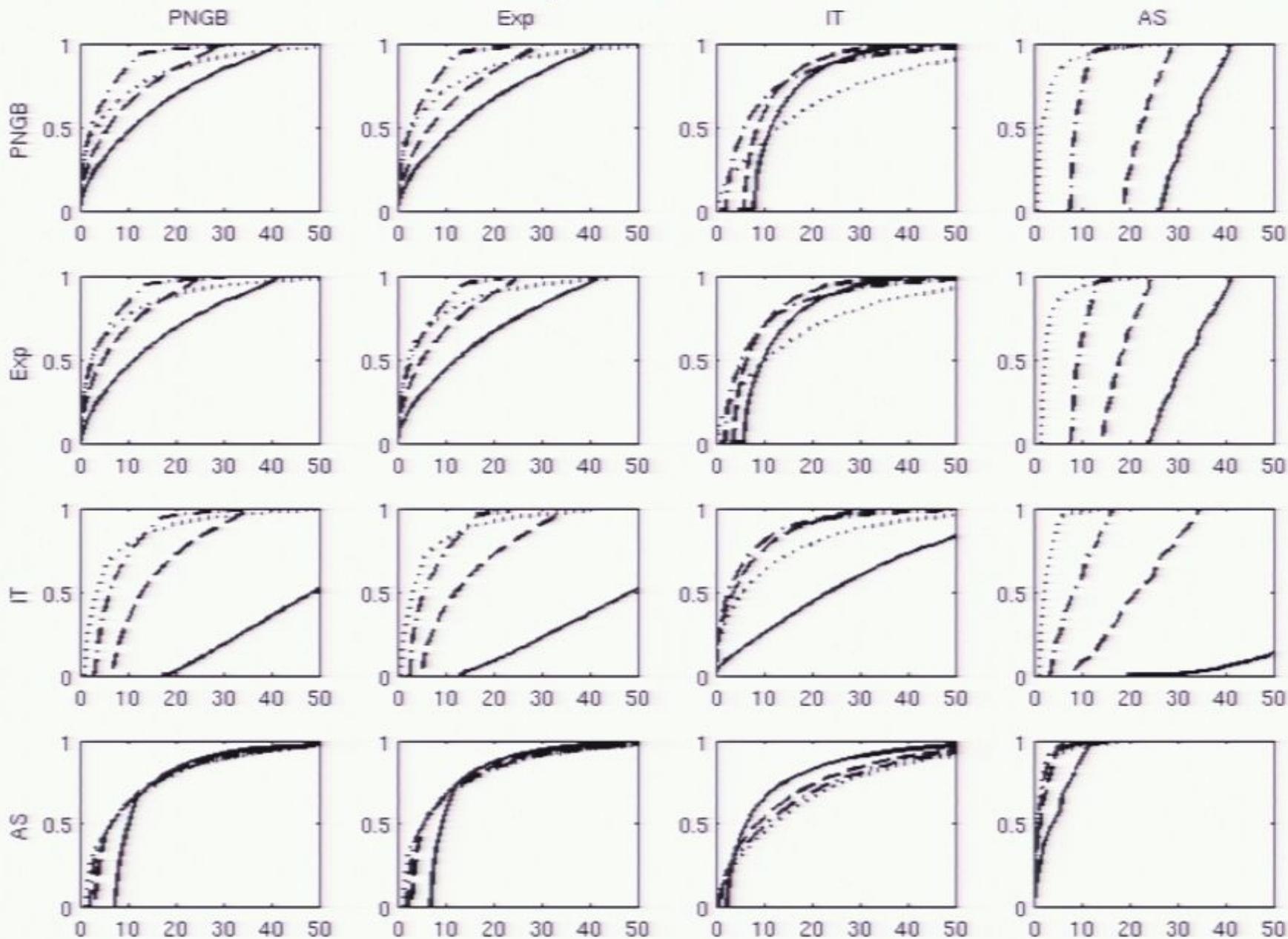
Stage 4 space Test Points



Comparison Model

DETF Stage 4 ground

Test Point Model



DETF Stage 3 photo

A tabulation of χ^2 for each graph where the curve crosses the x-axis (= gap)
 For the three parameters used here,
 95% confidence $\rightarrow \chi^2 = 7.82$,
 99% $\rightarrow \chi^2 = 11.36$.

