

# Activation monitoring and prediction models

2<sup>nd</sup> Slow Extraction Workshop, 9 - 11<sup>th</sup> November 2017



A.C. Araujo Martínez, D. Björkman, M.A. Fraser,  
B. Goddard, V. Kain, P.M. Schicho, C. Theis, H. Vincke, CERN

# Overview

- Why model the build-up of Induced Radioactivity?
- History of SE-induced radioactivation of SPS
- Introduction to empirical models of  $IR(t)$
- A practical implementation of  $IR(t)$ :
  - Instrumentation, data logging, fitting
- Predictive power of  $IR(t)$  :
  - Cool-down times for future operational scenarios:
    - YETS 2017-18 and future SPS BDF operation
- Conclusion:
  - Future studies and development, e.g. online tool

# Why monitor and model the build-up of IR?

- Slow extraction is a lossy, resonant process and it's efficiency is sensitive to the smallest machine parameters
  - e.g. alignment to 100  $\mu\text{m}$ , power converter ripple of a few ppm etc.

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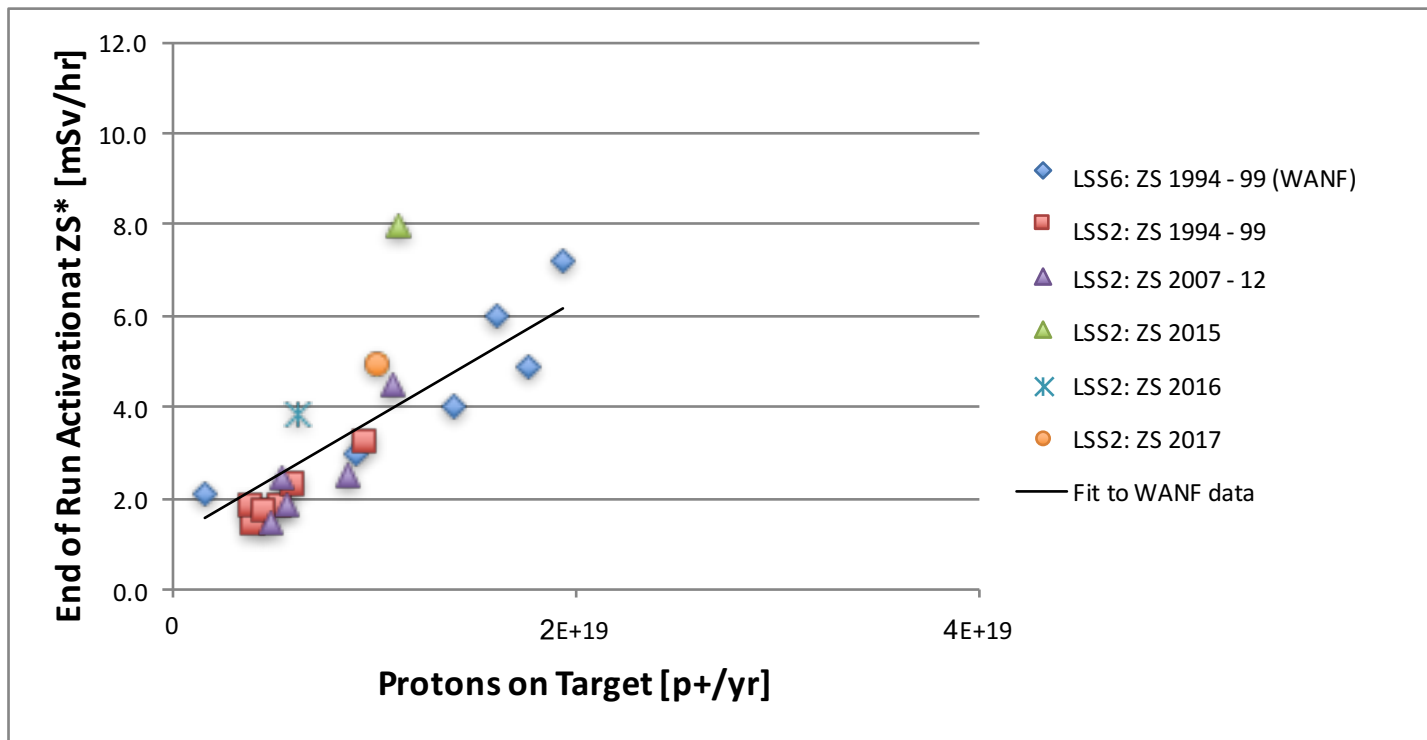
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  - We learnt this the hard-way at CERN SPS with elevated activation in 2015
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- Stable operation is not trivial and constant monitoring, warnings and interlocking are mandatory
  - We learnt this the hard-way at CERN SPS with elevated activation in 2015
  - Feedforward, or feedback OP systems, will be the solution in the longer term
- We want to monitor and empirically model the  $\text{IR}(t)$ :
  - Full numerical calculations are too computationally intensive: we are look for a simple but reliable empirical approximation
  - Keep an eye on activation levels before the end-of-year RP survey: avoid local radioactive hotspots!
  - Understand cool-down times as a function of Protons On Target (POT)
  - Predict activation levels for future operational scenarios as a function of extraction efficiency and POT

