



Improvement of a correction algorithm for laser-misalignment effects in TS density profiles at the W7-X stellarator



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Outline



Background

- unphysical density profiles — mainly caused by laser movement

Developed Correction Algorithm

- Identify “calibration cell” with Autoencoder
- Retrospective Correction

Improvement & Results

- New way to identify “calibration cell”

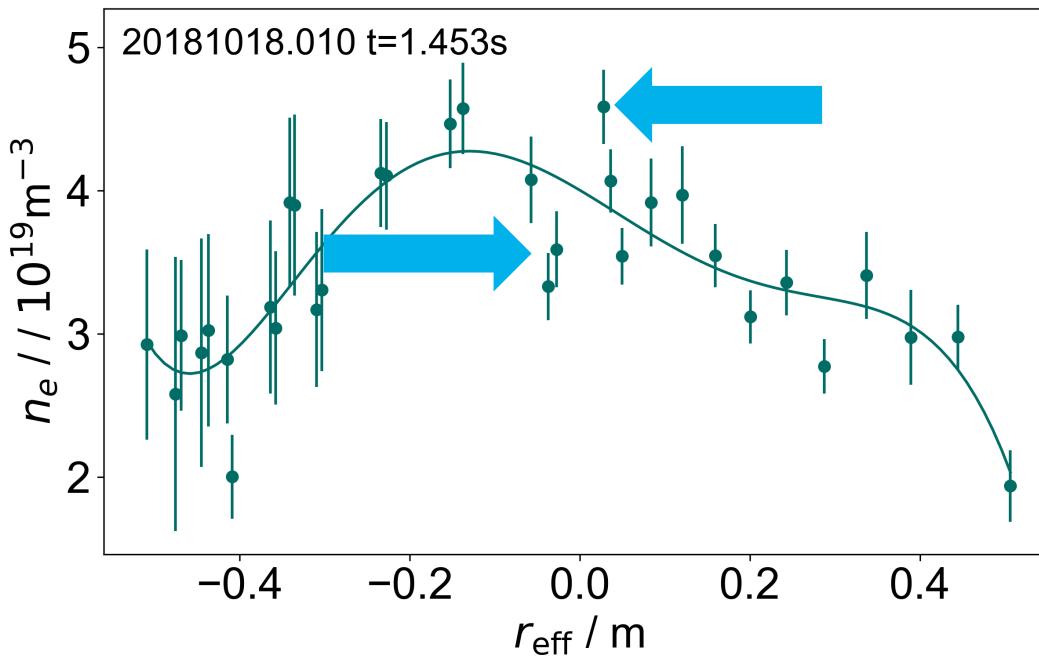
Summary and Outlook

Background

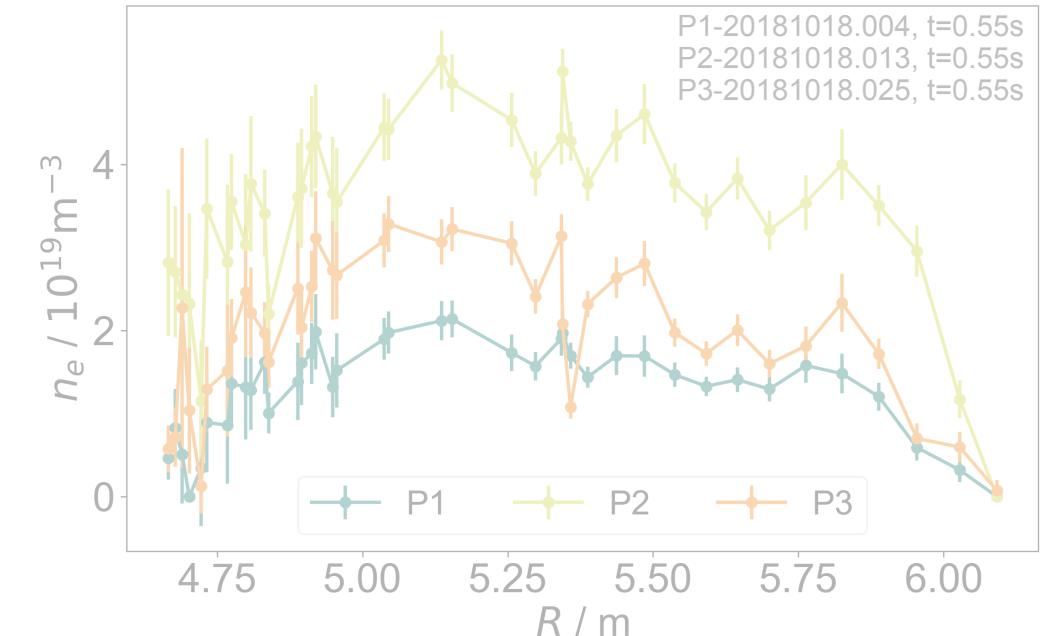


Distorted TS density profiles

- not smooth or symmetric over r_{eff}



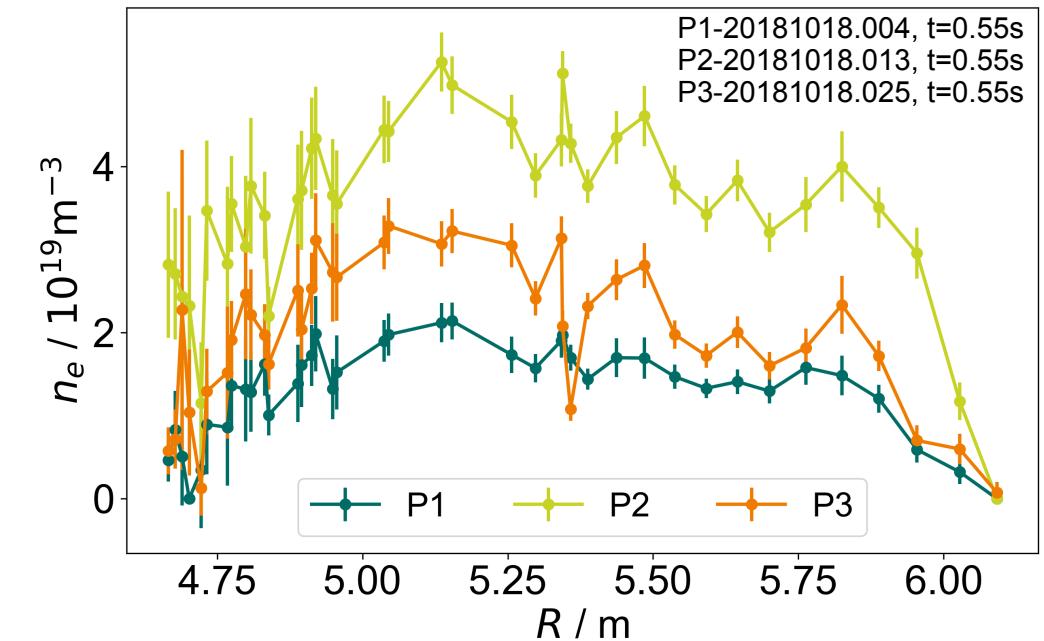
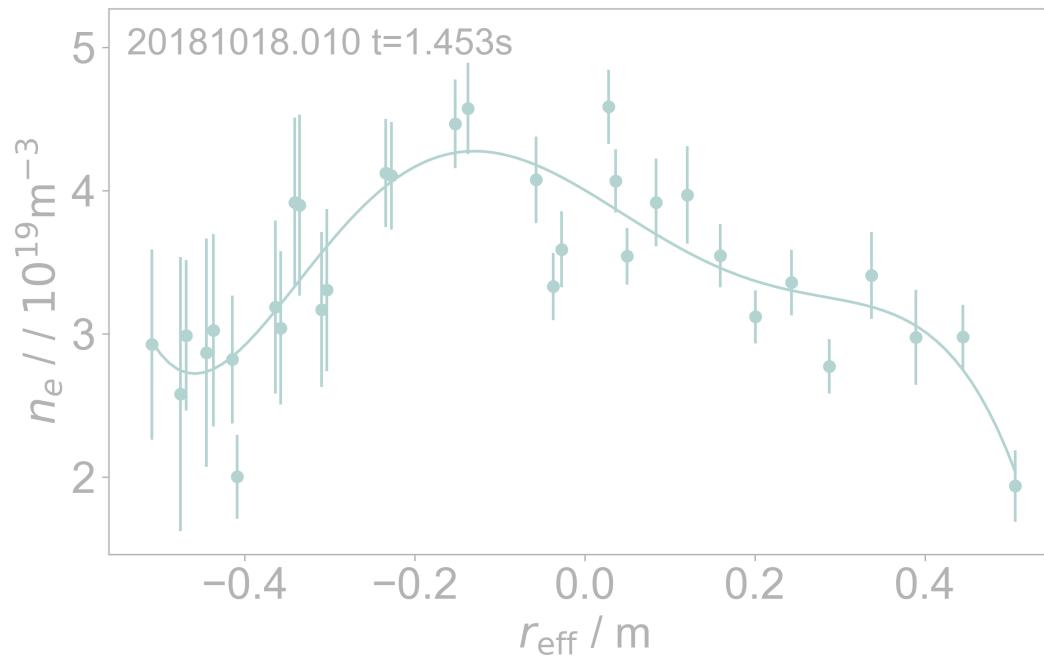
- Measured profiles for similar plasma parameters are very different



[slide courtesy J. Frank]

Distorted TS density profiles

- not smooth or symmetric over r_{eff}
- Profiles for similar plasma parameters are very different



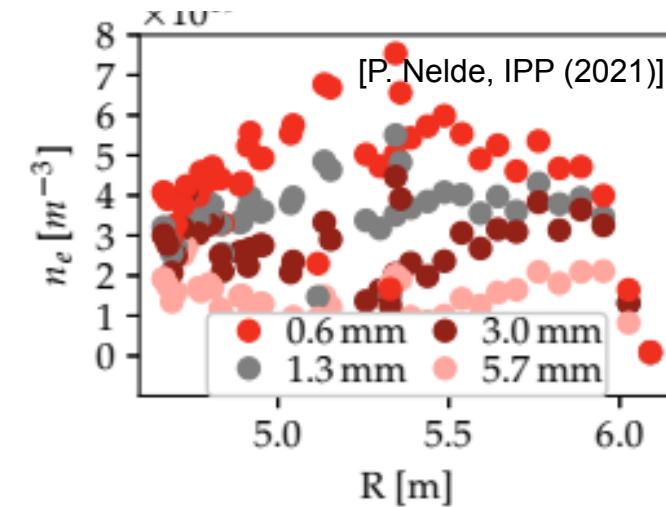
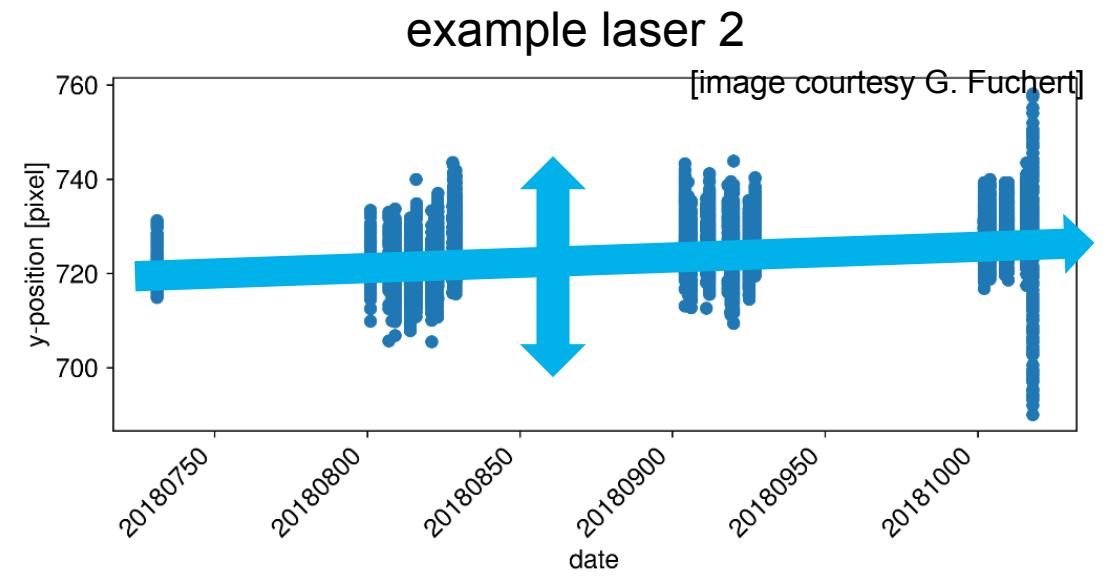
[slide courtesy J. Frank]

Error Sources — Laser Movement



Master's thesis of Philips Nelde:

- biggest error source: **unstable laser alignment (drift & vibration)**
→ variety of different profile shapes
- Even small drifts can reduce the measured density by 90 %

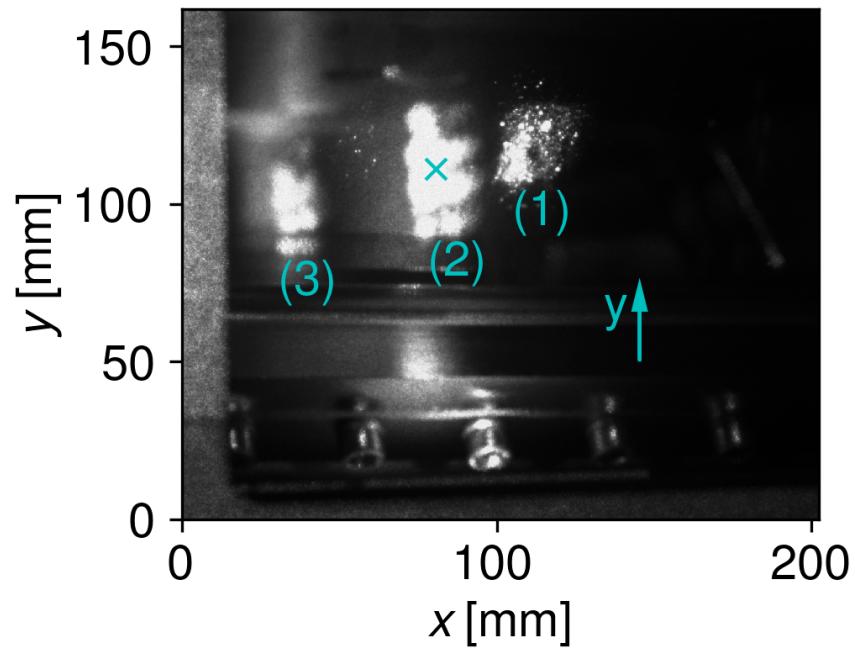


[slide courtesy J. Frank]

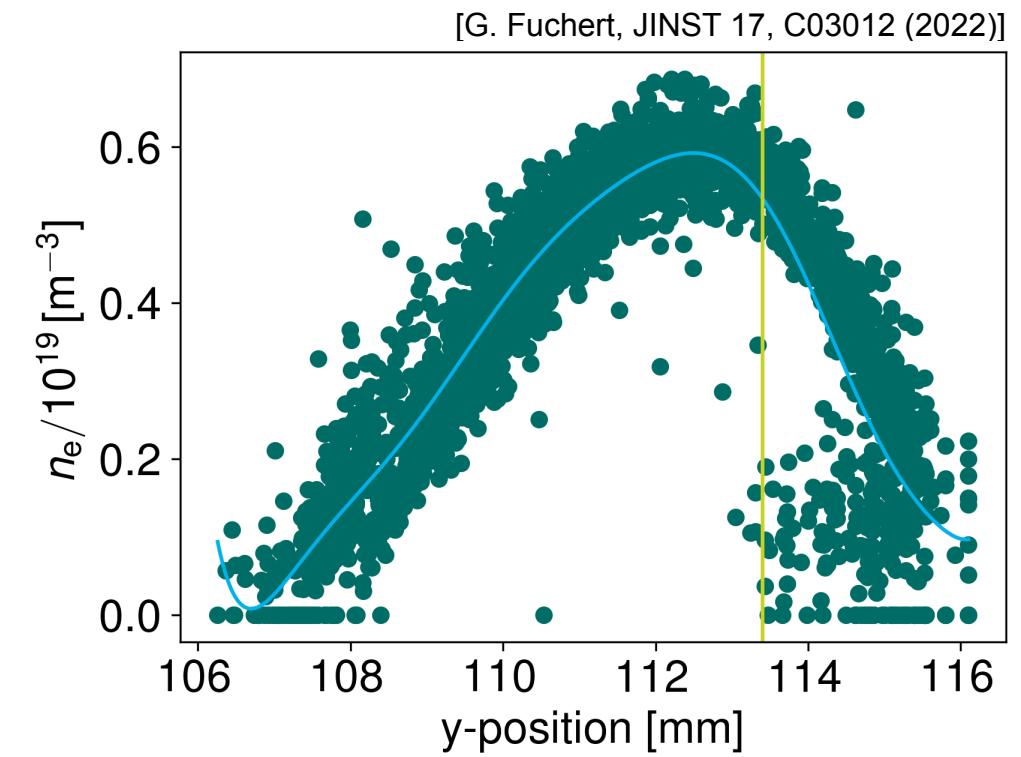
Correction by Camera

Proof of concept:

- Monitoring of entrance laser position with a camera
- Position dependence from alignment scan - each volume has now a relation between position on camera and measured density



Monitored position of the laser spot on the entrance window (1) and two reflections (2 and 3)



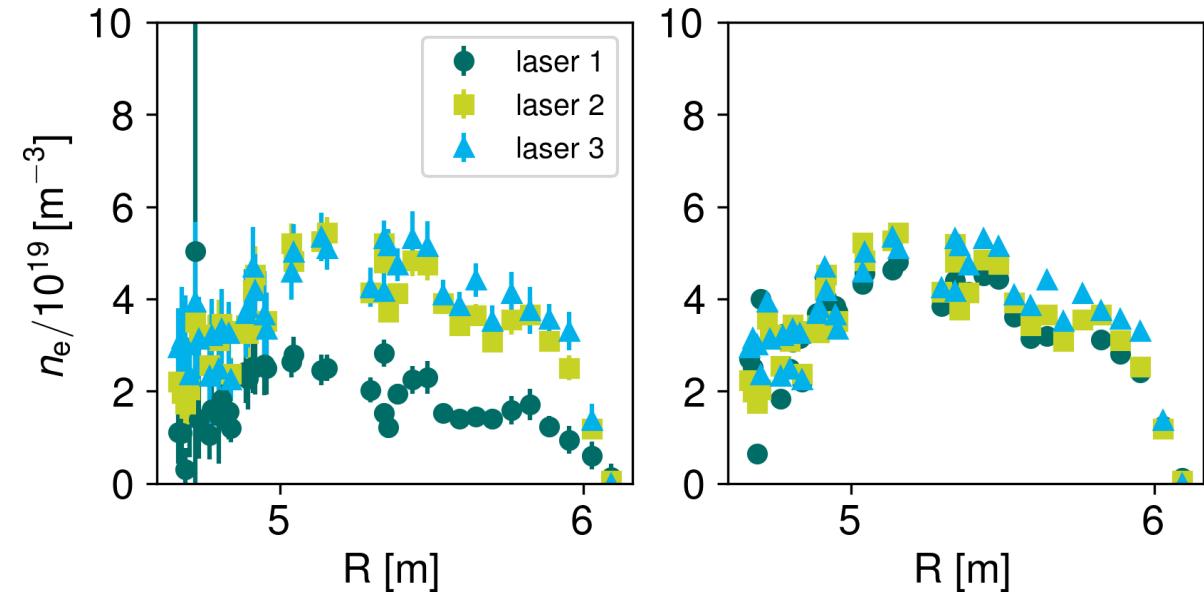
[slide courtesy J. Frank]

Correction by camera

- Correction of profiles influenced by intentional laser misalignment **successful**

Current OP:

- 2 cameras monitoring the laser entrance & exit
→ all degrees of freedom should be sufficiently covered



... but correction of OP 1.2b data is not possible since there was no camera observation of laser exit

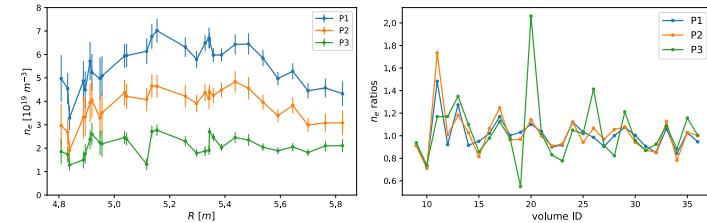
[slide courtesy J. Frank]

Developed Correction Algorithm

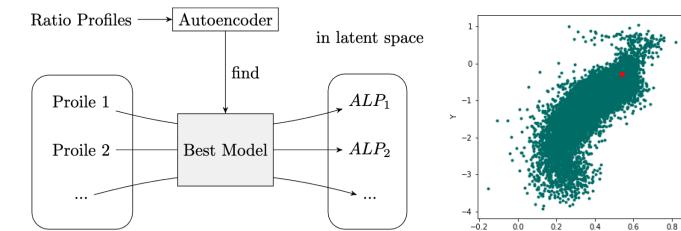
Procedure of the Correction Algorithm

Assume: Raman calibration performed correctly

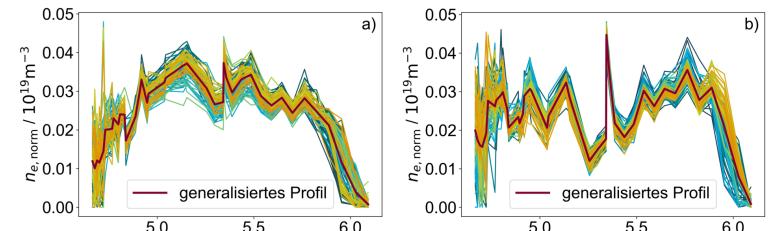
1. Prepare Data



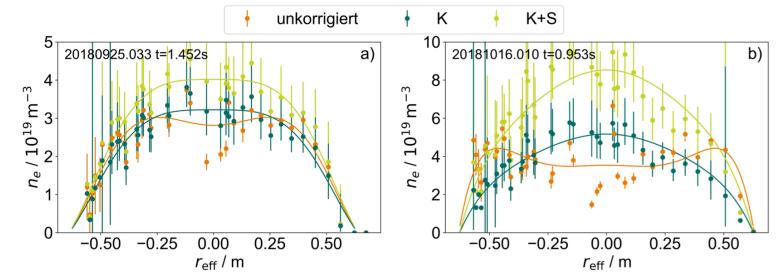
2. Learn “laser position” with *Autoencoder*



3. Identify *the calibration profile* through training result



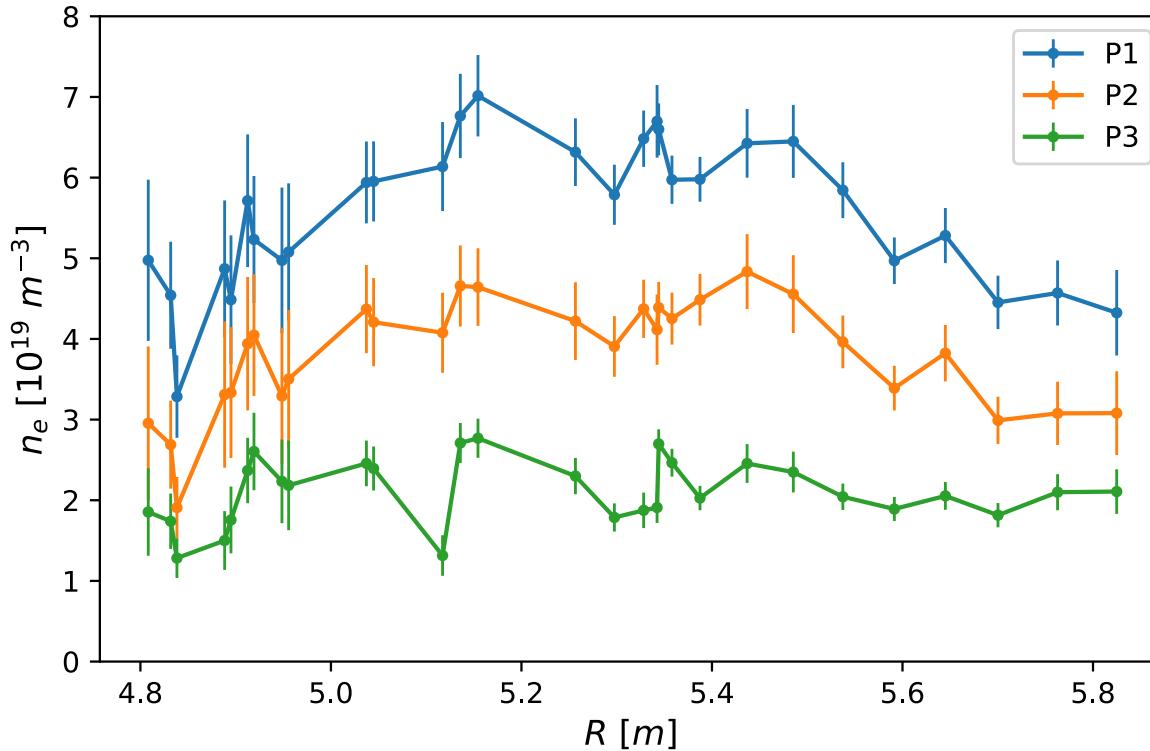
4. Correction



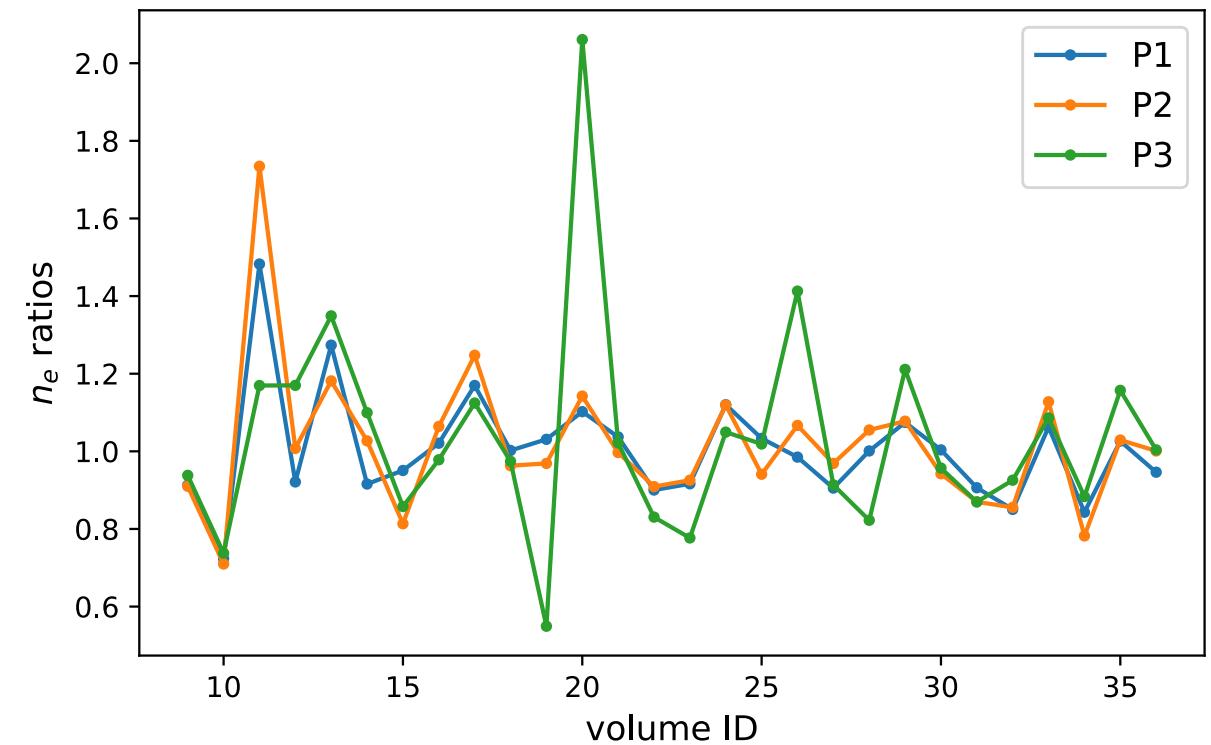
1. Prepare Data for Autoencoder

Ratio Profiles (OP 1.2b V8&10 Sep & Oct): Ratios of n_e measured in adjacent scattering volumes