Light orange > 95% rejection

Dark orange > 99% rejection

Blue: Ignore these because PNGB & Exp hopelessly similar, plus self-comparisons

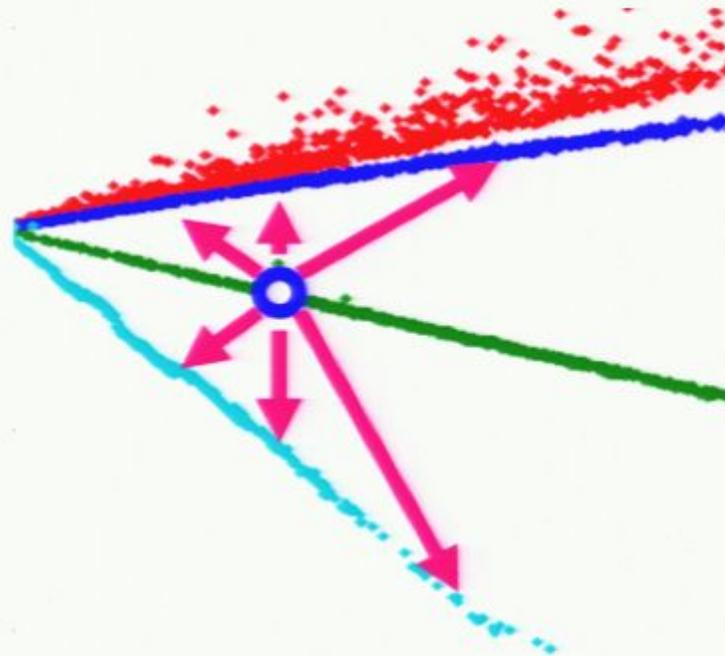
	PNGB	PNGB	Exp	IT	AS
Point 1	0.001	0.001	0.001	0.1	0.2
Point 2	0.002	0.01	0.01	0.5	1.8
Point 3	0.004	0.04	0.04	1.2	6.2
Point 4	0.01	0.04	0.04	1.6	10.0
Exp					
Point 1	0.004	0.001	0.001	0.1	0.4
Point 2	0.01	0.001	0.001	0.4	1.8
Point 3	0.03	0.001	0.001	0.7	4.3
Point 4	0.1	0.01	0.01	1.1	9.1
IT					
Point 1	0.2	0.1	0.1	0.001	0.2
Point 2	0.5	0.4	0.4	0.0004	0.7
Point 3	1.0	0.7	0.7	0.001	3.3
Point 4	2.7	1.8	1.8	0.01	16.4
AS					
Point 1	0.1	0.1	0.1	0.1	0.0001
Point 2	0.2	0.1	0.1	0.1	0.0001
Point 3	0.2	0.2	0.2	0.1	0.0002
Point 4	7.4	7.0	7.0	2.6	0.001

Comments on model discrimination

- Principle component $w(a)$ “modes” offer a space in which straightforward tests of discriminating power can be made.
- The DETF Stage 4 data is approaching the threshold of resolving the structure that our scalar field models form in the mode space.

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Followup questions:

- In what ways might the choice of DE parameters have skewed the DETF results?
- What impact can these data sets have on specific DE models (vs abstract parameters)?
- To what extent can these data sets deliver discriminating power between specific DE models?
- How is the DoE/ESA/NASA Science Working Group looking at these questions?

A:

