

# Results From An International MHD Data Mining Collaboration

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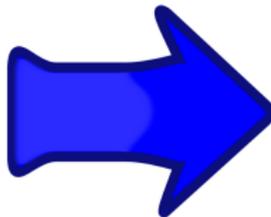
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# Outline

- 1 Introduction
- 2 Method
  - Overview
  - Preprocessing
  - Clustering
- 3 Results
  - H-1
  - TJ-II
  - Heliotron-J

# What Is Data Mining?

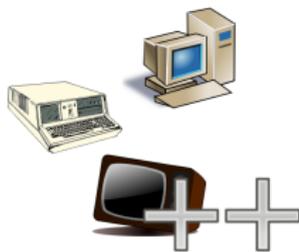
Very Large Databases



New Knowledge



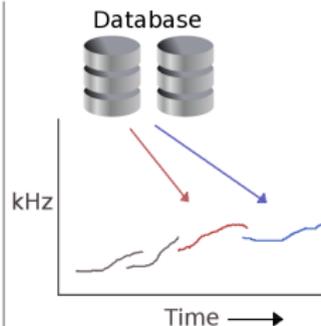
# Motivation



Can we obtain  
new results from  
old data?



Can we contribute  
to other MHD  
database projects?

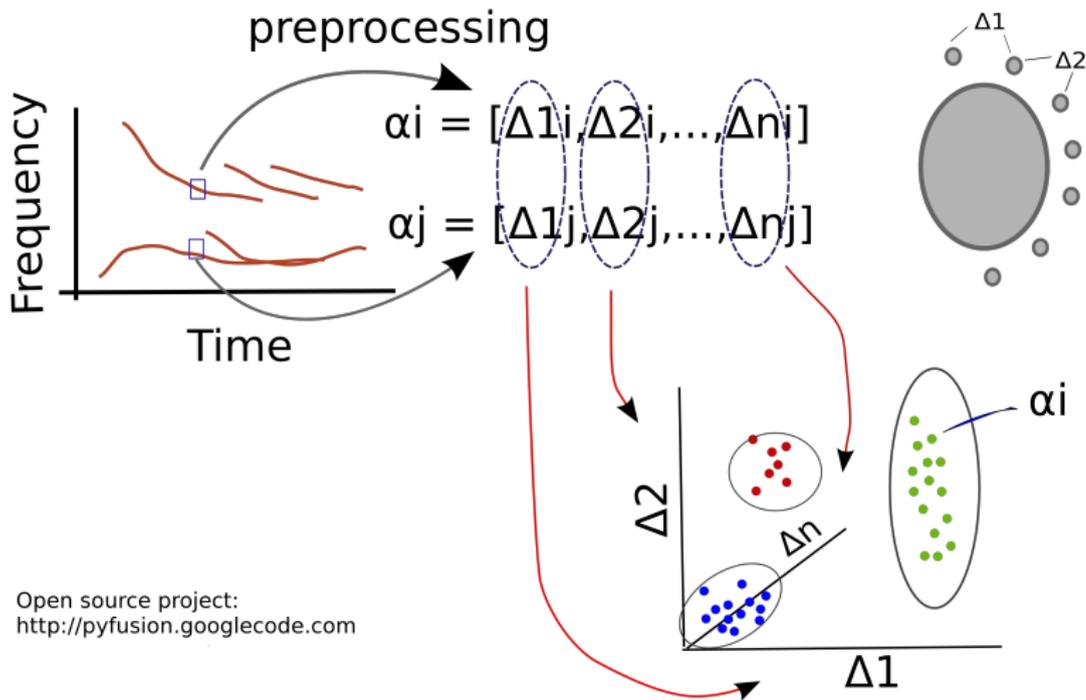


Can we use a MHD  
documentation database  
for identification of new  
data? in real-time?