Original Profile:
influenced by plasma condition & laser fluctuations

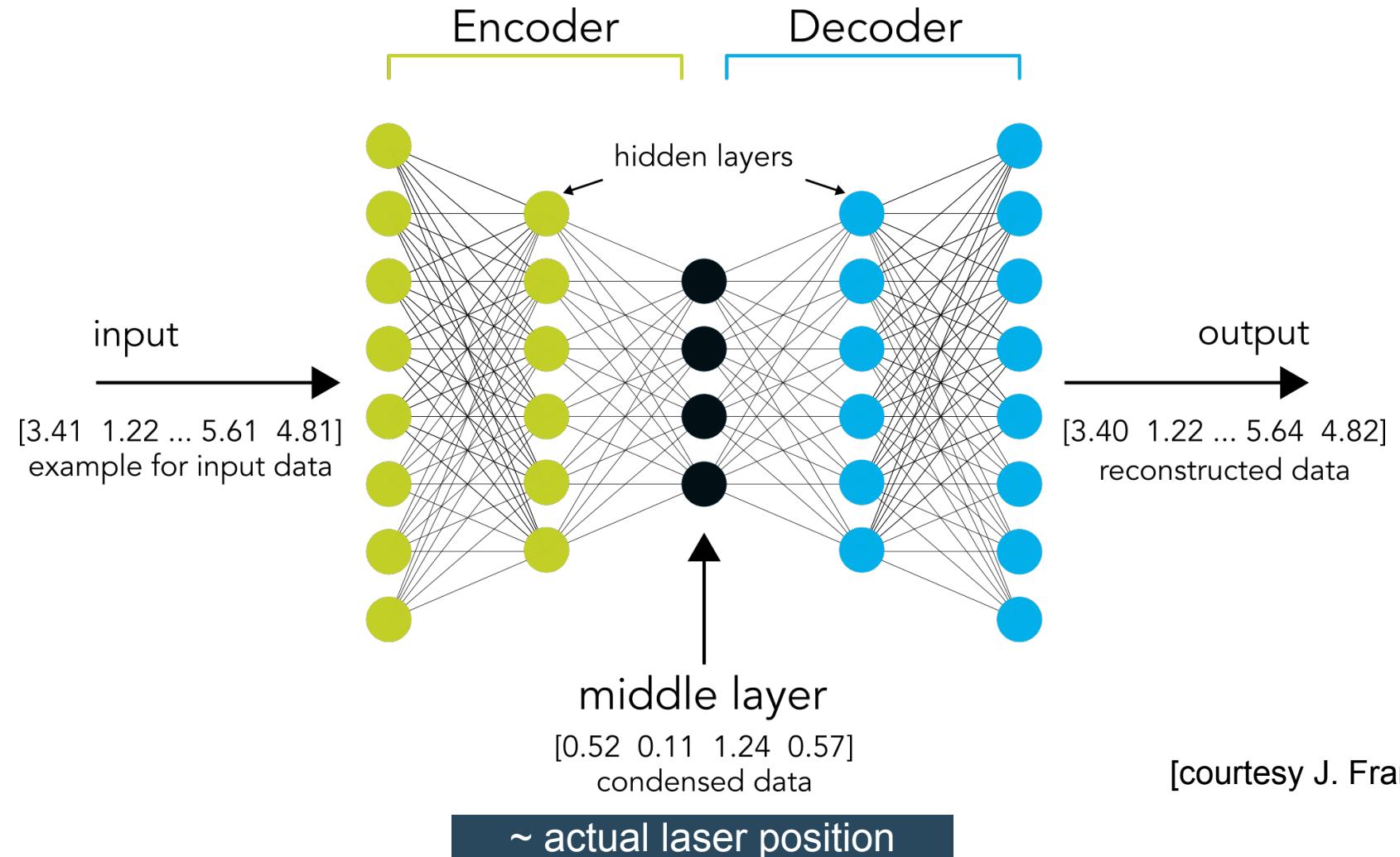


Ratio Profile:
predominantly influenced by laser fluctuations



2. Learn “laser position” with Autoencoder

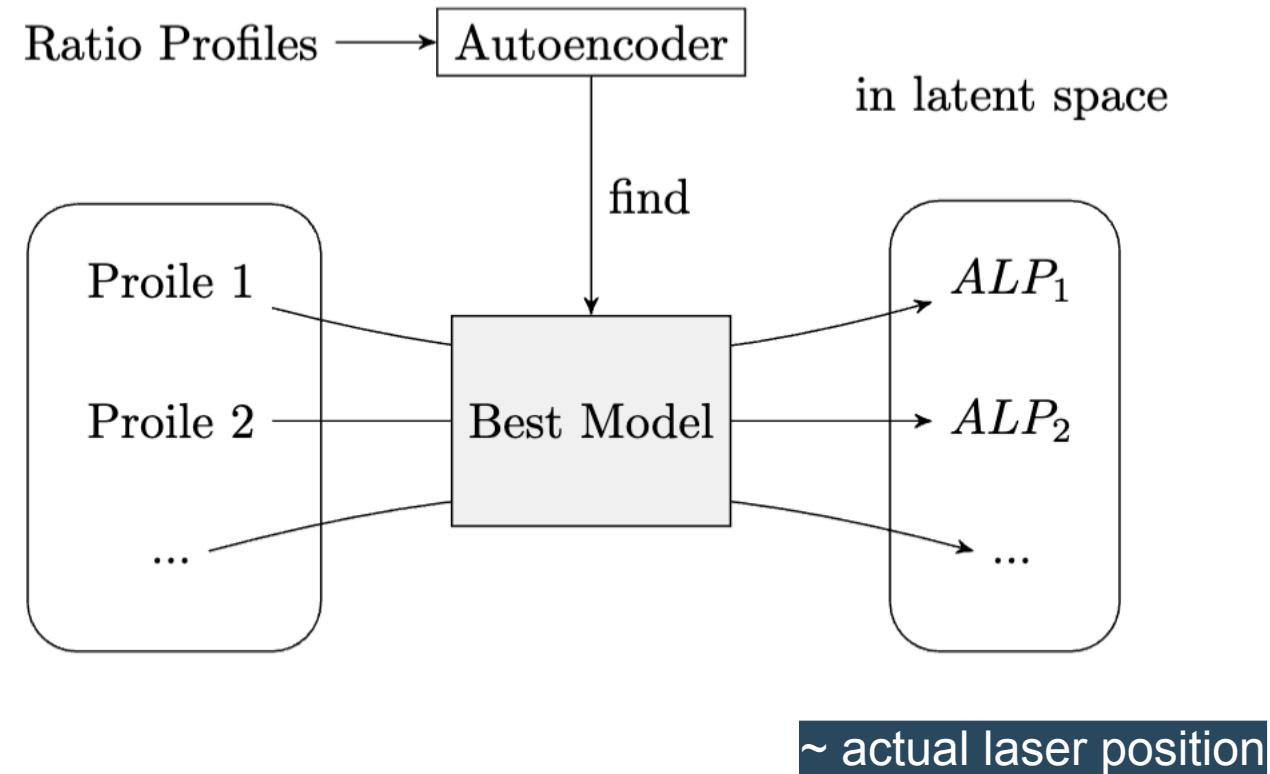
What is an Autoencoder?



[courtesy J. Frank]

2. Learn “laser position” with *Autoencoder*

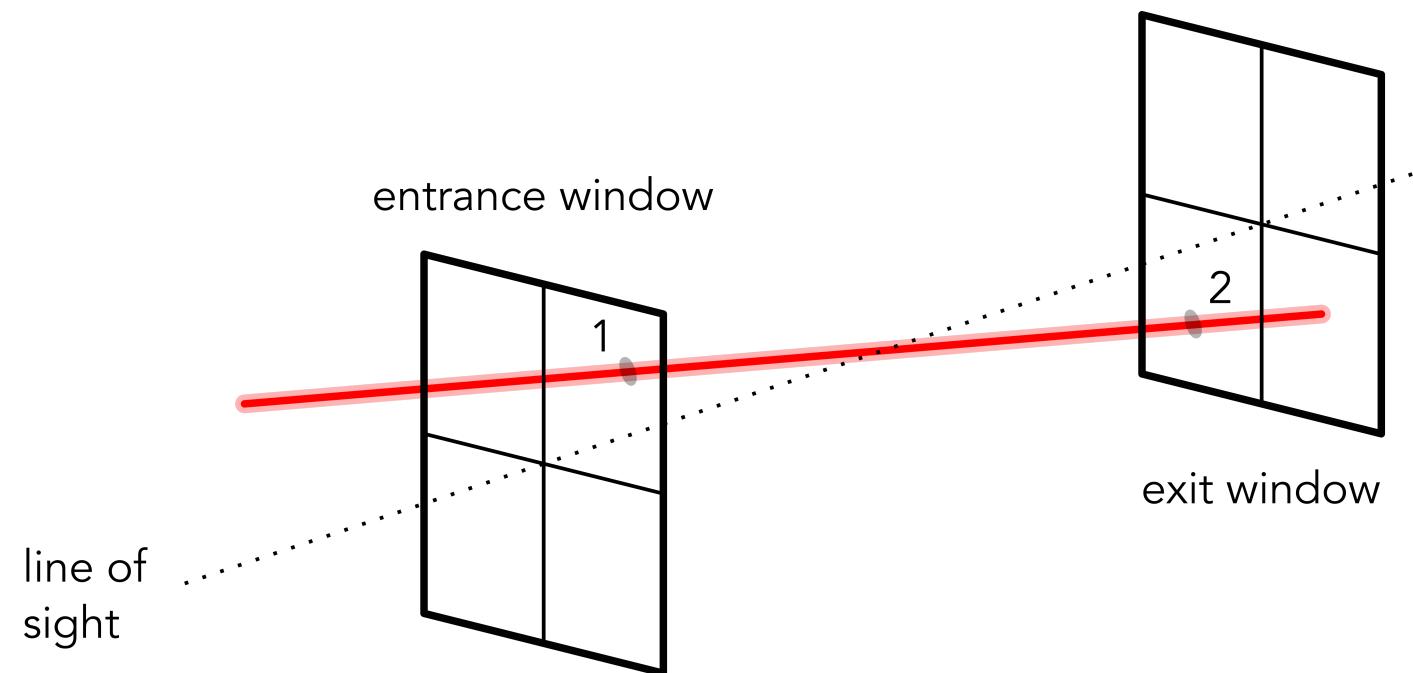
How?



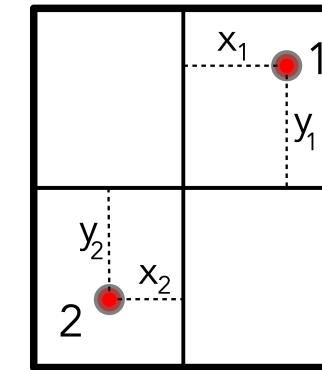
Interlude: Degrees of freedom

Expecting ≤ 4 degrees of freedom

- < 4 : Not necessarily all of them influence profile shape
- > 4 : probably reconstruct statistical effects



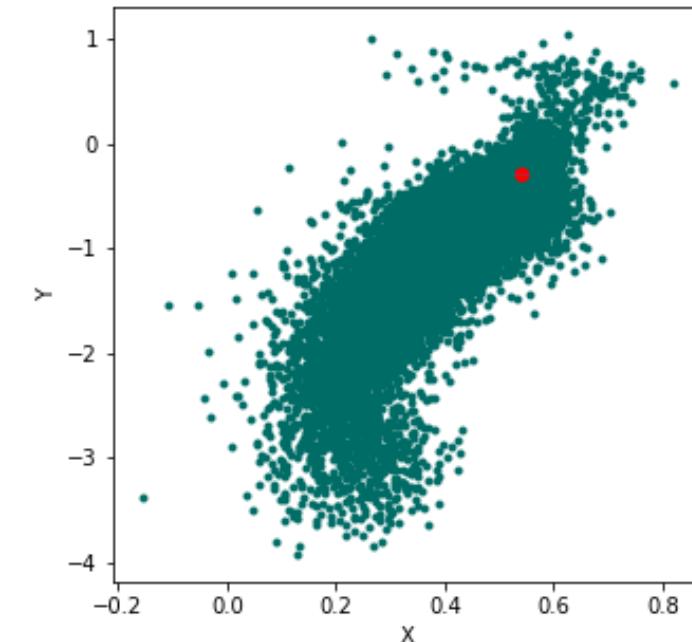
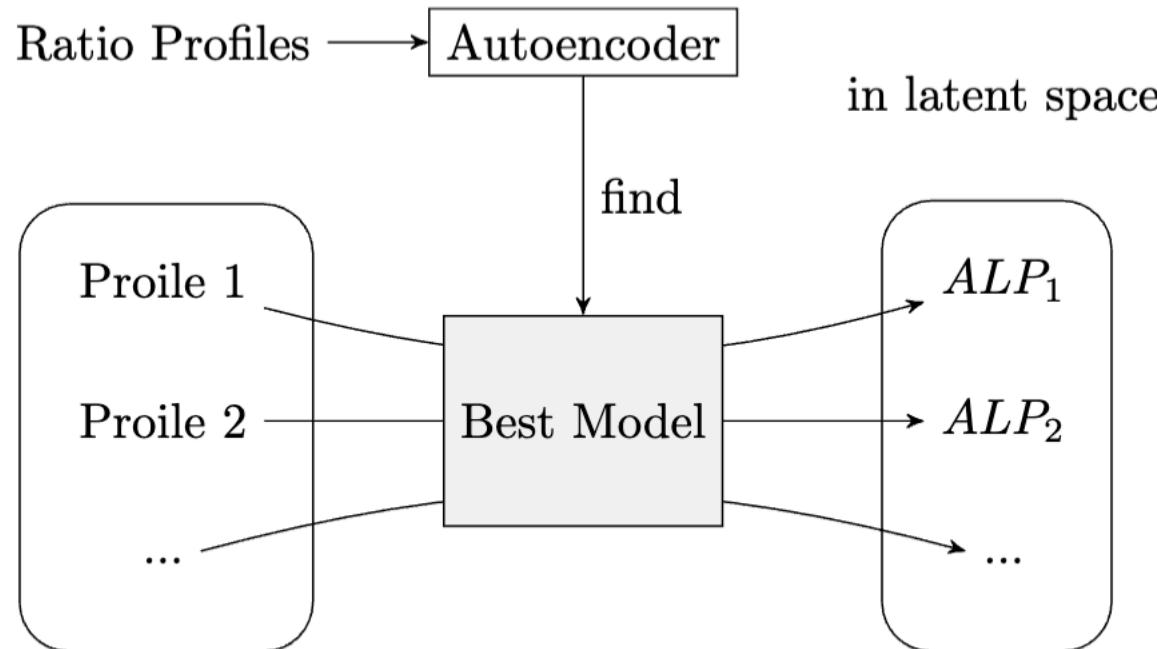
view through both
windows in line of sight



[courtesy J. Frank]

2. Learn “laser position” with *Autoencoder*

How?

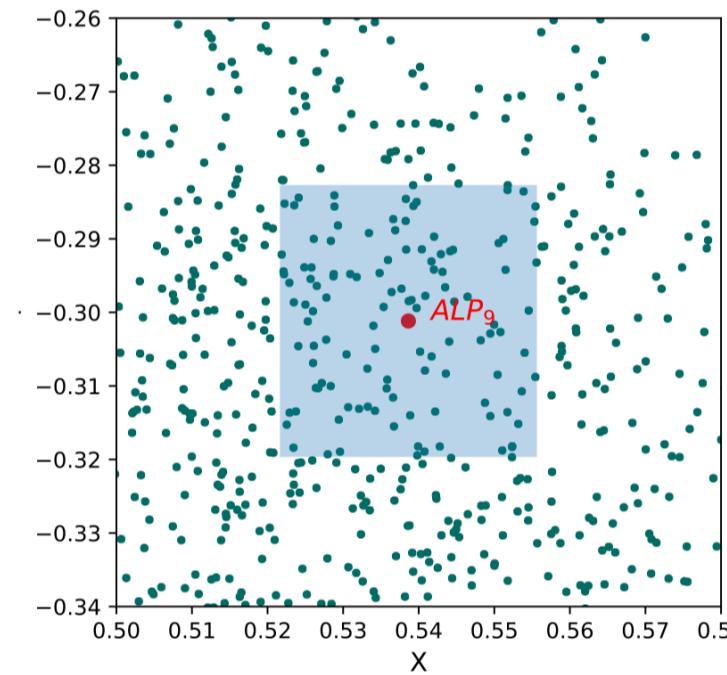
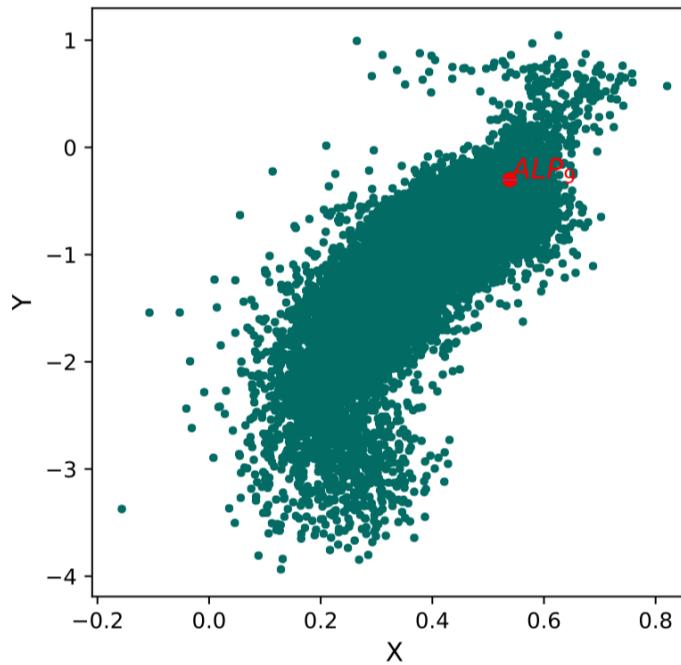