- DETF Stage 3: Poor
- DETF Stage 4: Marginal... Excellent within reach

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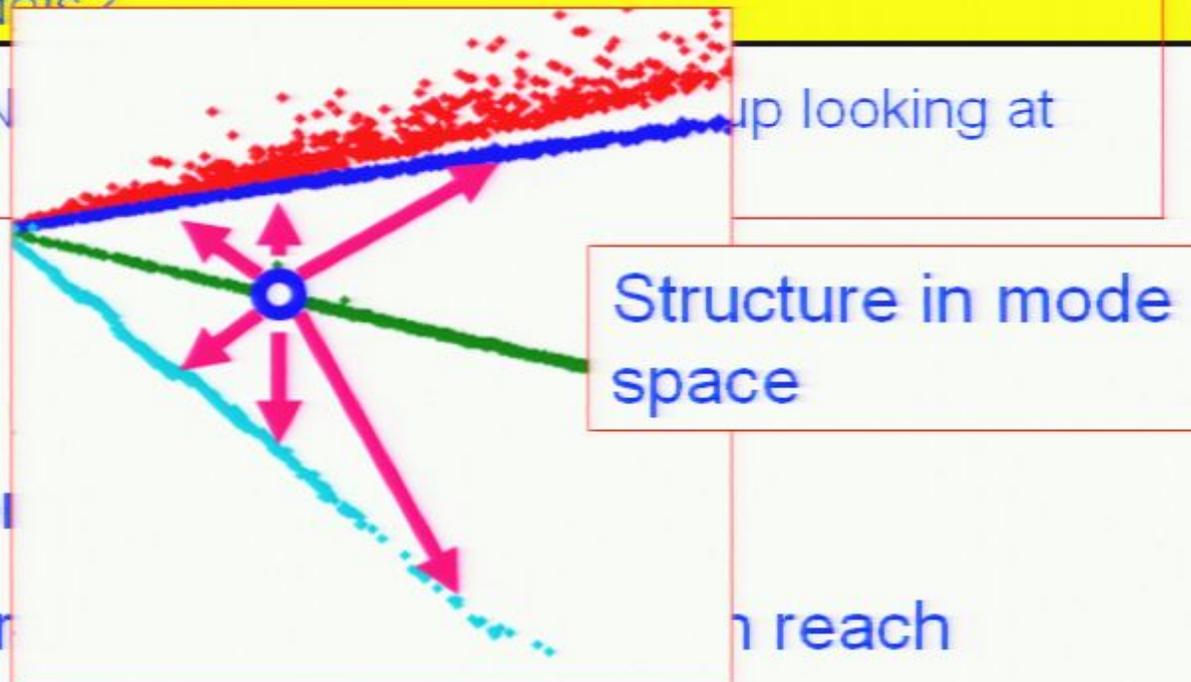
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- DETF Stage 4: Mar



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DoE/ESA/NASA JDEM Science Working Group

- Update agencies on figures of merit issues
- formed Summer 08
- finished ~now (moving on to SCG)
- Use w -eigenmodes to get more complete picture
- also quantify deviations from Einstein gravity
- For today: Something we learned about normalizing modes

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

$$\Delta\vec{w} = \sum_i c_i \vec{f}_i$$

The c_i are independently measured quantities with errors σ_i

Define

$$\vec{f}_i^D \equiv \vec{f}_i / \sqrt{\Delta a}$$

which obey continuum normalization:

$$\sum f_i^D(k) f_j^D(k) \Delta a = \delta_{ij}$$

then

$$\Delta\vec{w} = \sum_i c_i^D \vec{f}_i^D$$

where

$$c_i^D \equiv c_i \times \sqrt{\Delta a}$$

Q: Why?

A: For lower modes, f_j^D has typical *grid independent* “height” $O(1)$, so one can more directly relate values of $\sigma_i^D \equiv \sigma_i \times \sqrt{\Delta a}$ to one’s thinking (priors) on $\Delta \vec{w}$