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# Overview



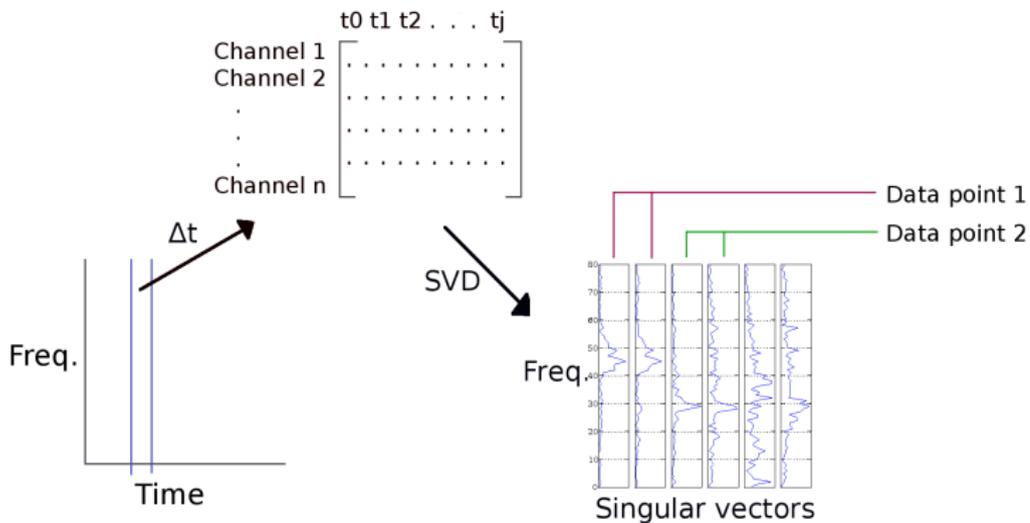
clustering in n-dimensional phase space

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# Preprocessing

- 1) Dataset selection: choose shots with some set of Mirnovs at sufficient sampling rate.
- 2) Separate and filter modes, map to  $\Delta\phi$ -space



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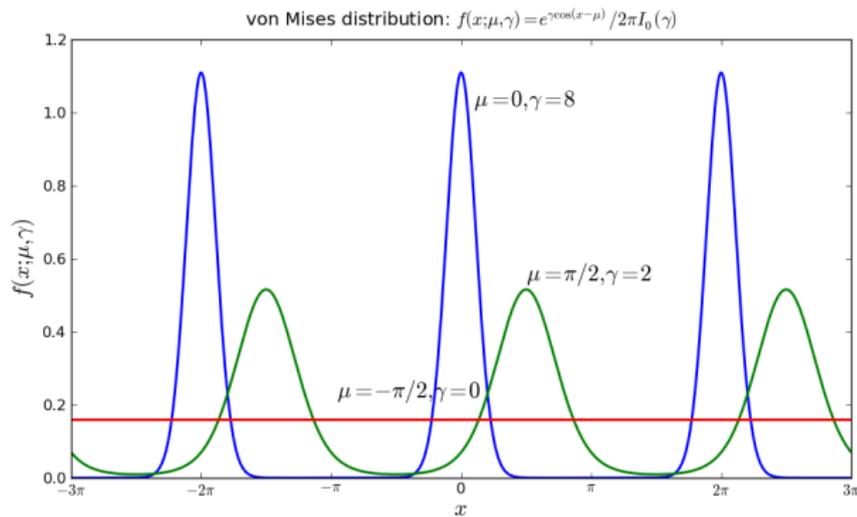
# Clustering 1

- Common approach, *Expectation Maximisation* algorithm, with Gaussian clusters.
- But,  $\Delta\phi$  is periodic.  $\Delta\phi_i = \Delta\phi_i \pm 2n\pi$
- Solution 1: use  $\sin(\Delta\phi)$ ,  $\cos(\Delta\phi)$ , however Gaussian approximation is poor as  $\sin, \cos \rightarrow \pm 1$ .