e.g: ALPs in 2D latent space

3. Identify the **best cell** (calibration profile) through ALPs (training result)

Assume: profiles measured with similar laser position are close in latent space

3.1 Assign a **cell** to each profile



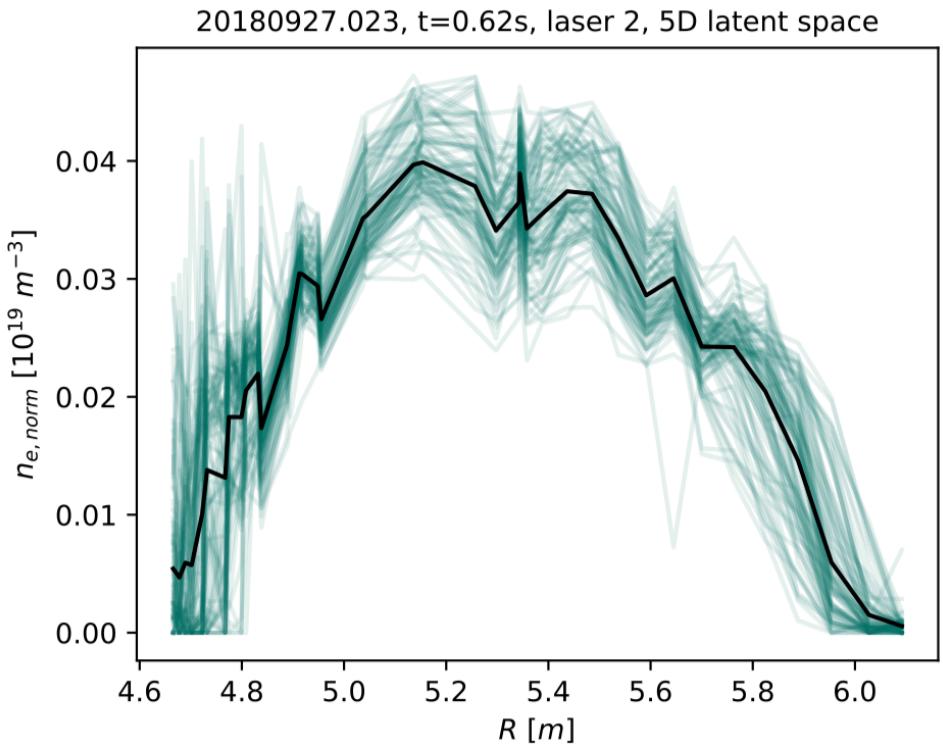
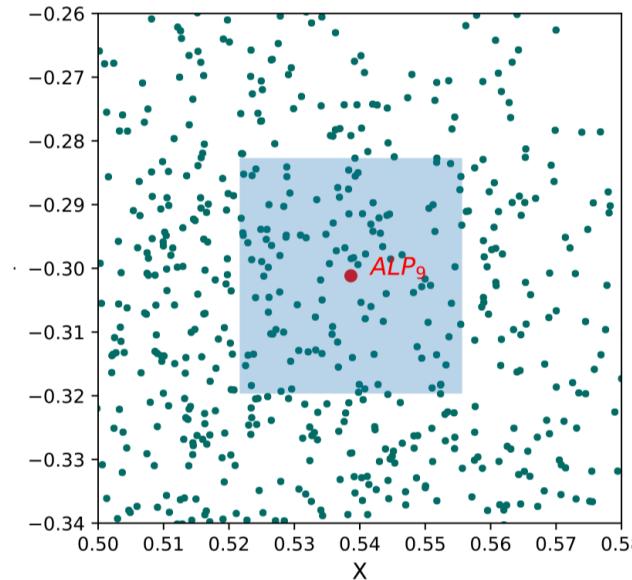
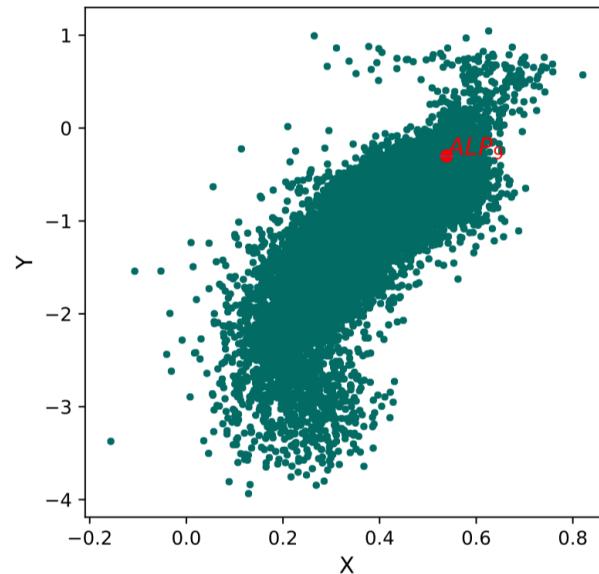
- E.g. Cell of profile 9 in 2D latent space
- Euclidean distance metric:

$$d(p, q) = \sqrt{\sum_{i=1}^N (p_i - q_i)^2},$$

3. Identify the *best cell* though ALPs

Assume: profiles measured with similar laser position are close in latent space

3.2 Get generalised profile



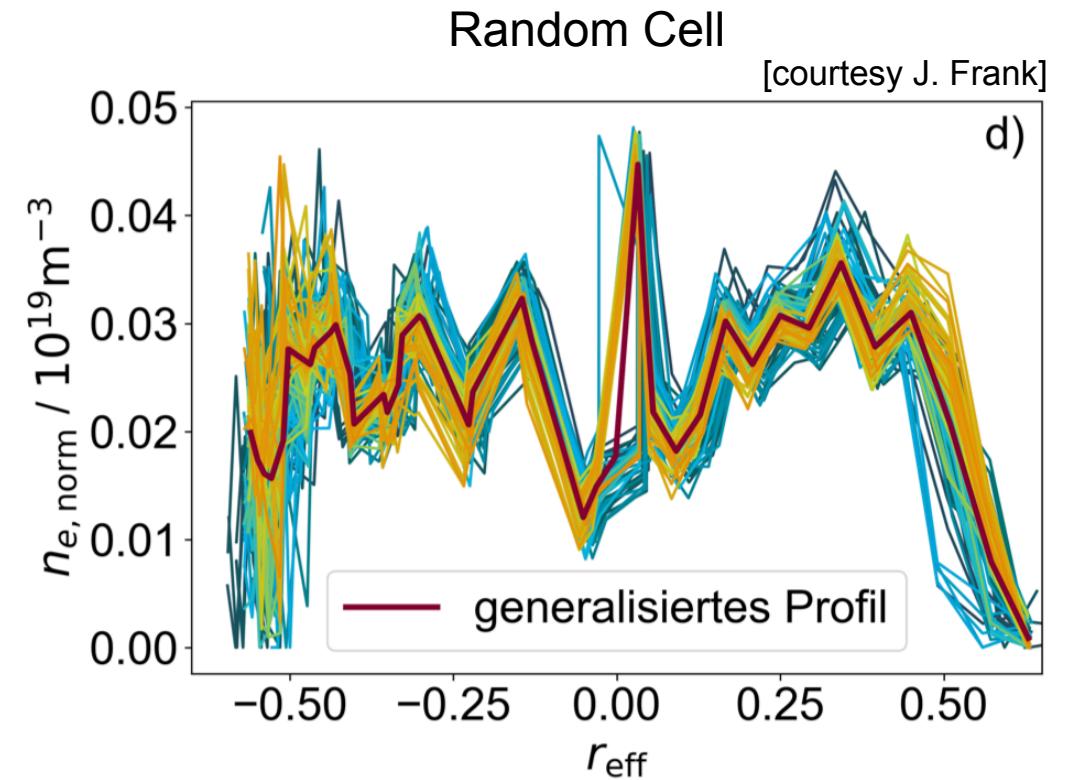
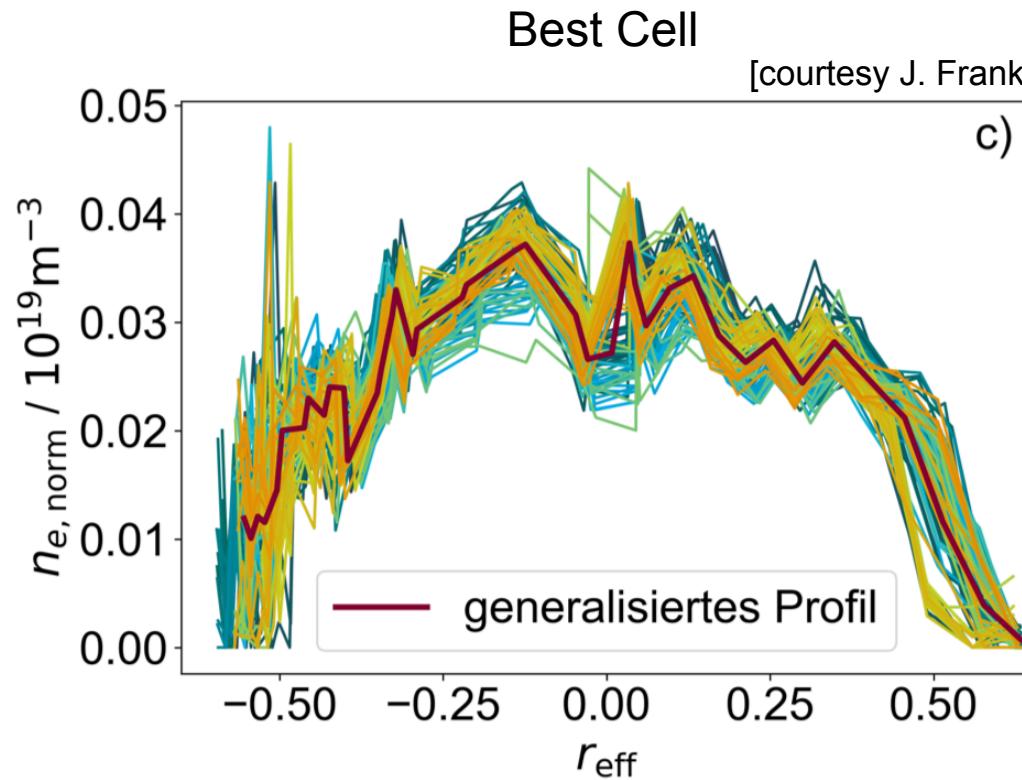
Mean Profile of normalised profiles
(*generalised profile*)

3. Identify calibration position through ALPs

Assume: profiles measured with similar laser position are close in latent space

3.3 Identify the best cell

By evaluating how ideal the shape is (Smooth & symmetrical)



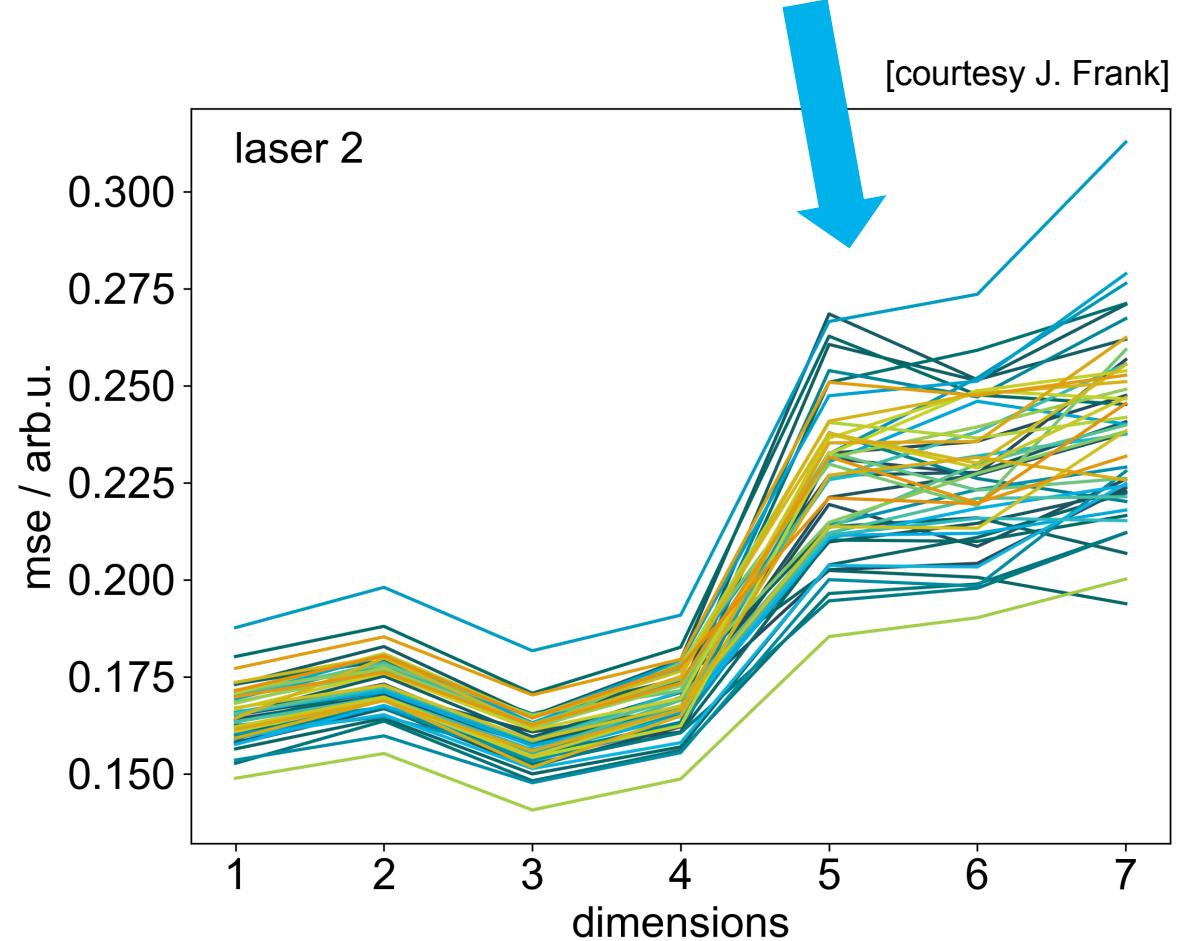
3. Identify calibration position though ALPs

Assume: profiles measured with similar laser position are close in latent space

3.4 Compare best cell of different Dimensionalities

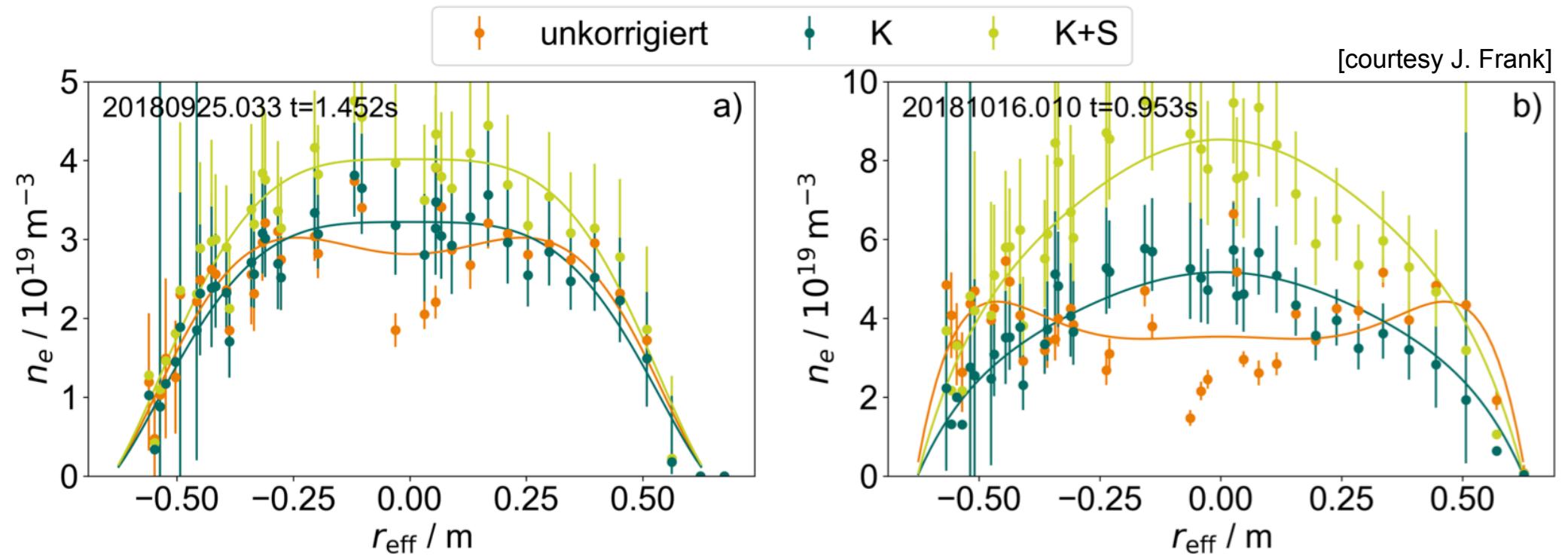
- >4D seem to worsen the correction
- 1-4D are rather similar
- Minimum mse seems to be at **3D**