# Activation levels at ZS in LSS2

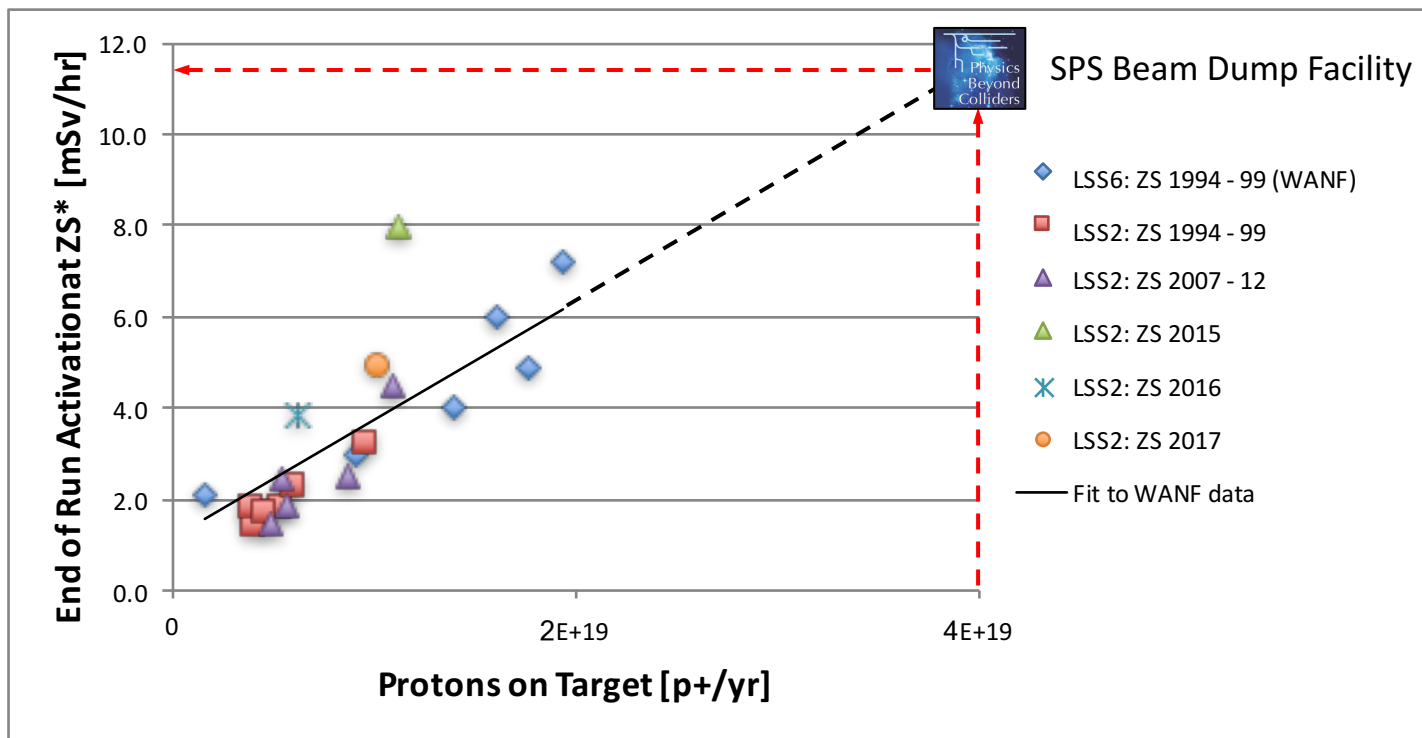
- Analysis of archived yearly end-of-run RP survey data shows a rough linear correlation of activation vs. POT:



