



Comprehensive Analysis of Micro-Bunching Instabilities using Machine Learning

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Outline



Motivation

- Introduction of the Clustering Method k-means
- Pre-Processing
- Analysis of Micro-Structure Dynamics
- Analysis of Further Characteristics
- Outlook



- operation of storage rings with short electron bunches increases coherent synchrotron radiation (CSR) power
- leads to micro-structure dynamics within the bunch
- indirect measurement: resulting fluctuations in the emitted CSR power
- direct measurement: electron distribution, difficult due to the small scale of the micro-structures
- \Rightarrow simulation of longitudinal dynamics with the simulation code Inovesa

Schönfeldt, P. *et al.* Parallelized Vlasov-Fokker-Planck solver for desktop personal computers. *Phys. Rev. Accel. Beams* **20** (2017)

Motivation Measured Fluctuations of the CSR power (Bursting)



 \Rightarrow fluctuations occur with characteristic frequencies

Data taken at ANKA, courtesy of Miriam Brosi



Motivation Measured Bursting Spectrogram





Data taken at ANKA, courtesy of Miriam Brosi



Motivation Simulated Bursting Spectrogram





- identify micro-structures on the simulated longitudinal bunch profiles
- correlate findings to the emitted synchrotron radiation, i.e. the CSR power spectrogram
 - ⇒ large amounts of data to analyze!
 (151 bunch currents with 10 000 time steps each,
 i.e. 1.5 million bunch profiles in the simulation data set)
 - ⇒ application of the machine learning technique *k*-means to reveal micro-structures within a fixed bunch current

Clustering method k-means

Principle of the k-means algorithm



the data set



initialization



iter. #1: assignment



iter. #1: update



iter. #2: assignment



iter. #2: update



Clustering method k-means



Introduction

- clustering is an example of unsupervised learning, a sub field of machine learning
- can be used as an exploratory tool of data analysis to reveal underlying structure
- the *k*-means method is one of many different clustering algorithms
- relatively simple procedure which can be understood intuitively
- aims to achieve an appropriate categorization of the given data set
- number of categories or clusters *k* is a free parameter
- cluster means can be used as reasonable representatives to analyze the found clusters

Pre-Processing



Initial Application of k-means



Pre-Processing



After Re-Centering Procedure



Analysis of Micro-Structure Dynamics

Analysis of Micro-Structure Dynamics



Different Bursting Regimes





Cluster Centers, $I_{reg} = 0.88 \text{ mA}, k = 2$





Cluster Centers Ref. to Mean, $I_{reg} = 0.88 \text{ mA}, k = 2$





Cluster Centers Ref. to Mean, $I_{reg} = 0.88 \text{ mA}$, k = 4





Correlation to CSR Power, $I_{reg} = 0.88 \text{ mA}, k = 2$





Correlation to CSR Power, $I_{reg} = 0.88 \text{ mA}, k = 2$





Longitudinal Phase Space, $I_{reg} = 0.88 \text{ mA}, k = 2$







Longitudinal Phase Space, $I_{reg} = 0.88 \text{ mA}, k = 2$

Analysis of Micro-Structure Dynamics



Different Bursting Regimes





Cluster Centers, $I_{saw} = 1.15 \text{ mA}, k = 4$





Cluster Centers Ref. to Mean, $I_{saw} = 1.15 \text{ mA}, k = 4$





Correlation to CSR Power, $I_{saw} = 1.15 \text{ mA}, k = 4$





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Correlation to CSR Power, $I_{saw} = 1.15 \text{ mA}, k = 4$





Longitudinal Phase Space, $I_{saw} = 1.15 \text{ mA}, k = 4$







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Modulation Frequencies for all Bunch Currents





Modulation Frequencies for all Bunch Currents





Modulation Amplitudes for all Bunch Currents





Modulation Amplitudes for all Bunch Currents





Correlation of Modulation Frequency and Amplitude

- f_{mod} > 75 GHz colored red
- same color scheme applied to modulation amplitudes A_{mod}



Cluster Label Spectrogram



Spectral Analysis of Categorical Time Series



Stoffer, D. S. *et al.* Spectral analysis for categorical time series: Scaling and the spectral envelope. *Biometrika* **80**, 611–622 (1993)

Analysis of Further Characteristics

Dynamics of a Localized Charge Density



Charge Density at CoM Position, $I_{reg} = 0.88 \text{ mA}$

- analysis of temporal changes of the charge density at a fixed position within the electron bunch
- the gained time signal resembles the corresponding CSR power signal



Dynamics of a Localized Charge Density



Spectrogram for CoM Position



Dynamics of a Localized Charge Density



Spectrogram for a Position far away from CoM



Low Current Bursting



2nd Data Set: CSR Power Spectrogram



Low Current Bursting



Longitudinal Phase Space, $I_{low} = 20 \,\mu A$, k = 2

- micro-structures in the low current bursting regime look similar as well
- however, there is one structure less than for the regimes above!



Vacuum Chamber Height



Ref. Cluster Centers, $I_{\text{bunch}} = 1.4 \text{ mA}$, k = 2, g' = 0.5 g



Vacuum Chamber Height



Long. Phase Space, $I_{\text{bunch}} = 1.4 \text{ mA}$, k = 2, g' = 0.5 g





Outlook

Outlook Ideas and Open Questions



- modulation frequencies are very similar across different bunch currents, but show a slight decay
 - \Rightarrow Where does this come from?
 - \Rightarrow Does it maybe depend on changes in the bunch length?
- systematic studies of the *f*_{mod} dependence on different parameters,
 e.g. the vacuum chamber height or bending radius
- reproduce these results on measured data (EO setup)
- additional ideas/suggestions?

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Thank you for your attention!

Backup

Simulation Parameters



1st Data Set

Physical parameter	Value
RF voltage U_0 revolution frequency f_{rev} synchrotron frequency f_s damping time τ_d harmonic number h parallel plates distance g initial electron distribution $\varphi(z, E, t_0)$ simulation time t	1 MV 9 MHz 30 kHz 5 ms 50 3.2 cm 2-dim. Gaussian 250 $T_{\rm s}$
bunch current Ibunch	0.5 mA to 2.0 mA
Control parameter	Value
grid size n _{grid} time steps n _{steps}	256 10 000

Simulation Parameters



2nd Data Set

Physical parameter	Value
RF voltage U_0 revolution frequency f_{rev} synchrotron frequency f_s damping time τ_d harmonic number h parallel plates distance g initial electron distribution $\varphi(z, E, t_0)$ simulation time t	1.3 MV 2.72 MHz 5.76 kHz 2.66 ms 184 3.2 cm 2-dim. Gaussian 500 <i>T</i> _s
bunch current <i>I</i> _{bunch}	$0.5\mu\text{A}$ to $200\mu\text{A}$
Control parameter	Value
grid size <i>n</i> _{grid} time steps <i>n</i> _{steps}	256 10 000

Micro-Structures on Energy Profiles



Cluster Centers Ref. to Mean, $I_{reg} = 0.88 \text{ mA}, k = 4$





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