→ Agrees with expectation from geometric configurations



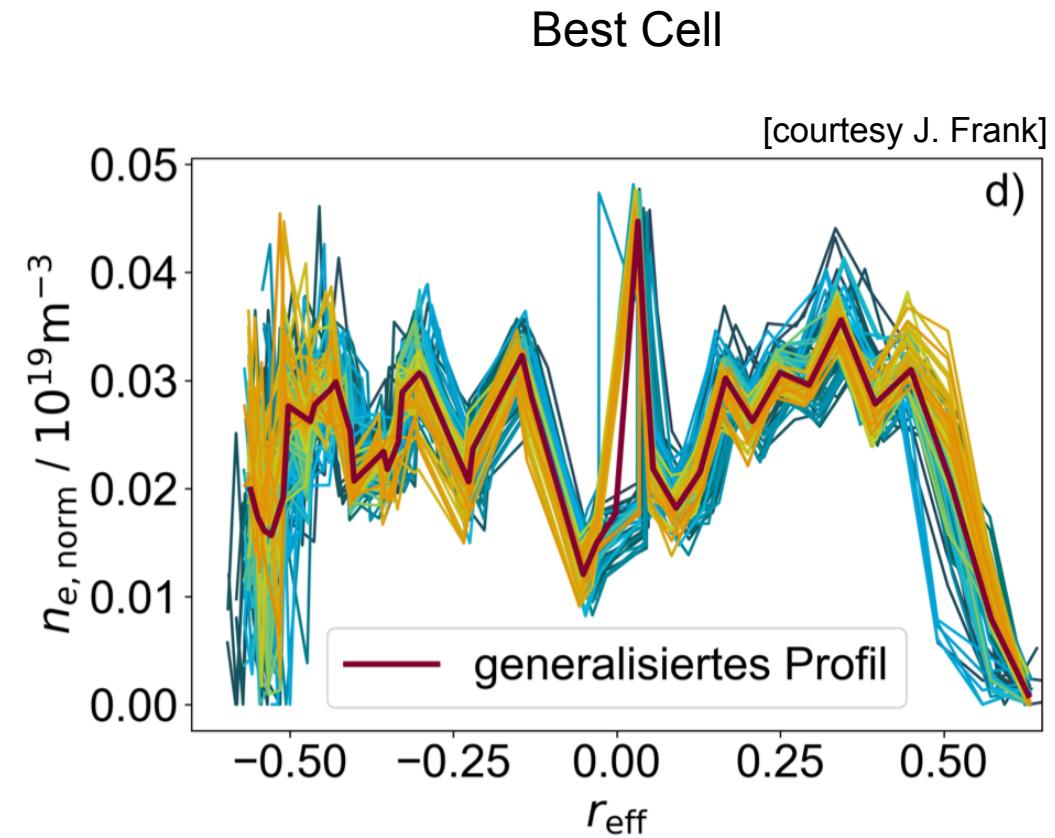
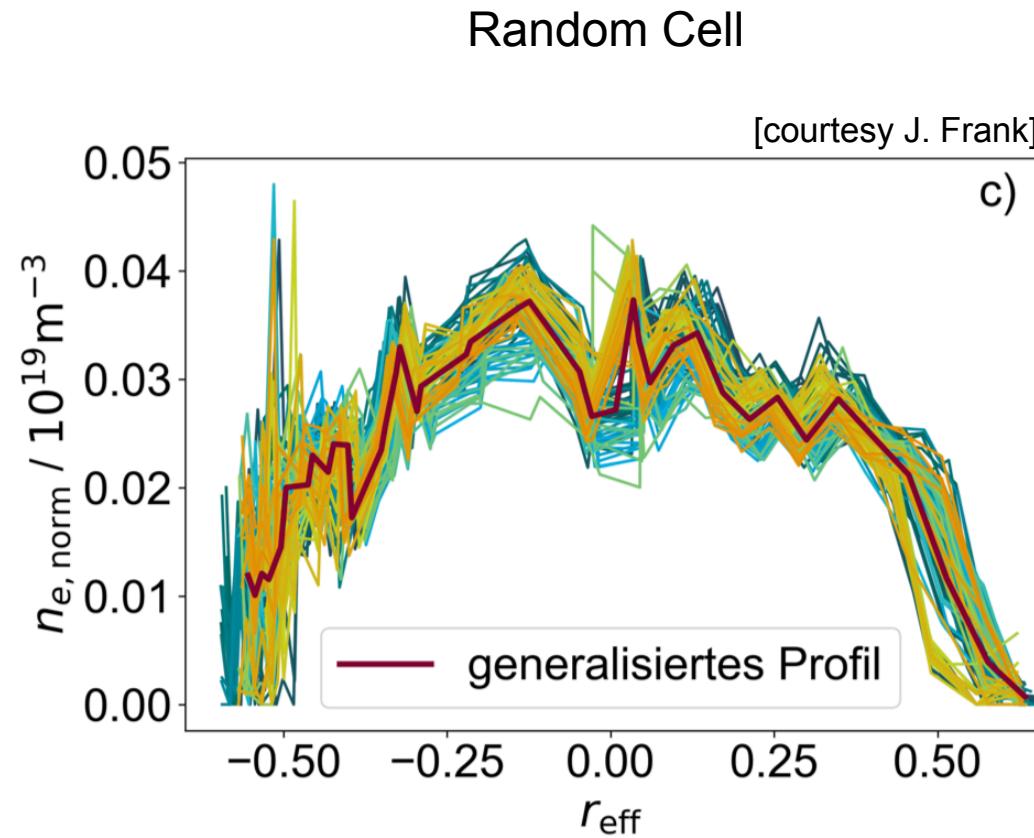
4. Retrospective Correction

$$n_{k,corrected} = n_k \cdot v_k \cdot s \quad v_k = \frac{n_{k,gen,cali}}{n_{k,gen}} \quad s = \frac{\int n_{cali}}{\int n}$$



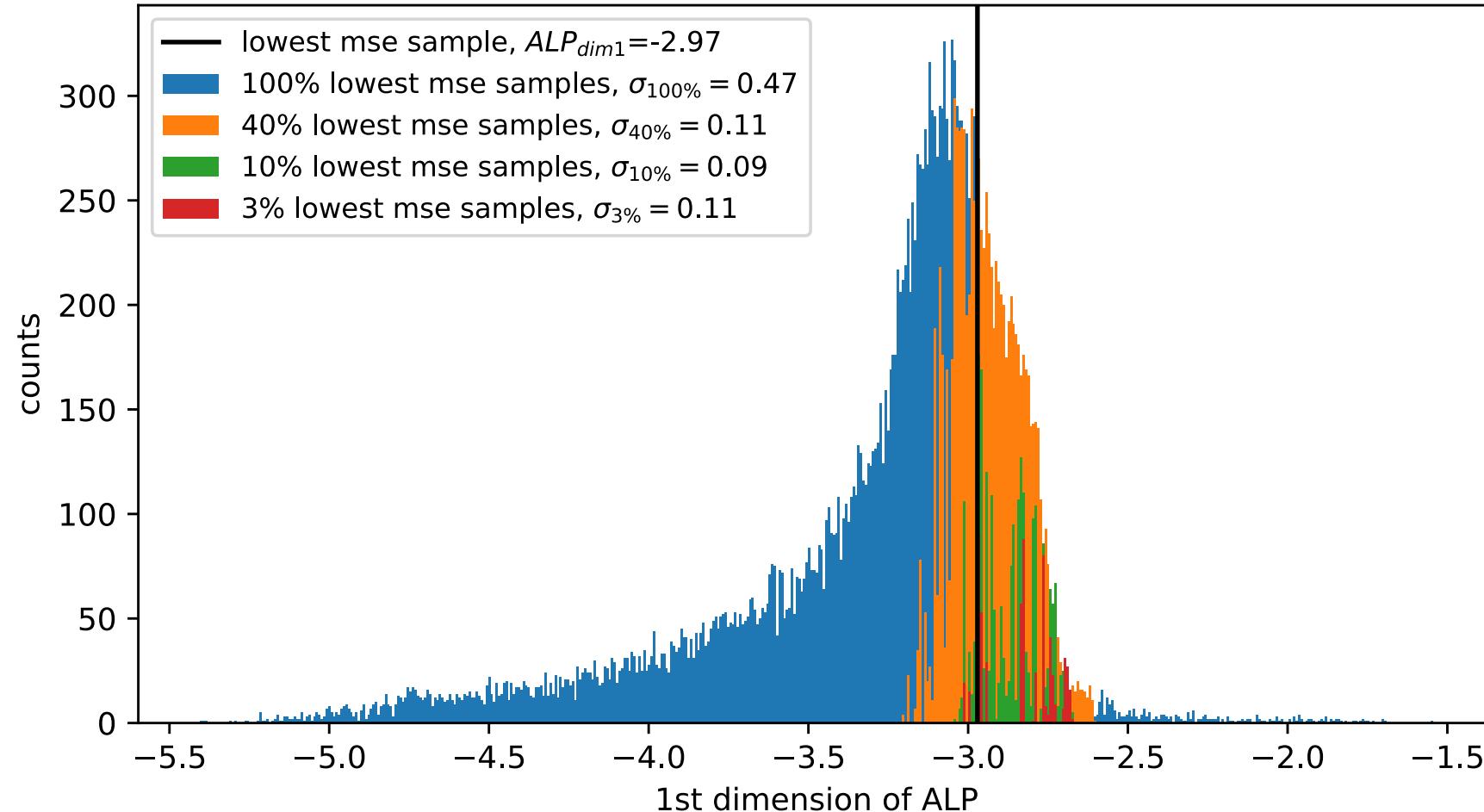
Instability in the Existing Method (Calibration Cell Selection)

Possible result:

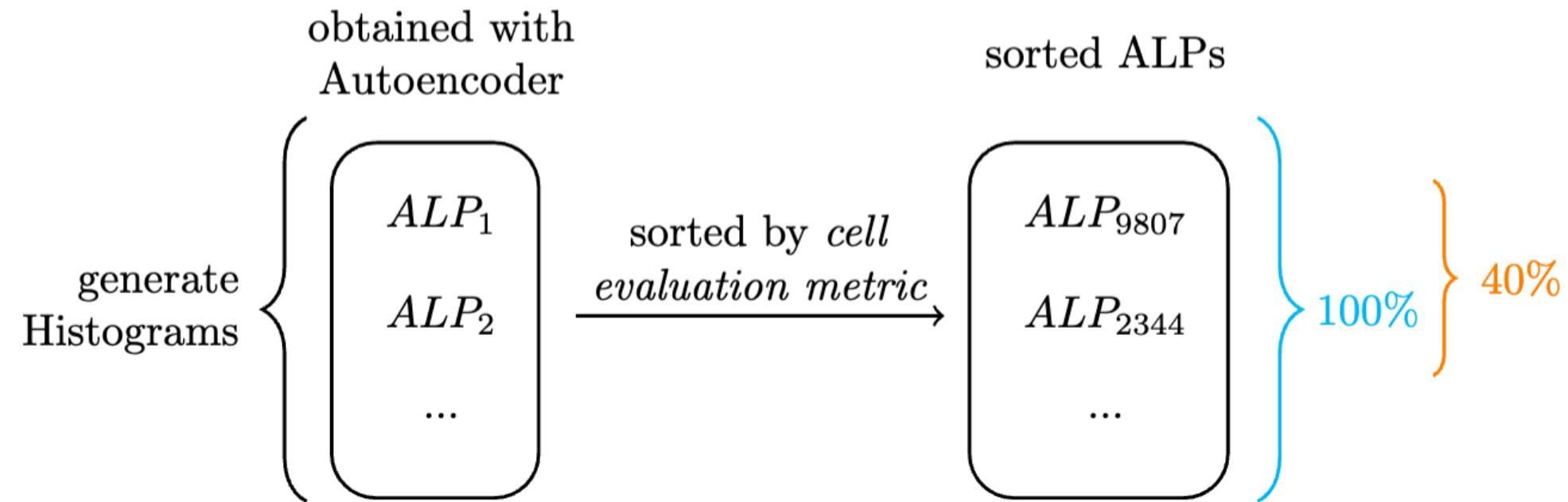




Color-coded histogram (what?)

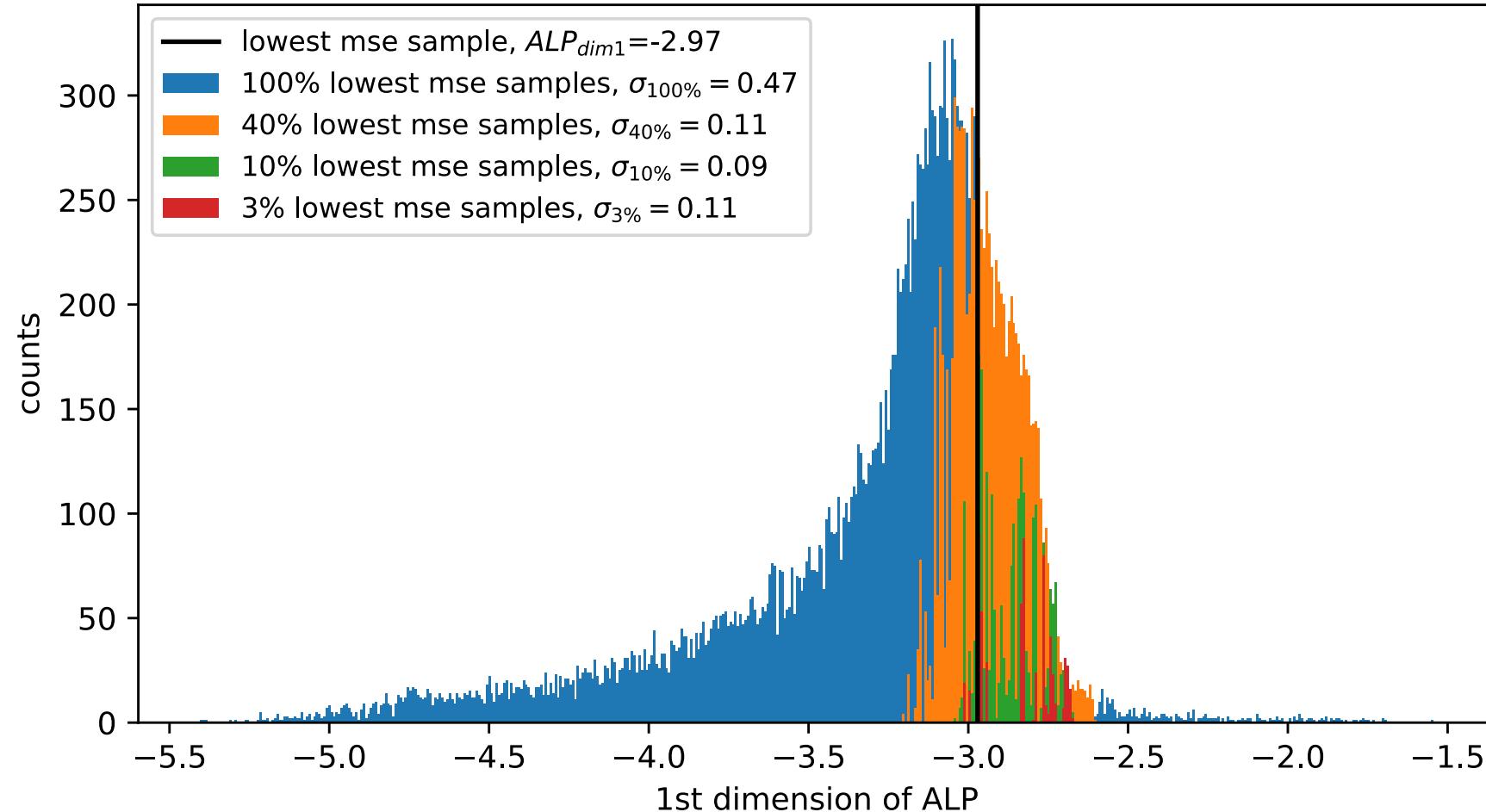


Instability in the Existing Method (Calibration Cell Selection)



Instability in the Existing Method (Calibration Cell Selection)