# Clustering 2

- von Mises distribution is periodic analogue of Gaussian:  

$$f(\Delta\phi; \mu, \gamma) = \frac{e^{\gamma \cos(\Delta\phi - \mu)}}{2\pi I_0(\gamma)}$$
.  $I_0(\gamma)$  is Bessel function of order 0.
- Bessel function complicates EM
- Now using Minimum Message Length (MML) clustering with von Mises distributions.



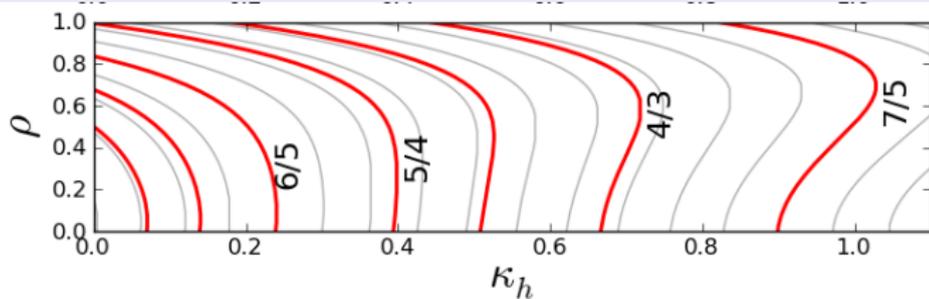
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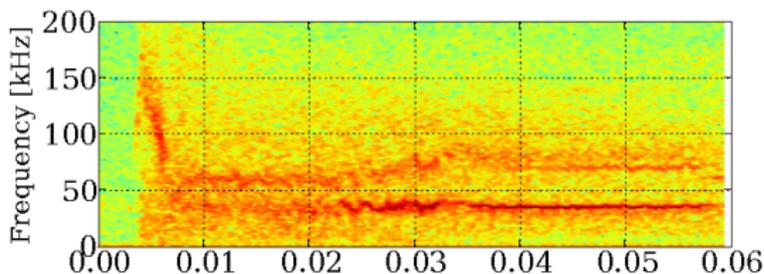
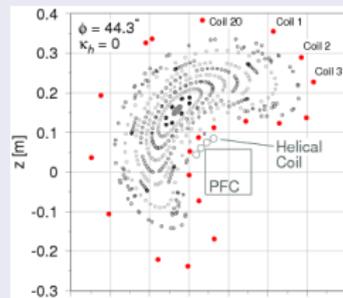
H-1

