



# Long Short-Term Memory Models for Online Monitoring of Beam Phase and Current

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## **Introduction (CAFe)**



### CAFe: China ADS Front-end (SRF facility, in CW mode)

- Demonstrate the feasibility of the 10-mA high-power CW proton beam for the China-initiative ADS (CiADS) project
- In March 2021, CAFe achieved its design goal with the successful commissioning of a 10 mA, 205 kW CW proton beam at an energy of 20 MeV





## **Introduction (ESS)**





Ion Species	Protons	
Output Energy	2	GeV
Frequency	352.21/704.42	MHz
Pulse Length	2.86	Ms
Peak Current	62.5	mA
Protons per Pulse	1.1 x 10 <sup>15</sup>	
<b>Repetition Rate</b>	14	Hz
Duty Cycle	4	%
Average Beam Power	5	MW
Accelerating Structures	RFQ, DTL, SC	
	Spokes/Elliptical	
Accelerator Length	~365	m

#### The normal-Temperature Front End



	Buncher1	Buncher2	Buncher3
R/Q[Ω]	73.85	76.7	76.9
$Q_L$ [arb.units]	8878	8893	9470
<i>f</i> <sub>0.5</sub> [kHz]	19.83	19.80	18.6
$f_{RF}$ [MHz]	352.21	352.21	352.21
$\beta$ (v=c) [arb.units]	1.03	1.02	1.06





## **Limitations of Traditional Phase scan Methods**

- synchronous phase and its measurements
- To maximize the energy gain, the particle must enter the cavity at a specific point in the RF field's oscillation, which corresponds to the synchronous phase  $(\varphi_b)$ .

 $V_{acc} = V_c \cdot \cos(\varphi_b)$ 

• The measurement of the synchronous phase is performed by the BPM.

#### Limitations

- 1. Off-Line: Takes up machine operation time
- 2. Phase Drift: hard to track phase drift caused by environmental factors
- **3. Long Scanning Times**: Inefficient for large accelerators with numerous RF cavities.



Phase drift caused by temperature changes in CAFe





#### **Proposed solution: on-line beam measurement**

- It can address the challenges of traditional BPM based phase scan method to determine beam synchronous phase <u>and beam current</u>.
- It can provide continuous beam phase monitoring and online beam information during accelerator operation. It can solve two major problems in accelerator operation:
  - 1. Beam Loading and its compensation
  - 2. Beam Trip and Its Recovery





### **On-line measurement during operation:**

#### Beam Loading and its compensation

- Beam loading is very hard to fully compensate in normal conducting linac, which affects both beam transmission, and beam phase (more than 10 deg beam phase changes observed when adaptive feedorward is not on (feedback only) for RFQ)
- Static feedforward is necessary, but knowing beam information in real time (beam current, beam on/off status) is hard
  RFQ Cavity Field with Beam Loading under Feedback & Feedforward



The beam loading of 59mA in ESS RFQ





### **On-line measurement during operation:**

#### Beam Trip and Its Recovery:

• detect beam trip, disclose beam information (status, phase, current, energy gain, etc) is essential for preventing overshoot due to no beam (AFF still on).

Overshoot at beam on	62.5mA	63.5mA
RFQ	5%	5%
DTL1	9%	11%
DTL2	8%	8%
DTL3	8%	8%
DTL4	7%	7%













### **Relationship between RF and Beam**

- Beam loading effect: the influence of the charged particle beam on the accelerating RF field within an RF cavity. When a beam of charged particles (like electrons or protons) passes through the cavity, it induces currents and fields that interact with the existing RF field, which can change the amplitude and phase of the RF field.
- Mechanism: The changes measured with respect to the RF field depend on the beam phase, charge and the accelerating field strength.