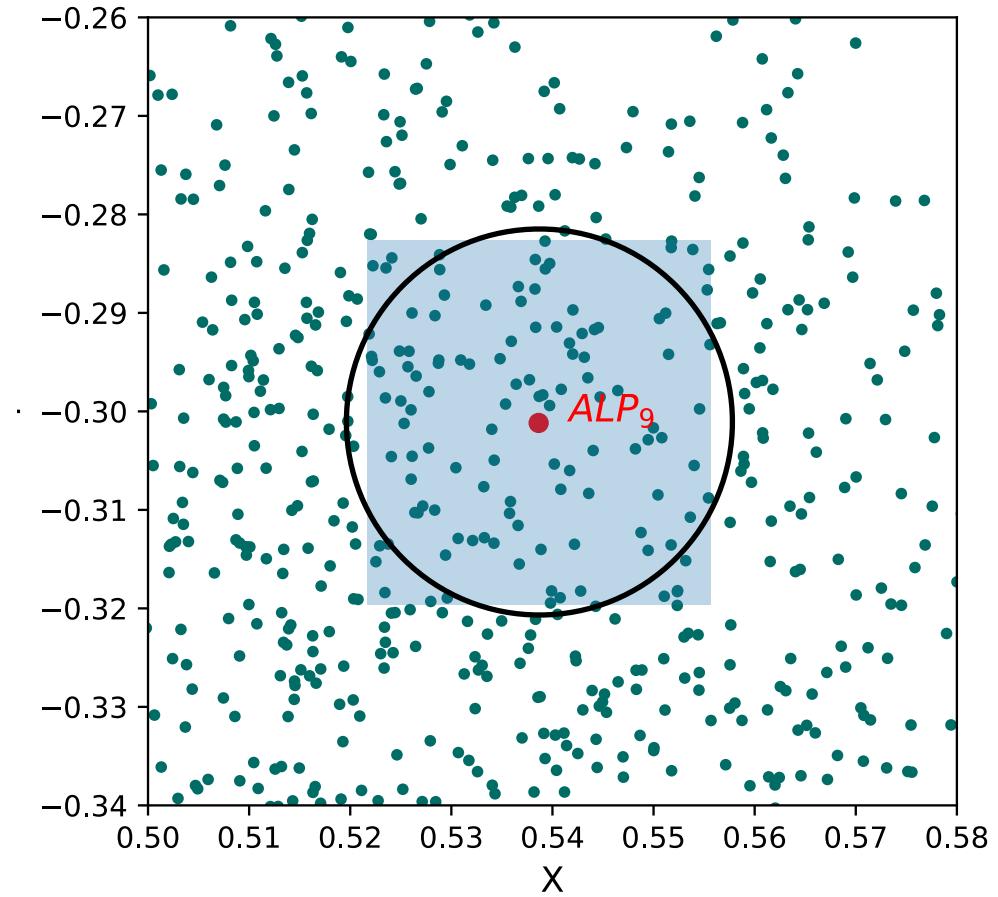
Histogram of ALPs in 1D latent space with different percentages



Improvement & Results

Spherical Cell Definition

- Change cell geometry from n-cube to n-ball
- Same distance metric
- Computation time has been reduced **from up to 2 weeks to only 6h** (on 4 CPU cores)



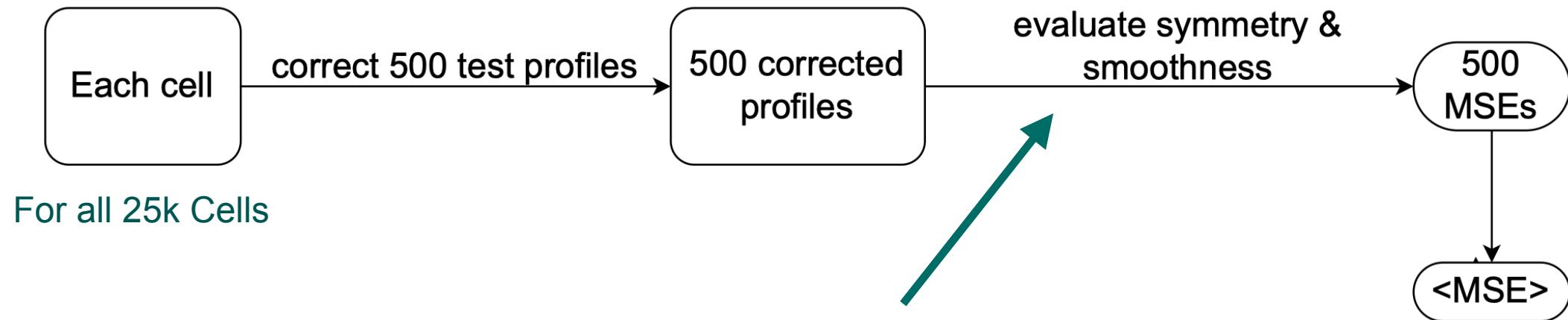
Correction-Based Calibration Cell Selection

Based on **how well** each cell corrects.



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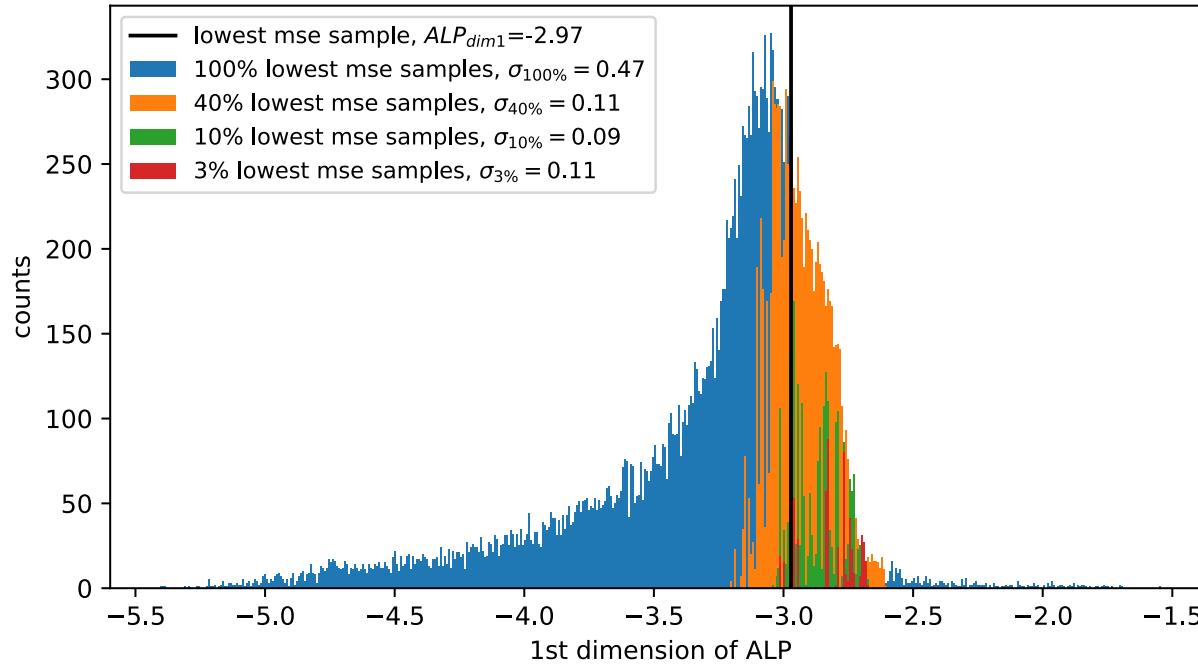
$$MSE_{in/out} = \frac{1}{S(I_{in/out})} \sqrt{\sum_{i \in I_{in/out}} \frac{F(r_{\text{eff}})_i - n'_{e,i}}{F(r_{\text{eff}})_i}}$$

$$\langle MSE \rangle = \frac{1}{2}(MSE_{in} + MSE_{out})$$

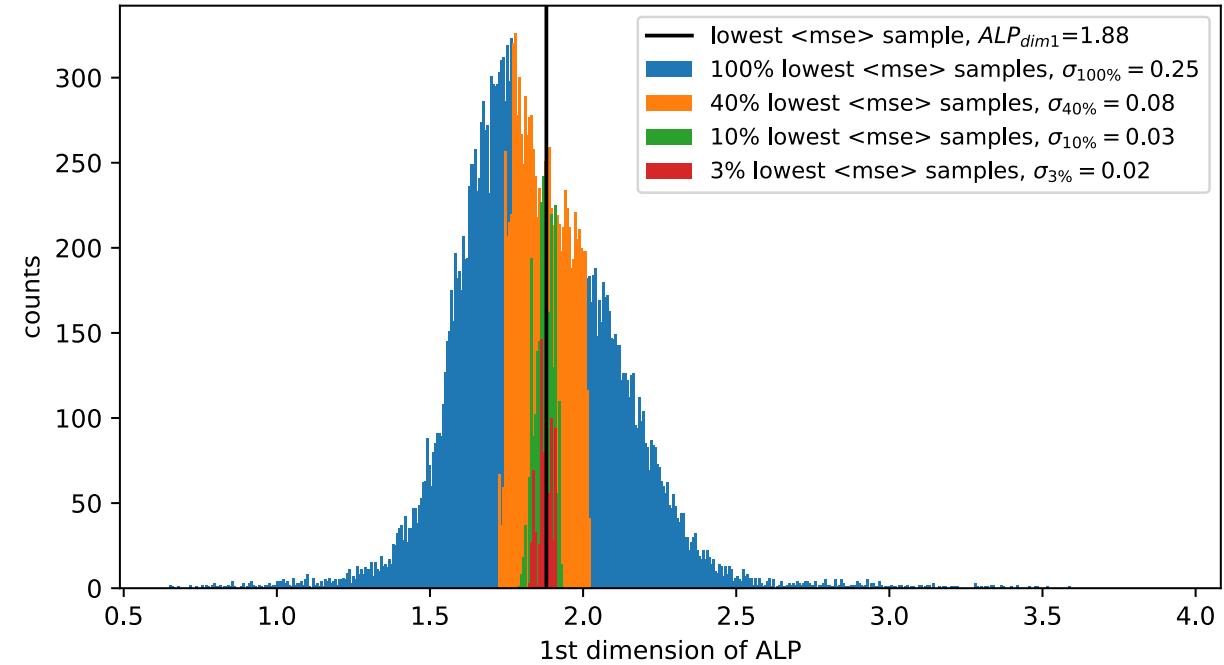
Correction-Based Calibration Cell Selection



ALPs sorted by old metric



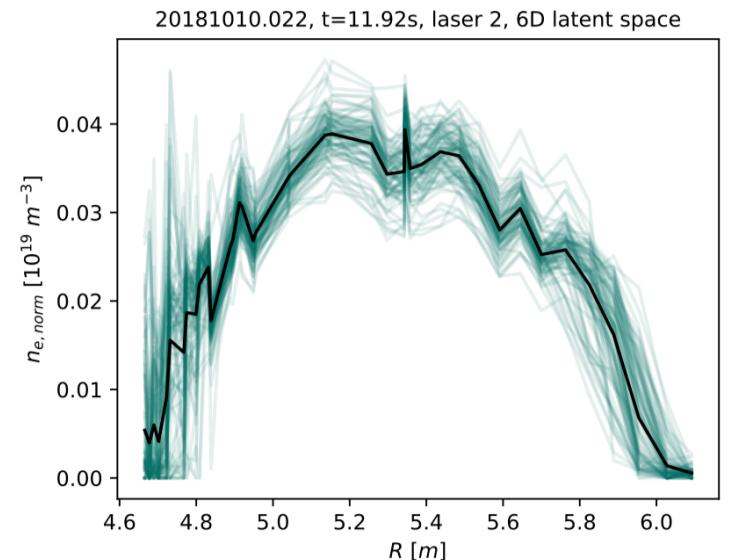
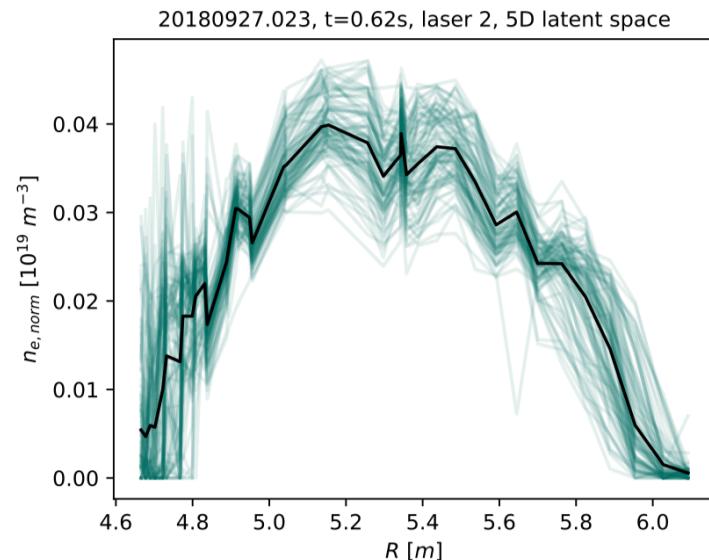
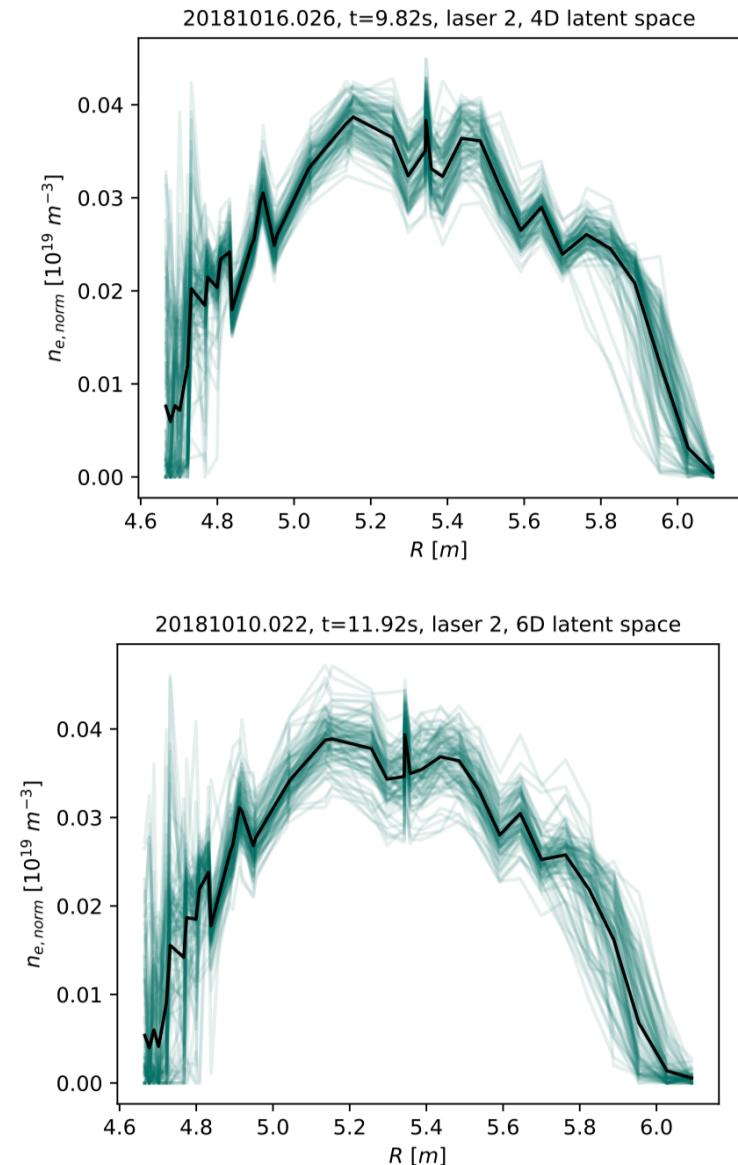
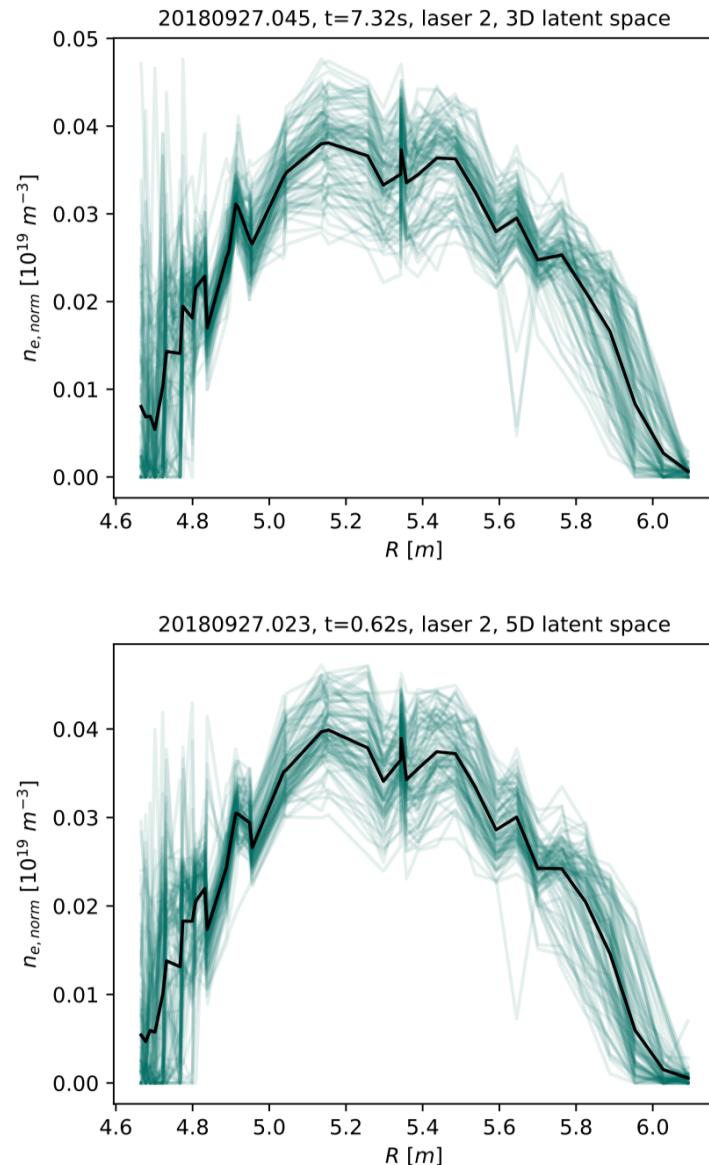
ALPs sorted by new metric