# Rotational transform ( $t$ ) parameter scan

## Configuration scan

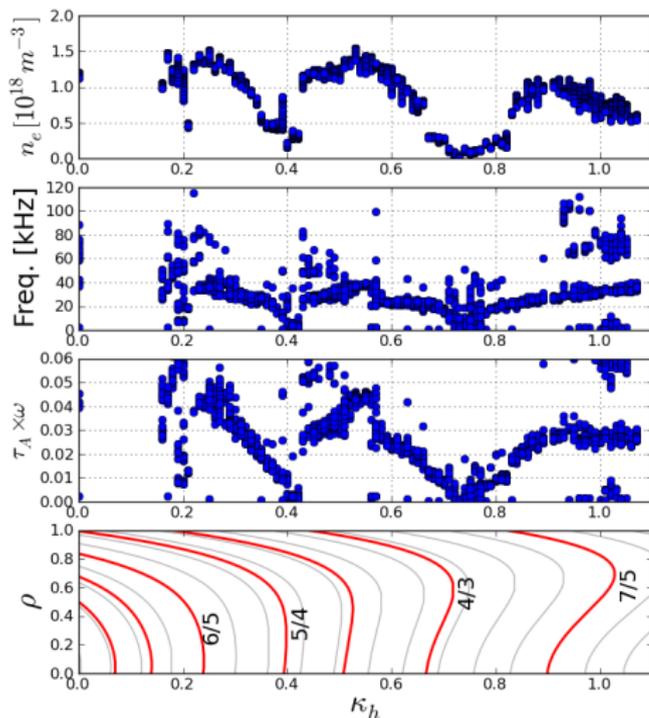


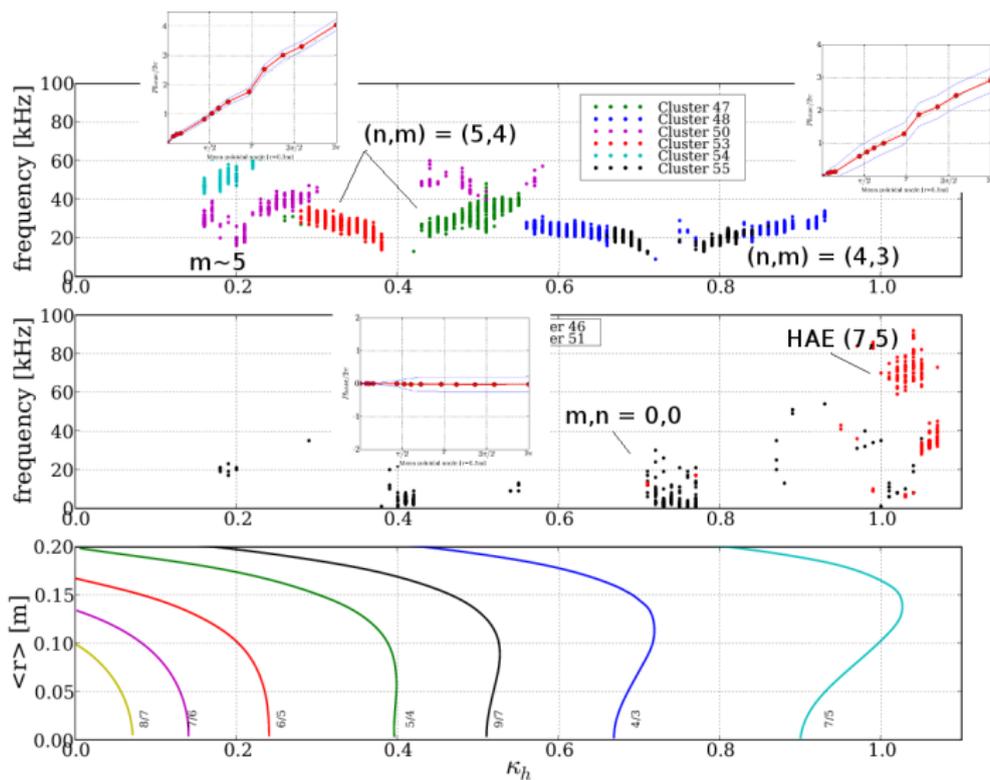
## Mirnov coils



# H-1 $t$ -scan, data mining

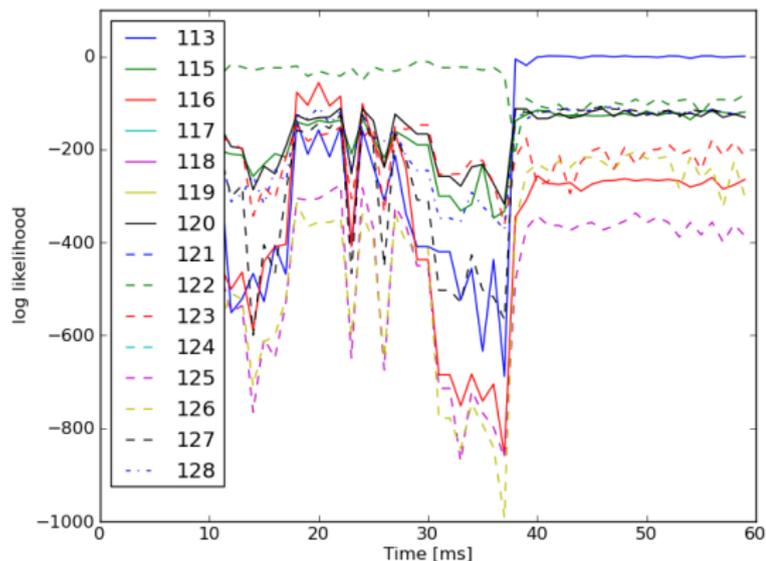
- Each data point from 1 ms time sample, each  $\kappa_h$  coordinate is a separate shot.
- Scaled by Alfvén transit time  $\tau_A = R_0 \sqrt{\mu_0 \rho} / B_0$ , the resonance modes are GAE-like ( $f \propto |n - \ell m|$ ), but at 1/3 frequency.