[1] P Pawlik et al, New method for beam induced transient measurement, Meas. Sci. Technol. 2007(18)
 [2] Nick Walker, Experience with beam-transient LLRF calibration in XFEL, DESY-TEMF collaboration meeting, 2020





#### ESS Buncher2 : open loop (this is also used in DESY and KEK)



**Open loop**:  $I_b = 63$ mA, Beam phase from  $-180^{\circ}$  to  $180^{\circ}$ , Beam width time= $5\mu s$ 



## **Transient Beam-loading Method**



#### Bottlenecks: Requires the cavity works w/o detuning (for Bun2) and under open-loop



#### Application Trans. Method at ESS Buncher2 : closed loop

The vector  $V_{tr}$  is **NOT** straight under closed-loop operation



## **Transient Beam-loading Method**



### **Closed-loop vs. Open-loop: Nonlinearity becomes increasingly complex**



#### Challenges faced by the transient beam-loading method when it comes to closed-loop:

**1.Generalization Issue:** current methods struggle with adapting to different operating conditions:

Open-Loop: when the system is not in a feedback loop.

Closed-Loop: when the system is in a feedback loop.

Detuning: when there are frequency mismatches or detuning.

**2.Accuracy Issue:** various nonlinear effects greatly increase the complexity of beam behavior, leads to a decrease in the accuracy of the physical methods. (space charge effects, beam emittance, nonlinearity of the accelerating fields).











### **Advantages of AI:**

Powerful nonlinear modeling ability; Rich function approximation ability; Automatic feature extraction ability; Adaptability to various complex data types

### How can AI technology be applied to beam measurements?

- Model design
- Data prepartion
- Training





![](_page_15_Picture_2.jpeg)

### How can AI technology be applied to beam measurements?

![](_page_15_Figure_4.jpeg)

![](_page_16_Picture_0.jpeg)

## **AI-based beam measurements**

![](_page_16_Picture_2.jpeg)

#### How can AI technology be applied to beam measurements?

![](_page_16_Figure_4.jpeg)

![](_page_17_Picture_0.jpeg)

![](_page_17_Picture_2.jpeg)

### How can AI technology be applied to beam measurements?

![](_page_17_Figure_4.jpeg)

![](_page_18_Picture_0.jpeg)

## **AI-based beam measurements**

![](_page_18_Picture_2.jpeg)

#### How can AI technology be applied to beam measurements?

![](_page_18_Figure_4.jpeg)

![](_page_19_Picture_0.jpeg)

![](_page_19_Picture_2.jpeg)

#### How can AI technology be applied to beam measurements?

- Model
- Data
- **Training** Pretraining: Using simulated data. Fine-tuning: Using real-world data.

#### **Performance Check**

![](_page_19_Figure_8.jpeg)

#### From simulation to real accelerator with Transfer learning: robust and reliable

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## **AI-based beam measurements**

![](_page_20_Picture_2.jpeg)

![](_page_20_Figure_3.jpeg)

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## **Challenges**

- Quantity and Quality of Data
- Simulation error: Using simulated data in the absence of experimental data, but the simulated data has biases.
- Signal-to-noise ratio in measured data: data noise, when the beam current is low, the noise heavily impacts calculation reliability

![](_page_21_Figure_7.jpeg)

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## Challenges

#### Model Generalization

• Al models may struggle to generalize in new cavities or conditions, necessitating continuous updates and validation to ensure reliability under varying operational conditions.

![](_page_22_Figure_6.jpeg)

### Model Interpretability

• "black box" nature of AI models, makes it difficult to understand their decision-making processes, which may affect their credibility in high-risk applications.

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#### **Benefits of AI Online Measurement**

- 1. High Accuracy: Provides precise measurements and strong generalization ability.
- 2. Non-Intrusive: Measures beam information without disturbing accelerator operations.
- **3. Continuous Monitoring:** Ensures stable and reliable performance. Real-time updates on beam status, phase, current, and energy gain.
- 4. Data-Driven: It can be applied to other accelerators to enhance their operation.

5. .....

#### **Challenges of AI algorithm**

- 1. Data Scarcity : requires large amount of high quality experimental data. Data scarcity limits the accuracy and applicability of the models.
- **2. Generalization Ability :** Al models may struggle to generalize in new environments or conditions.
- **3. Model Interpretability:** "black box" nature of AI models, makes it difficult to understand their decision-making processes.
- 4. Integration of Interdisciplinary Knowledge: requires the integration of knowledge from physics, engineering, and data science.
- 5. .....

![](_page_24_Picture_0.jpeg)

![](_page_24_Picture_1.jpeg)

# Thanks for your attention

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