- More concentrated
- Std decrease with lower <MSE>s
- Groups nested (Red enclosed in green in orange)

Calibration Cell e.g. laser 2, 3 – 6 neurons in latent space

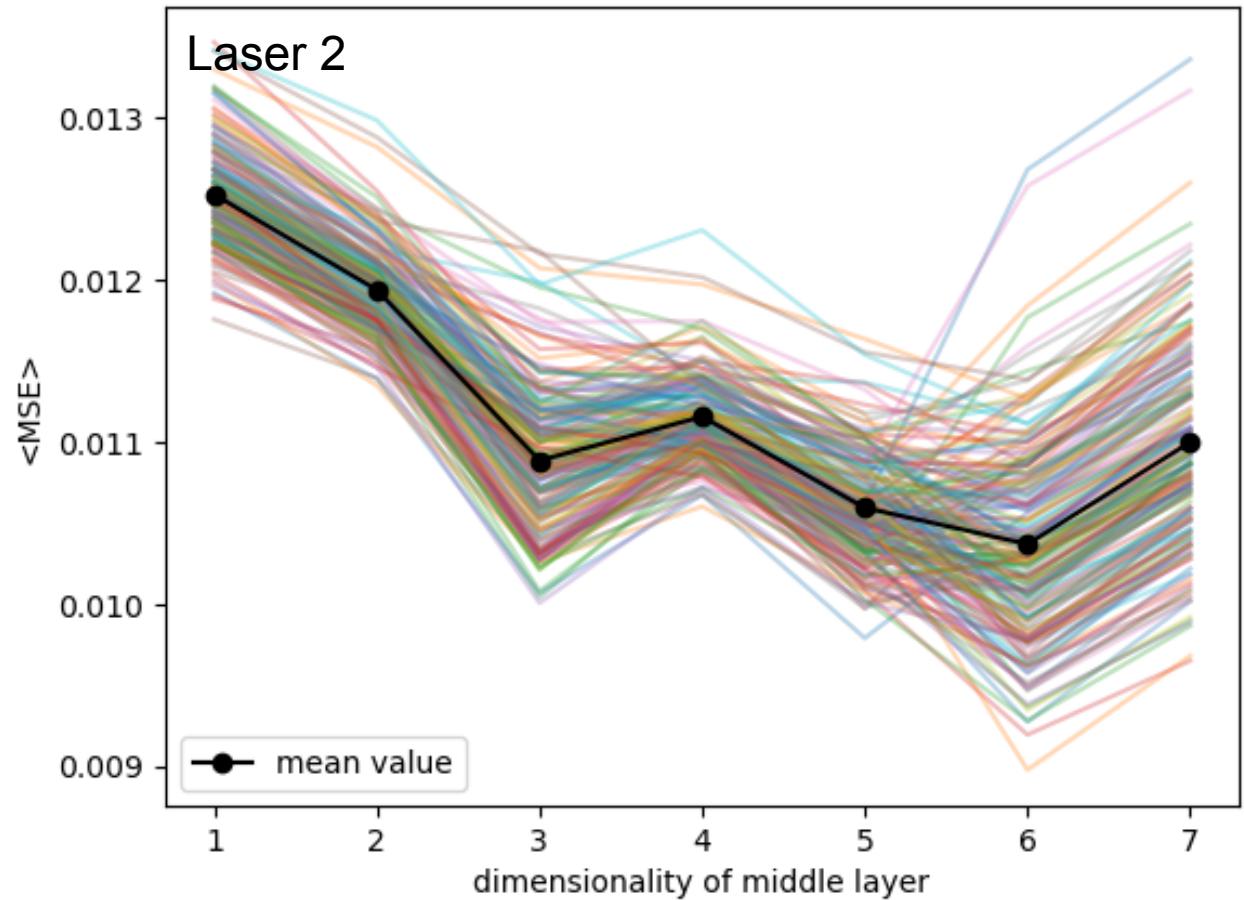
- similar
- Peaks in the centre — assumption not perfectly valid (smoothness)
- 3D & 5D: more dispersed compared to the others
- **4D**: concentrated — optimal



Dimensionality of the Latent Space

$\langle \text{MSE} \rangle$: how well the best cell within 1-7D latent space corrects

- Run 200 times — reduce noise; represent stability
- $\langle \text{MSE} \rangle$ — similar as in cell selection
- **Lower $\langle \text{MSE} \rangle$** ~ smoother & more symmetrical corrected profile \rightarrow ideal Dimensionality



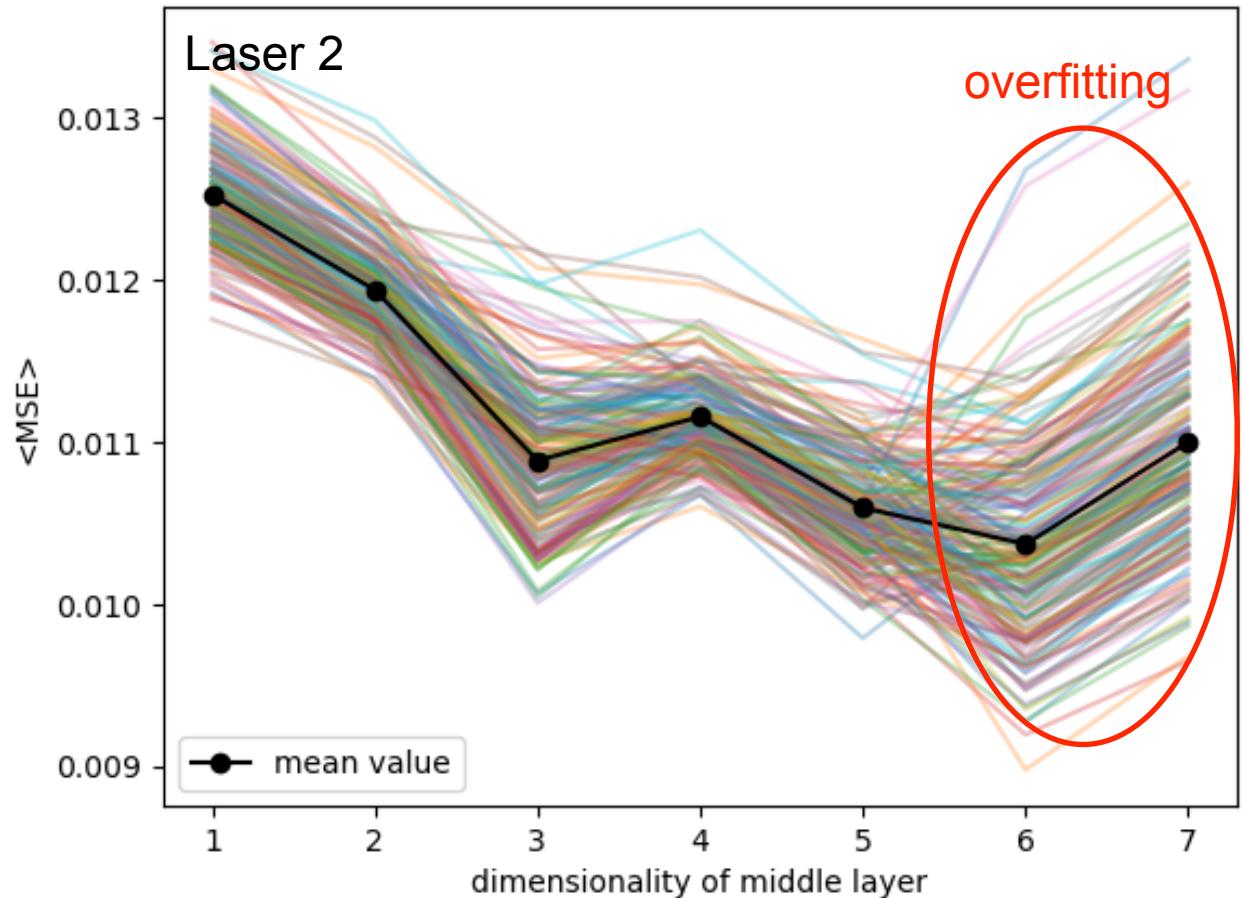
Dimensionality of the Latent Space

$\langle \text{MSE} \rangle$: how well the best cell separately identified with networks with 1D-7D latent space corrects

- Run 200 times — reduce noise; represent stability
- $\langle \text{MSE} \rangle$ — similar as in cell selection
- Lower $\langle \text{MSE} \rangle \sim$ smoother & more symmetrical corrected profile \rightarrow ideal Dimensionality

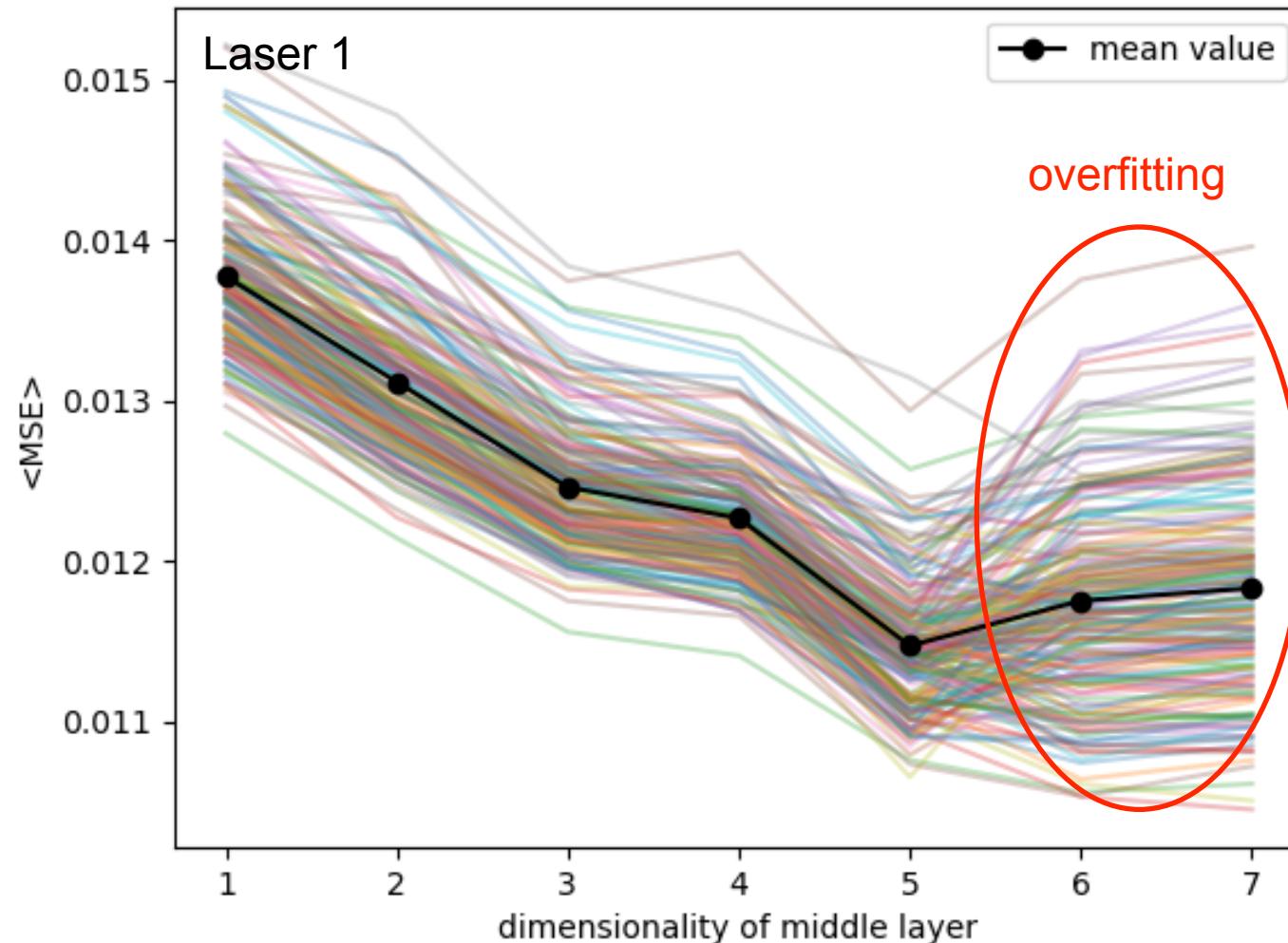
For laser 2:

- Low in 3D-7D
- Possible overfitting for $\geq 6\text{D}$

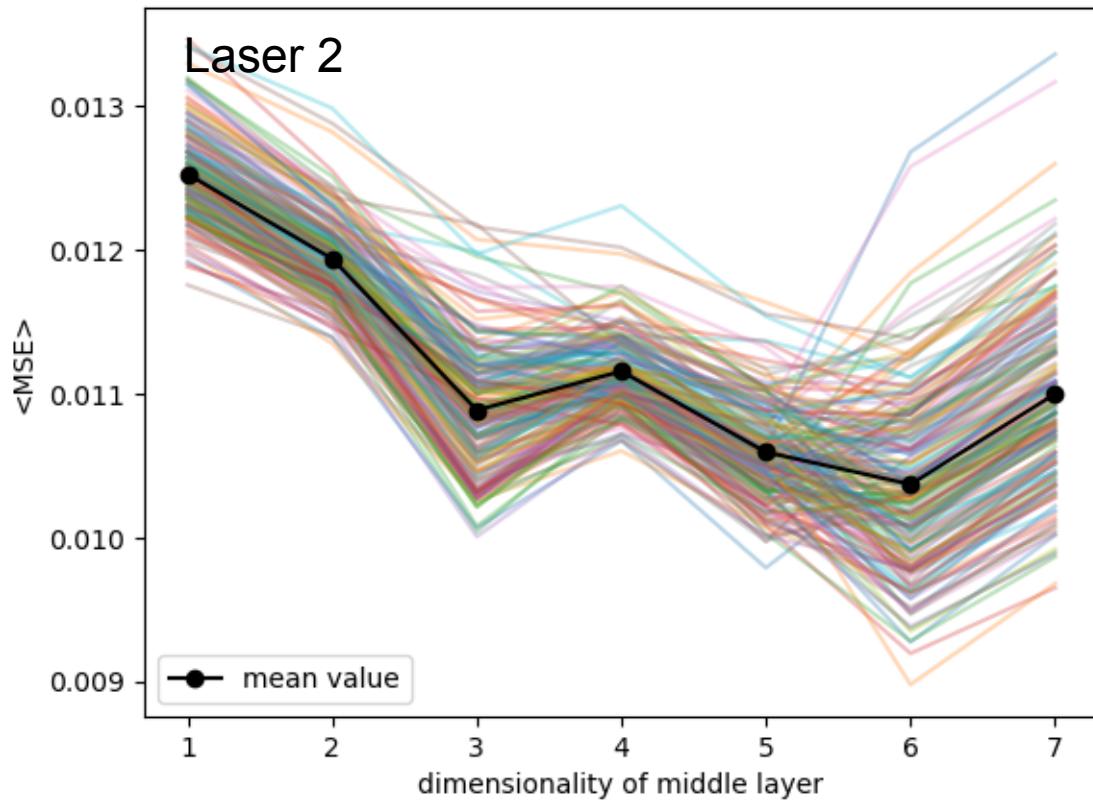


Dimensionality of the Latent Space

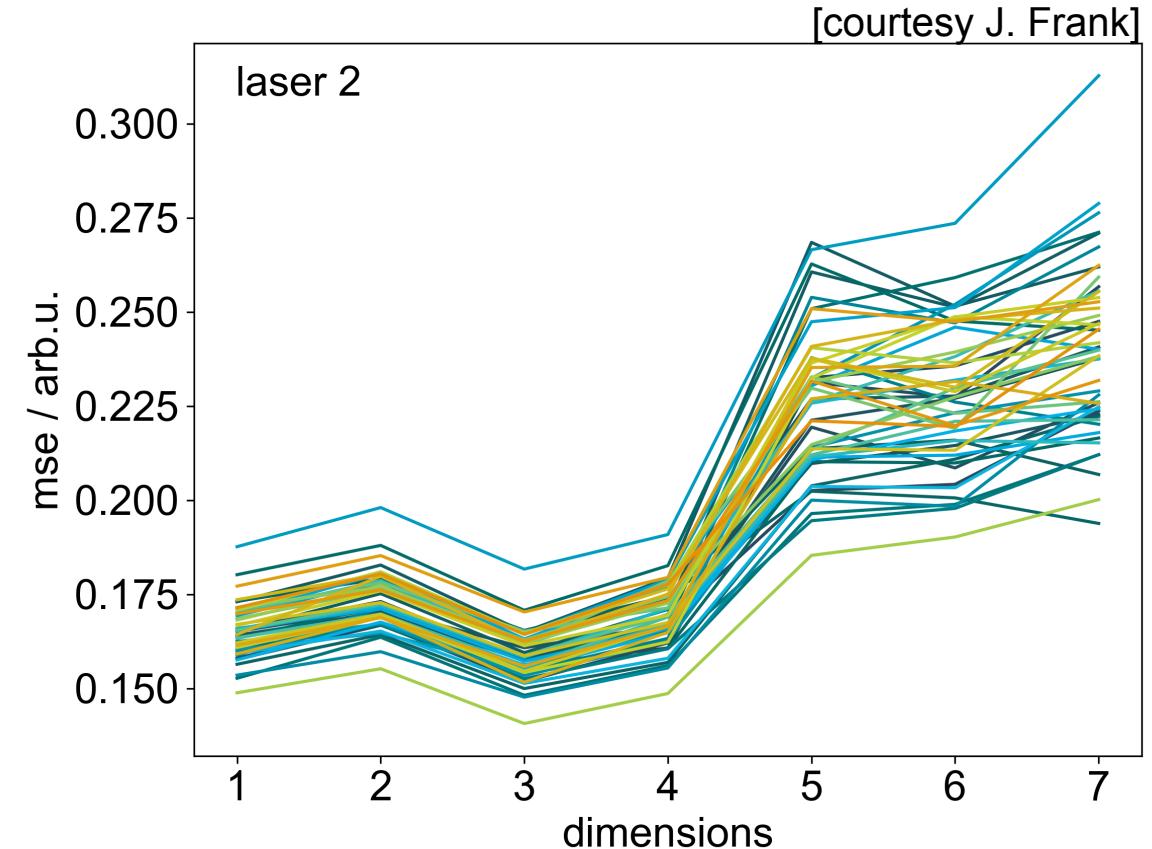
$\langle \text{MSE} \rangle$: how well the (newly identified) best cell of networks with dimensionality 1-7 corrects



Dimensionality of the Latent Space

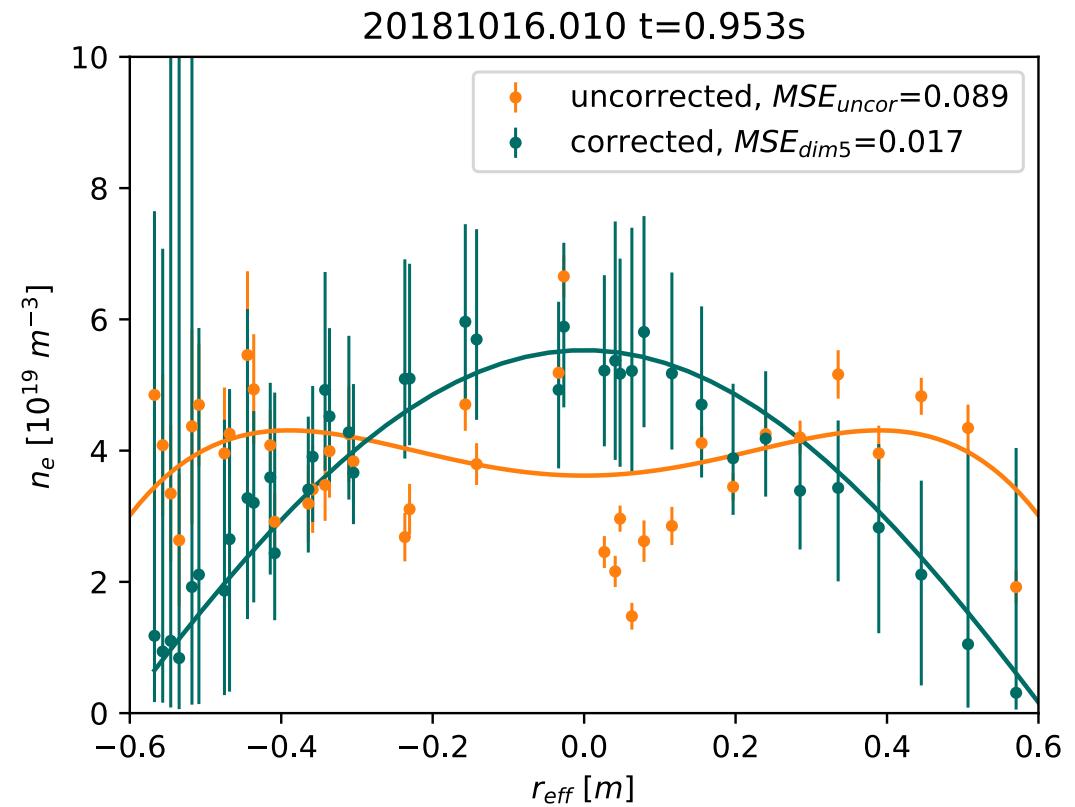
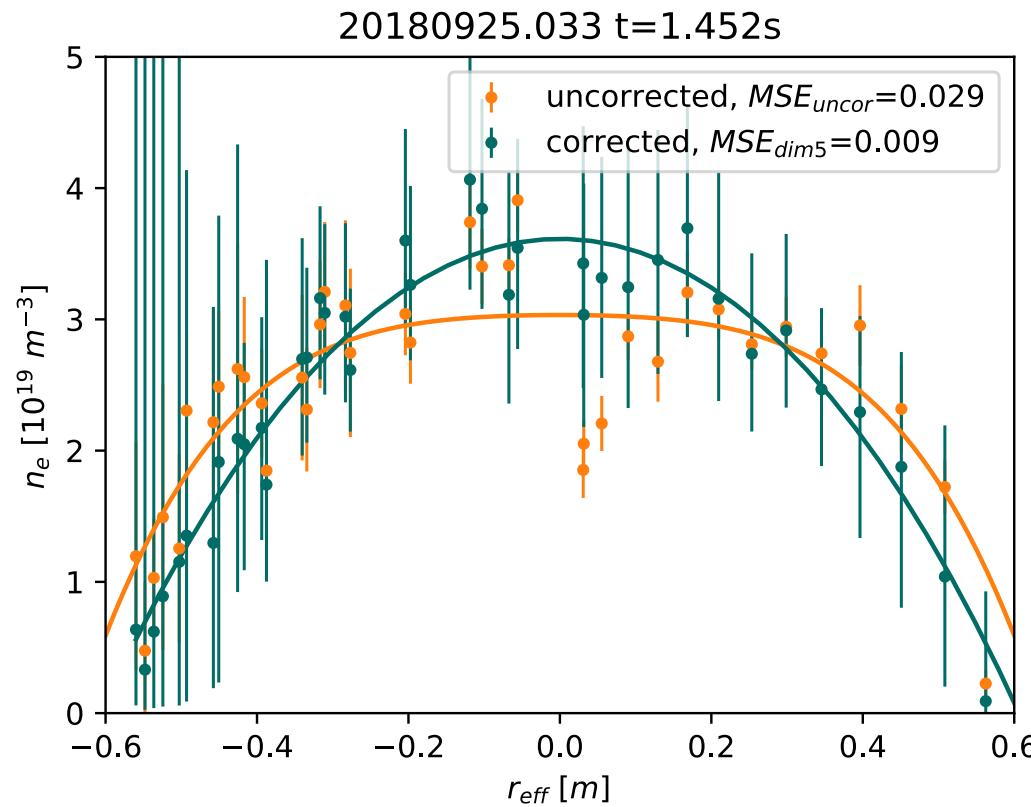


Best now: 5D



Best previously: 2/3D

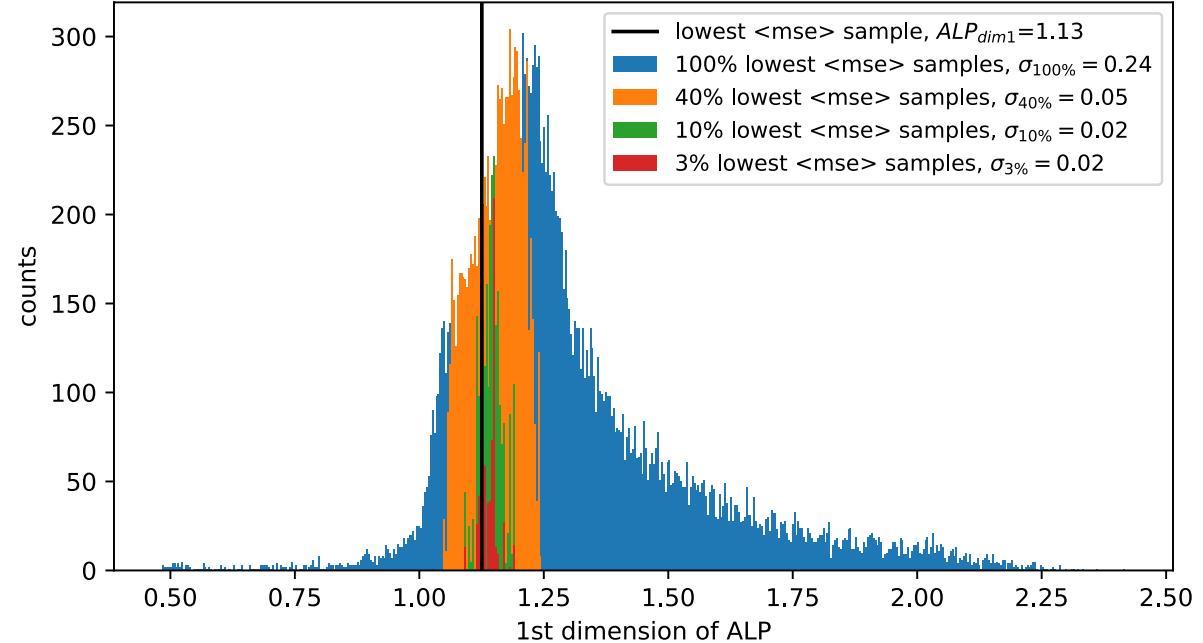
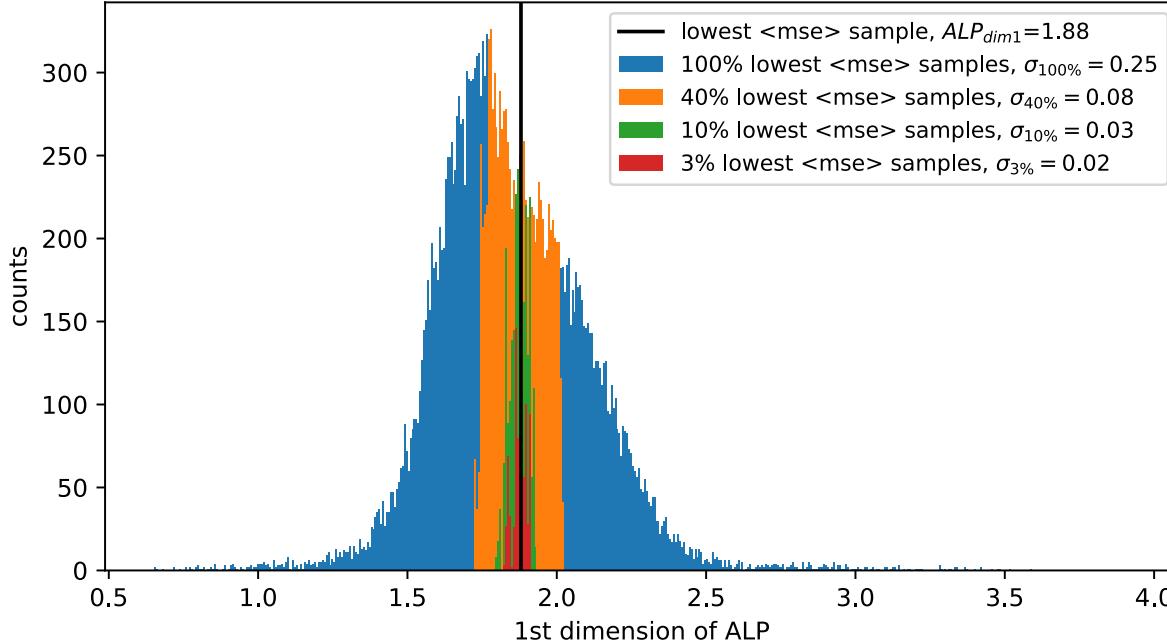
Corrected Profile Shapes



- **Similar result** as the previous method — possible same calibration cell is chosen
- but more confident — **more reliable**

Reproducibility

Evaluation results from the 2 runs (laser 2, 1D latent space)



- Qualitatively different as expected
- Similar contraction effects

Summary and Outlook

Summary



- Main focus: **more stable Calibration Cell identification** —> more reliable correction
- Significantly reduced computing time by changing the geometry of a cell —> **more efficient**
- Best latent space dimensionality: 4/5 — different as the previous study (3D) and expectation ($\leq 4D$), reason unclear
- Now: The entire process takes about **a week to compute on 4 CPU cores** per laser per latent space dimensionality. (Training 3-4d, evaluation 3d)

Limitation & Outlook



- not very obvious which **dimensionality for the latent space** is most suitable
 - probably can be improved with a **new cell clustering method** for different latent spaces
- only partial absolute correction has been achieved — solution needed (have ideas)