H-1  $t$ -scan, results

# H-1, probabilistic mode identification

Because modes are represented as multivariate von Mises distributions, we can trivially compute the likelihood of any new data being of a certain type of documented mode.

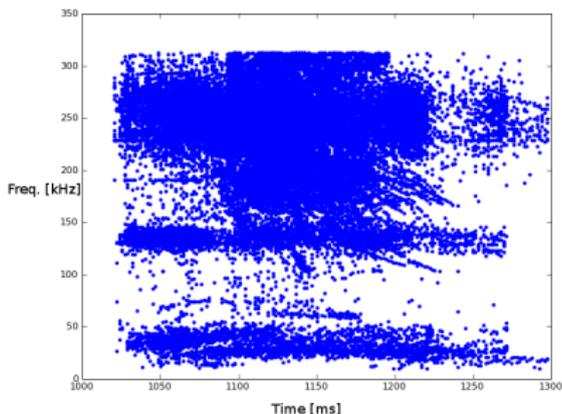


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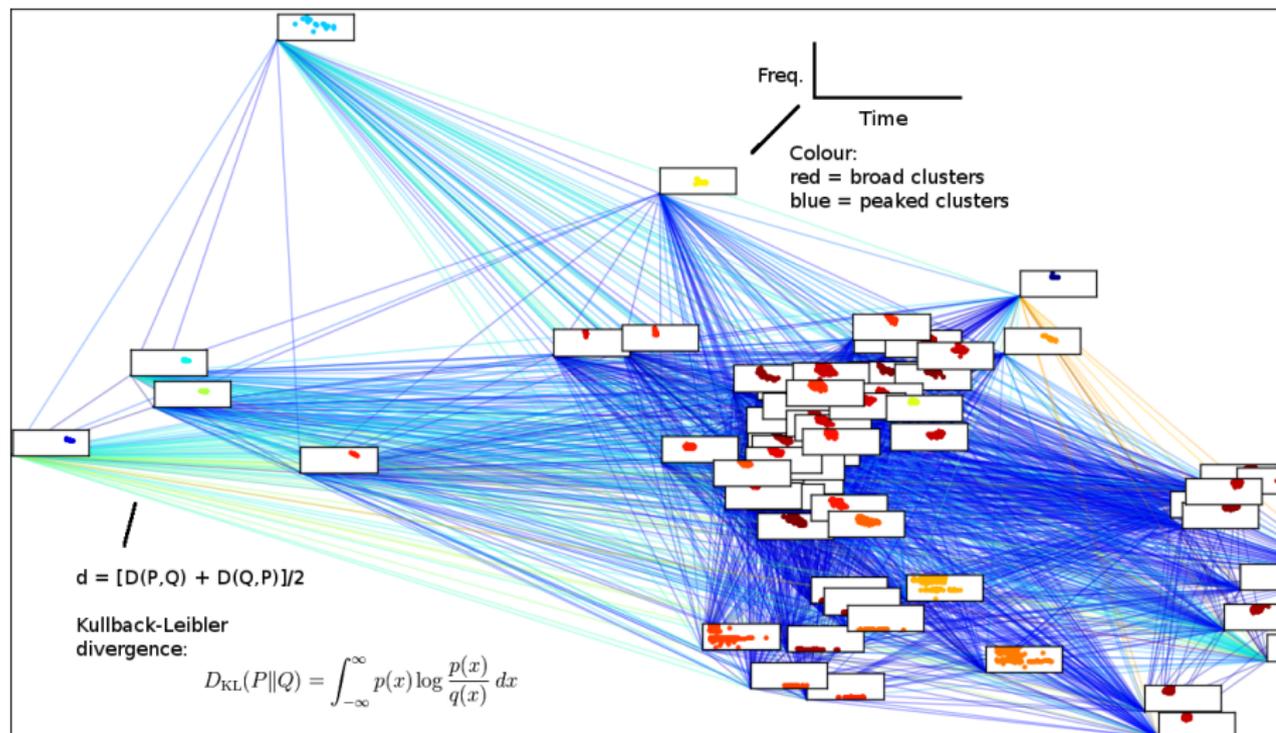
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# TJ-II Dataset