$$\Delta \vec{w} = \sum_i c_i \vec{f}_i = \sum_i c_i^D \vec{f}_i^D$$

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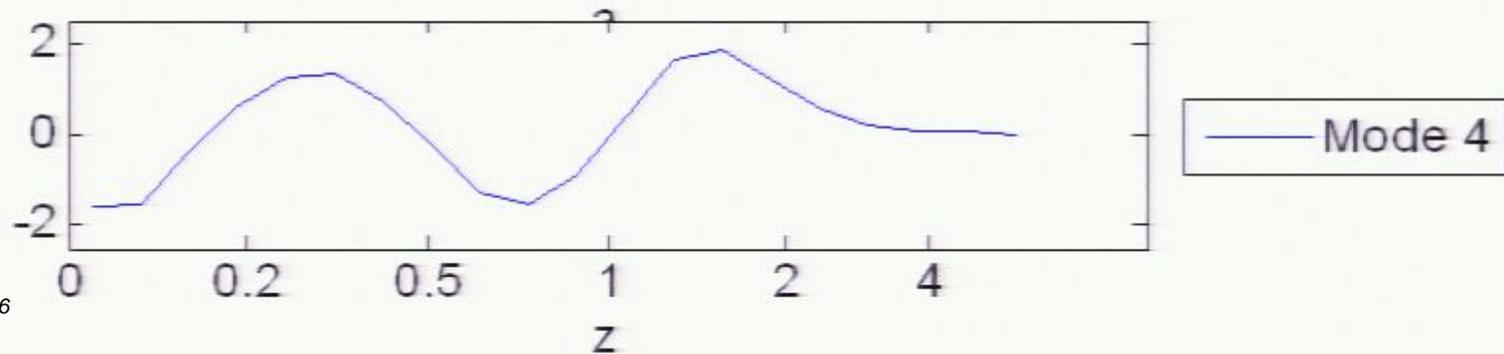
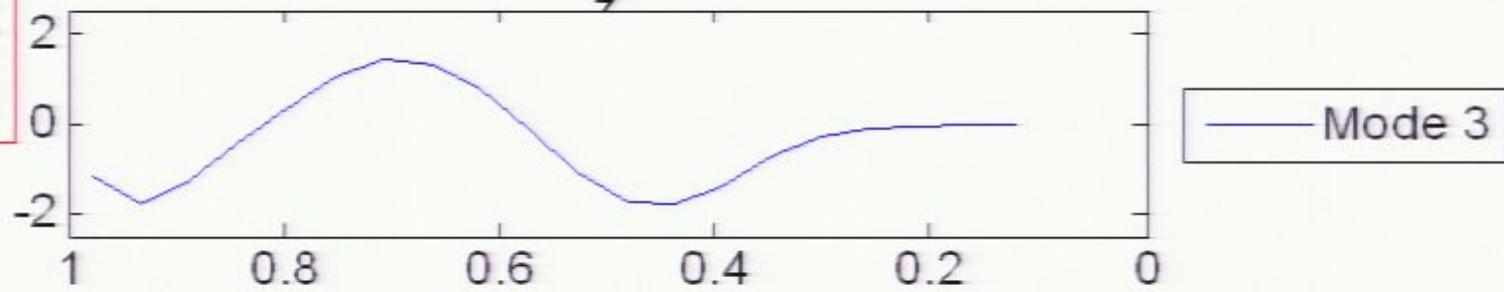
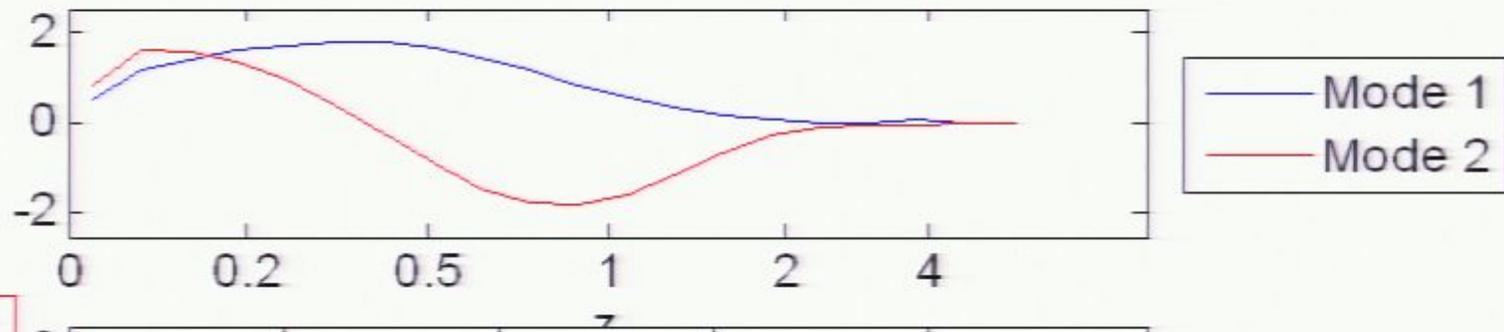
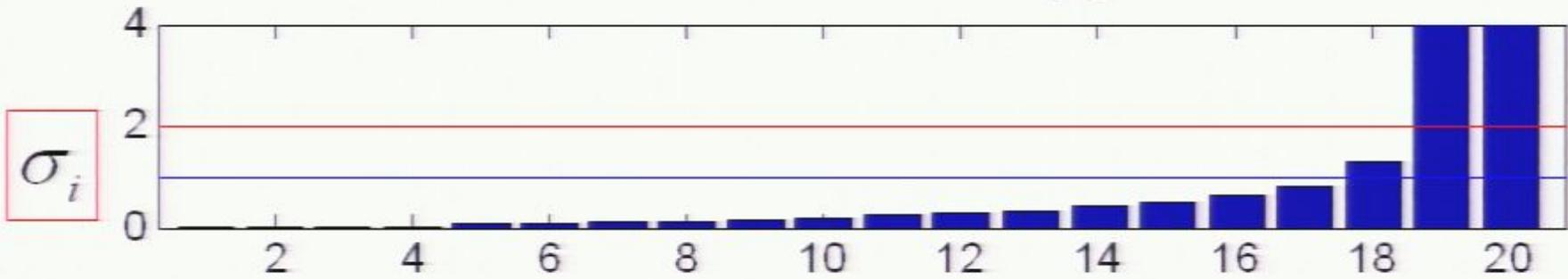
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DETF= Stage 4 Space Opt All $f_{k=6} = 1, Pr = 0$



