Optimising the NLK Injection using Reinforcement Learning

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Introduction: 4 Kicker bump



Figure 1: 4 Kicker bump, by [5]

Disadvantages:

- affects stored beam
- might cause the loss of stored electrons

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Introduction: Non-linear Kicker Injection



Figure 2: Kicker magnet structure by [3].



Figure 3: Kicker field strength plot.

- Advantages:
 - Only small effects on stored beam
- installed, but not used device (currently lower efficient)

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Problem Description

- Task: Find optimal NLK settings
- Goal:
 - Improve injection efficiency
- Steps:
 - Choose activation round and strength of NLK



Figure 4: Kicker field strength plot.

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Intuition



Figure 5: Figure showing the septum sheet at 15mm.

- Injected electrons survive multiple rounds without a kick
- But eventually crash into the septum sheet



Figure 6: Example of phase space. Blue first round, red second, green third and yellow fourth round. Modified from Lin Liu et al. [4]. Image was mirrored.

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Number of rounds electrons survive without kicker



Figure 7: Plot that shows the amount of rounds newly injected electrons survive without kick.

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Model setting

- In each round from 0, . . . , 1000:
 - Input: x, px information of 1000 electrons
 - Choose if we want to activate NLK
 - Choose strength of NLK
- NLK can only be activated once
- Goal: Maximise number of electrons that survive 1000 rounds
- When is an electron lost?
 - If it hits the septum sheet, that lies at 15mm
- Fulfil setting of a Markov Decision Process

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What have we done?

Given a deterministic simulation (Thor SCSI): We have

- Understood how the NLK works
- Created own stochastic simulation (250.000 times faster)
- Applied reinforcement learning algorithms (TD3) to find the optimal actions

Steps needed for own simulation:

- Round-to-round behaviour
 - Interpolation
- NLK influence
 - More complicated...

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Accepted NLK area



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Optimal NLK strength



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Non-linear Kicker Injection Summary Programming Results

Injection efficiency given kicker strength



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How to model the NLK

- check if electron is in accepted area
- interpolate to get optimal kicker strength
- use approx. for electron survival





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Advantages of this model

- very fast simulation
- round-to-round behaviour very accurate
- can be used to pre-train RL-Agents/ find good hyperparameters

We have trained RL-Agents for 1000-step 1000-electron and 1-step 1000-electron Injection

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Best performing 1000-electron 1000-step-RL-agent



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Performance Analysis

Results:

- Only kicking in round 0 is not optimal
- Identified good regions to kick electrons
 - \Rightarrow Adjust steerer settings to steer electrons in these regions

How to apply to BESSY II?

- Create an injection-injection model
- Activation round and strength of NLK has to picked immediately
- Only small changes in electrons position/shape from injection-injection

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Model setting Programming Results

Best performing 1-step 1000-electron RL-agent



Model performance, in own env

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What have we seen?

- Advantages using NLK injection
- Creating own environment
- Results in different environments

Left to do?

• Find the optimal steerer settings before the NLK

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Apply to BESSY II

References

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