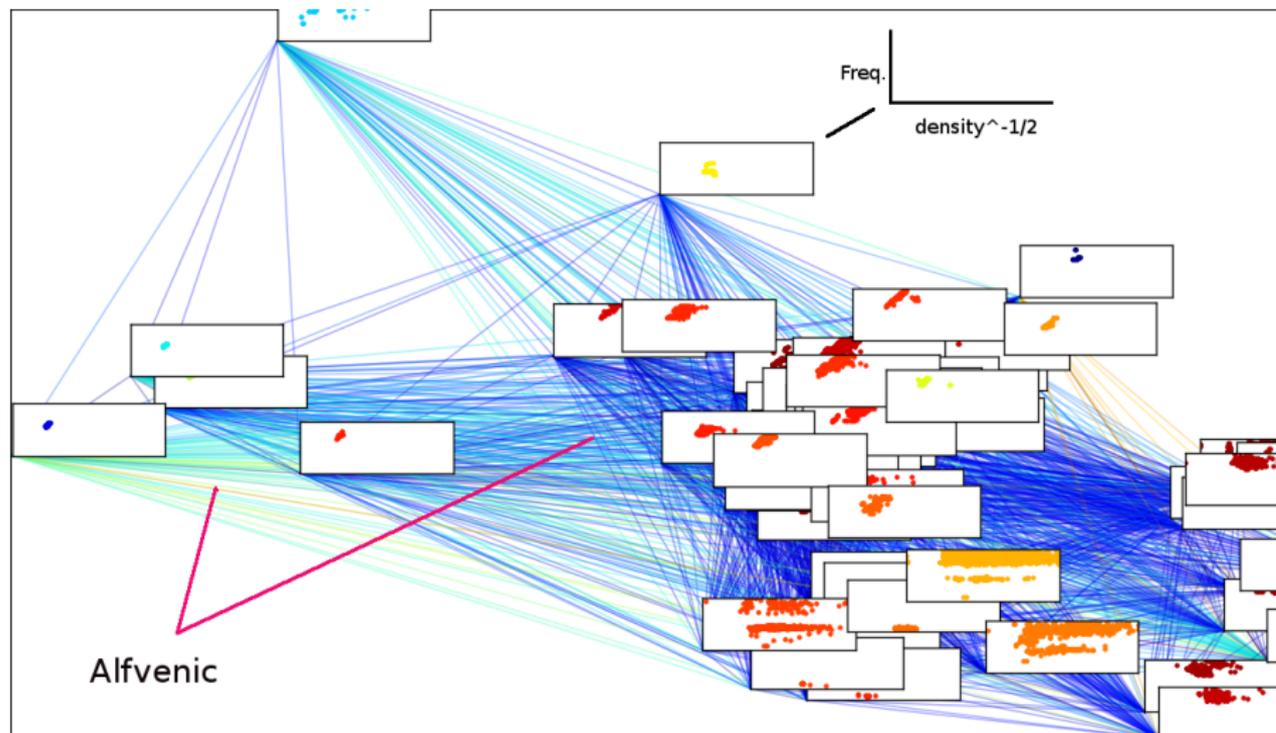
- Several datasets (different sets of Mirnovs) have been considered in TJ-II. Largest portion of database: 15 poloidal Mirnov coils.
- 3753 shots.  $7.1 \times 10^6$  datapoints  $\rightarrow 3.2 \times 10^4$  after filtering ( $f > 2$  kHz,  $p > 0.3$ ,  $S_{Mir1}(RMS) > 1.2$ )