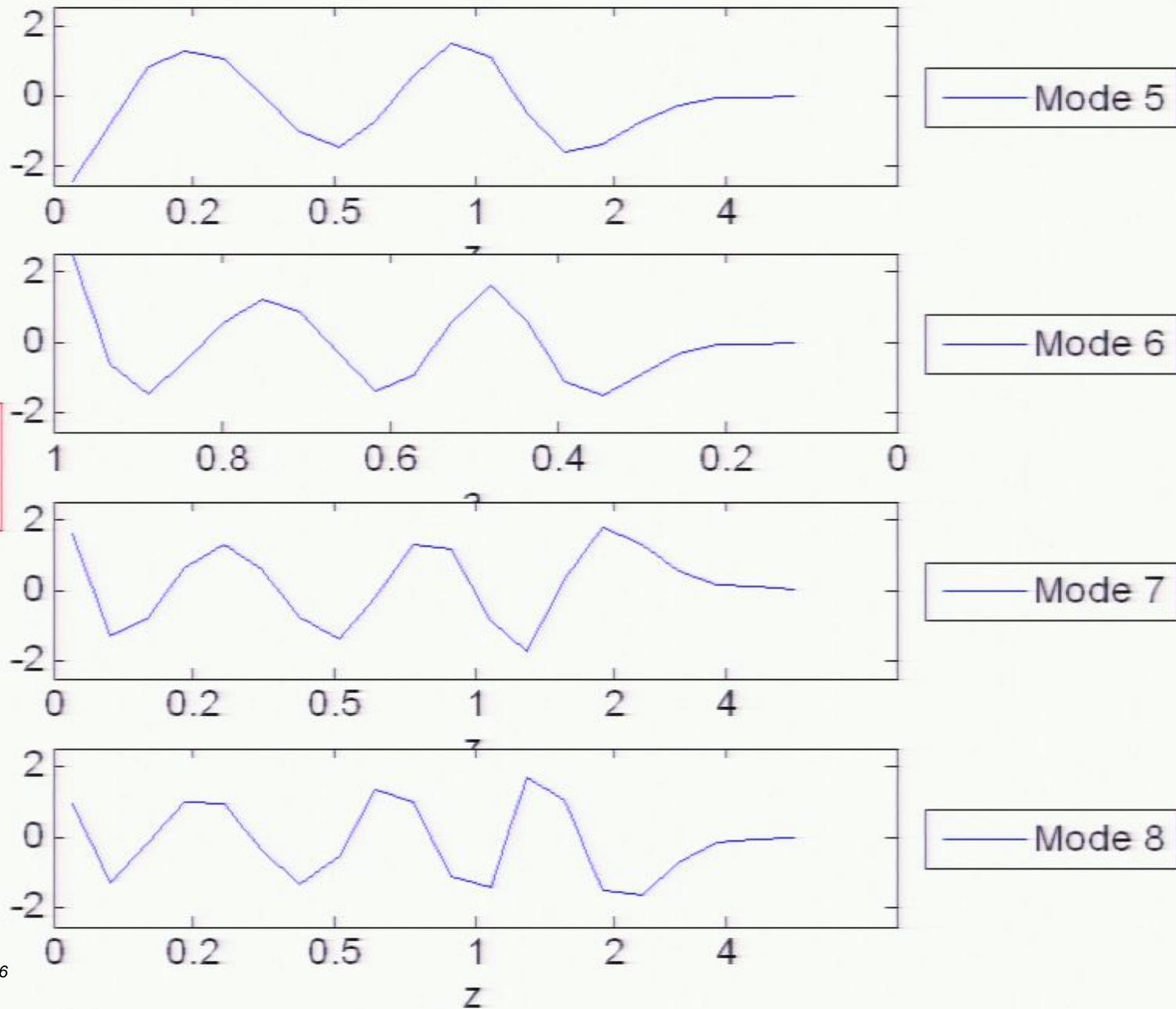
Principle Axes

\vec{f}_i

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$$\vec{f}_i$$

DETF= Stage 4 Space Opt All $f_{k=6} = 1, Pr = 0$



Upshot: More modes are interesting (“well measured” in a grid invariant sense) than previously thought.

Summary

→ In what ways might the choice of DE parameters have skewed the DETF results?

A: Only by an overall (possibly important) rescaling

→ What impact can these data sets have on specific DE models (vs abstract parameters)?

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Interesting contribution to discussion of Stage 4 (if you believe scalar field modes)

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