## TJ-II Clusters showing time-freq.

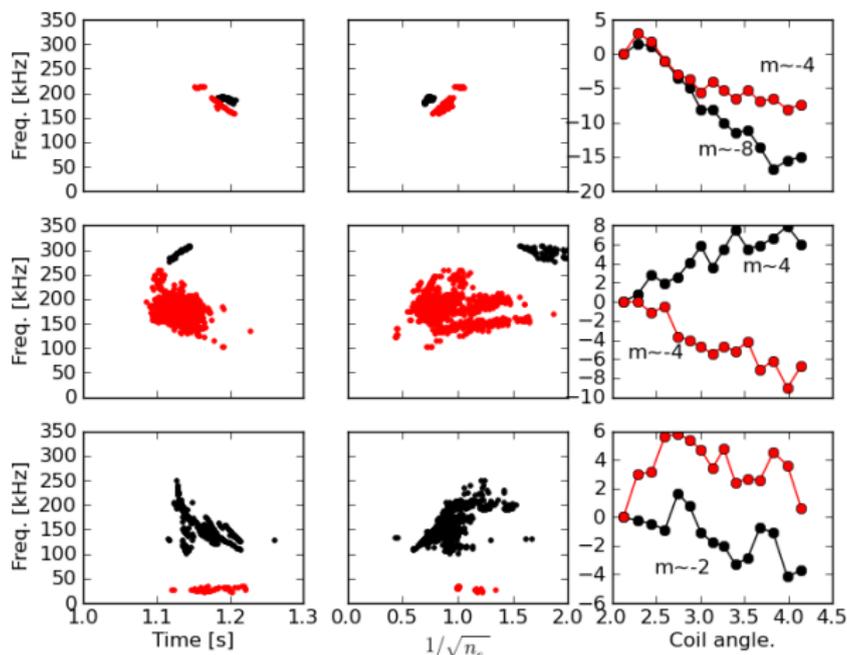


TJ-II

TJ-II Clusters showing  $n_e^{-1/2}$ -freq.

# TJ-II Mode structure

Various TJ-II modes. Phase-angle defined by von Mises clusters.



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# Heliotron J dataset

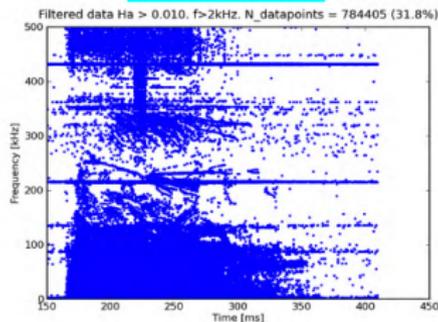
- 14 Mirnov coils with same timebase → 3786 shots
- Using 1ms samples, get 2.5 million datapoints (incl. pre-, post-shot noise etc.)



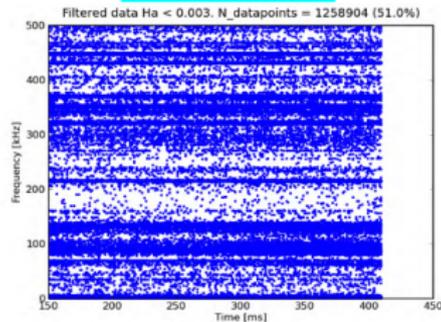
# Heliotron J, filtering

Remove non-plasma signal noise by keeping only time-slices where  $H_{\alpha} > 0.01$

**Ha filtered data**

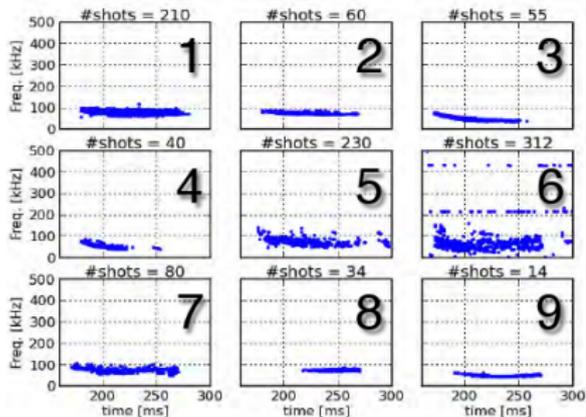


**Removed "noise"**

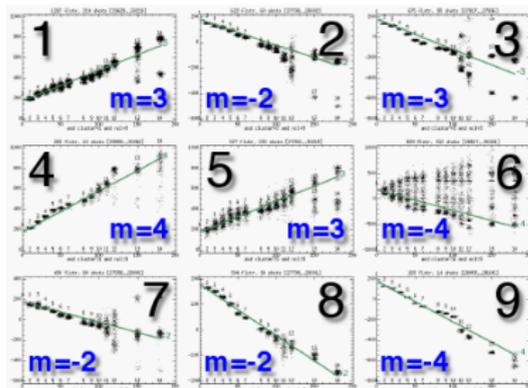


# Heliotron J results

## clusters

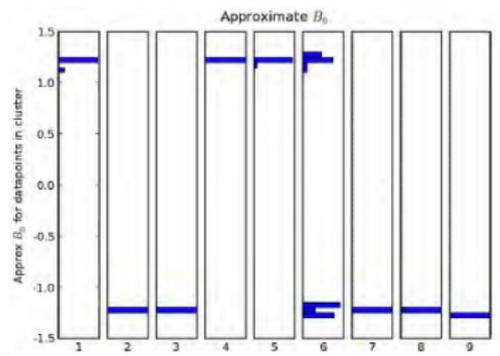
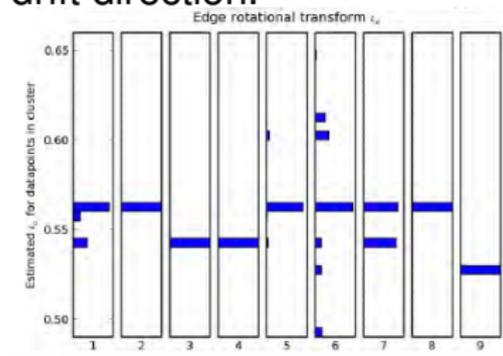


## mode analysis



# Heliotron J, mode parameters

$B_0$  flips phase-angle plot: modes propagate in ion diamagnetic drift direction.



# Summary

- We have a working, scalable method for extraction and unsupervised classification (clustering) of coherent MHD activity in fusion databases.
- Clustering results have been used to investigate physical properties of the modes.
- We have demonstrated using clustering results for probabilistic identification of modes in new data.

# Status and Plans

## H-1

Use new Mirnovs. Explore mass,  $B$  dependence of modes.

## Heliotron J

Expand database. Identify low, broad frequency MHD activity

## TJ-II

Find dependence of modes on other parameters, diagnostics.

## LHD/CHS

Status: Dedicated data mining server installed at NIFS.

## W7-AS

Status: Viable range of shots selected, working on connection between pyfusion and new W7-AS webservices interface.