\*Measured ~30h after shutdown at ~1 m from beam line, peak at ZS in LSS2 plotted. Correlation shown is a rough rule of thumb: time from end of last slow-extraction can vary, along with the length of the annual proton run and extraction rate. WANF data is ½-integer quad. driven “fast-slow” extraction: ~10 ms spills.

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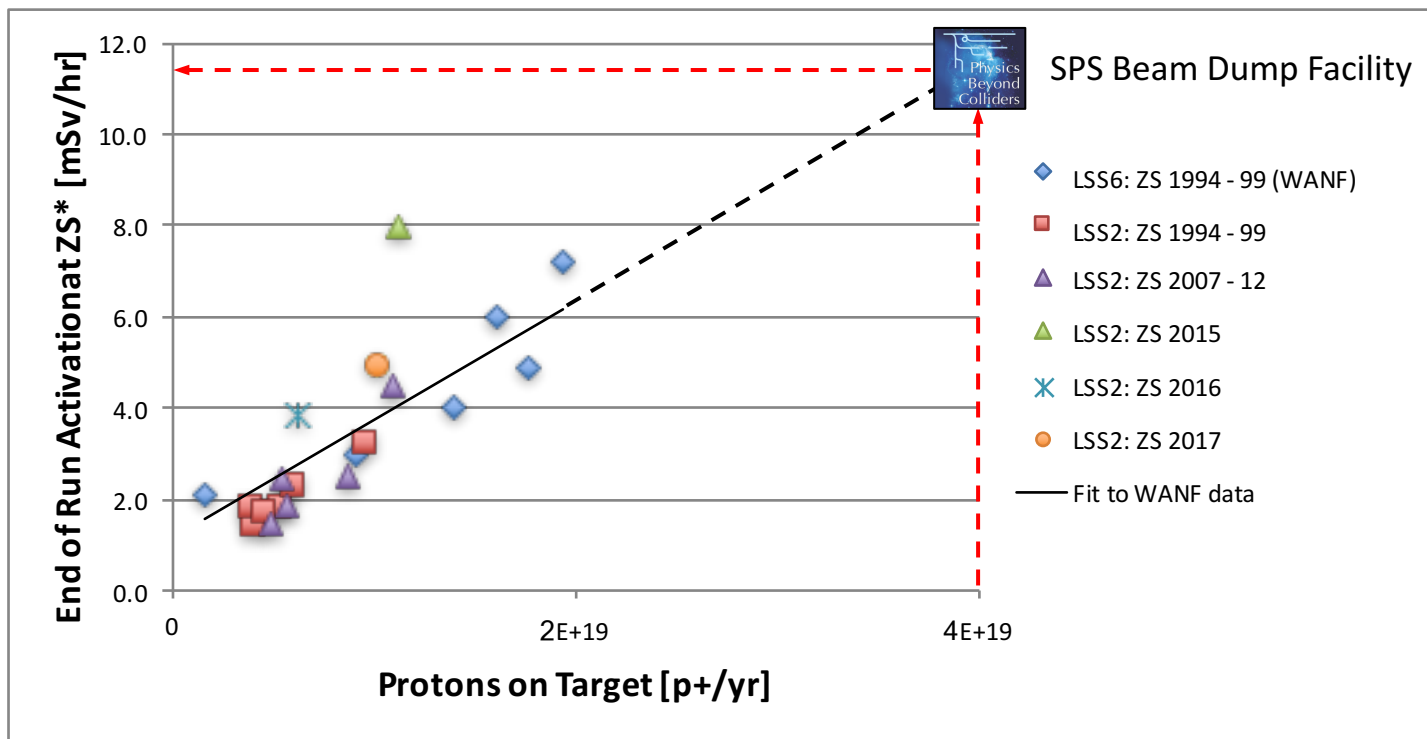
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  - i.e. background not recovering during year-end stops:

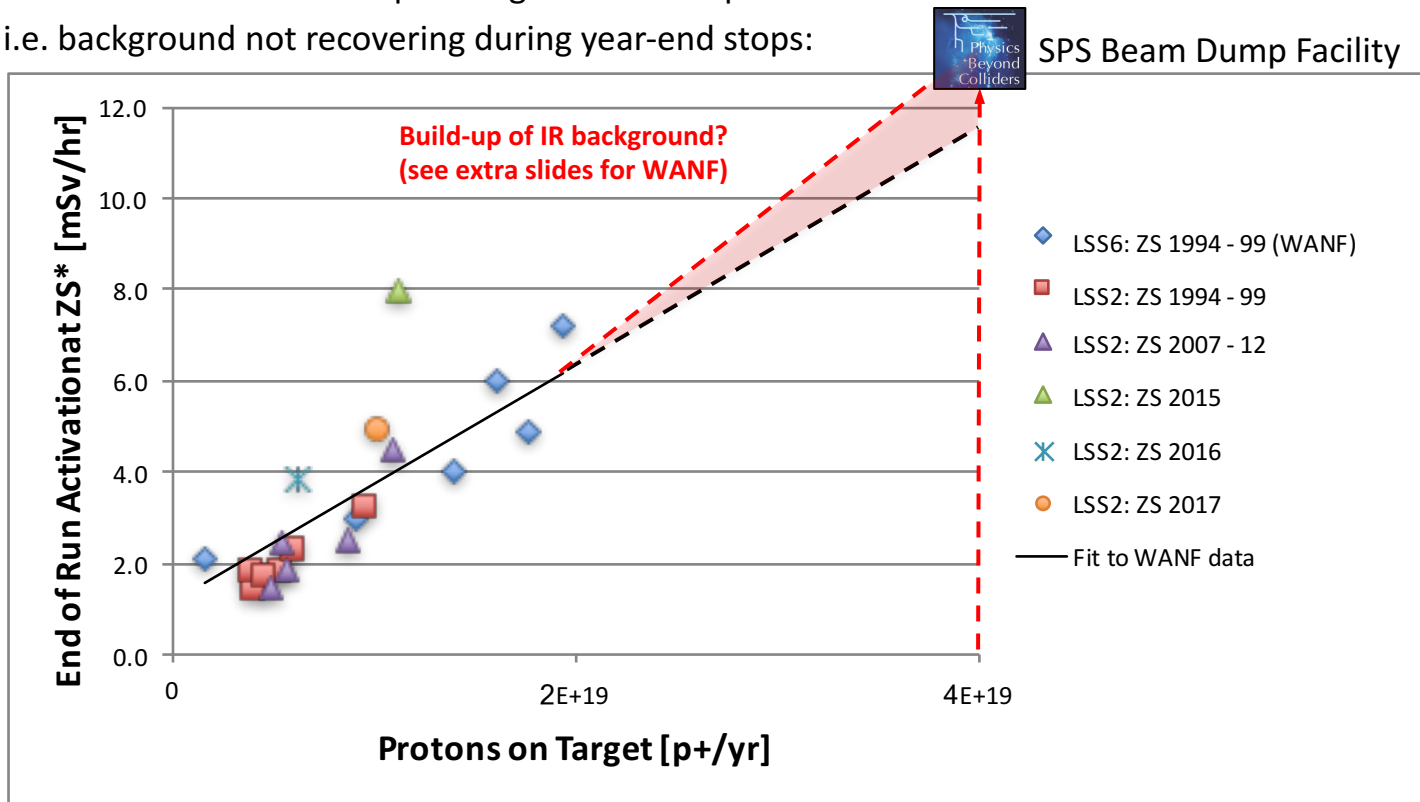


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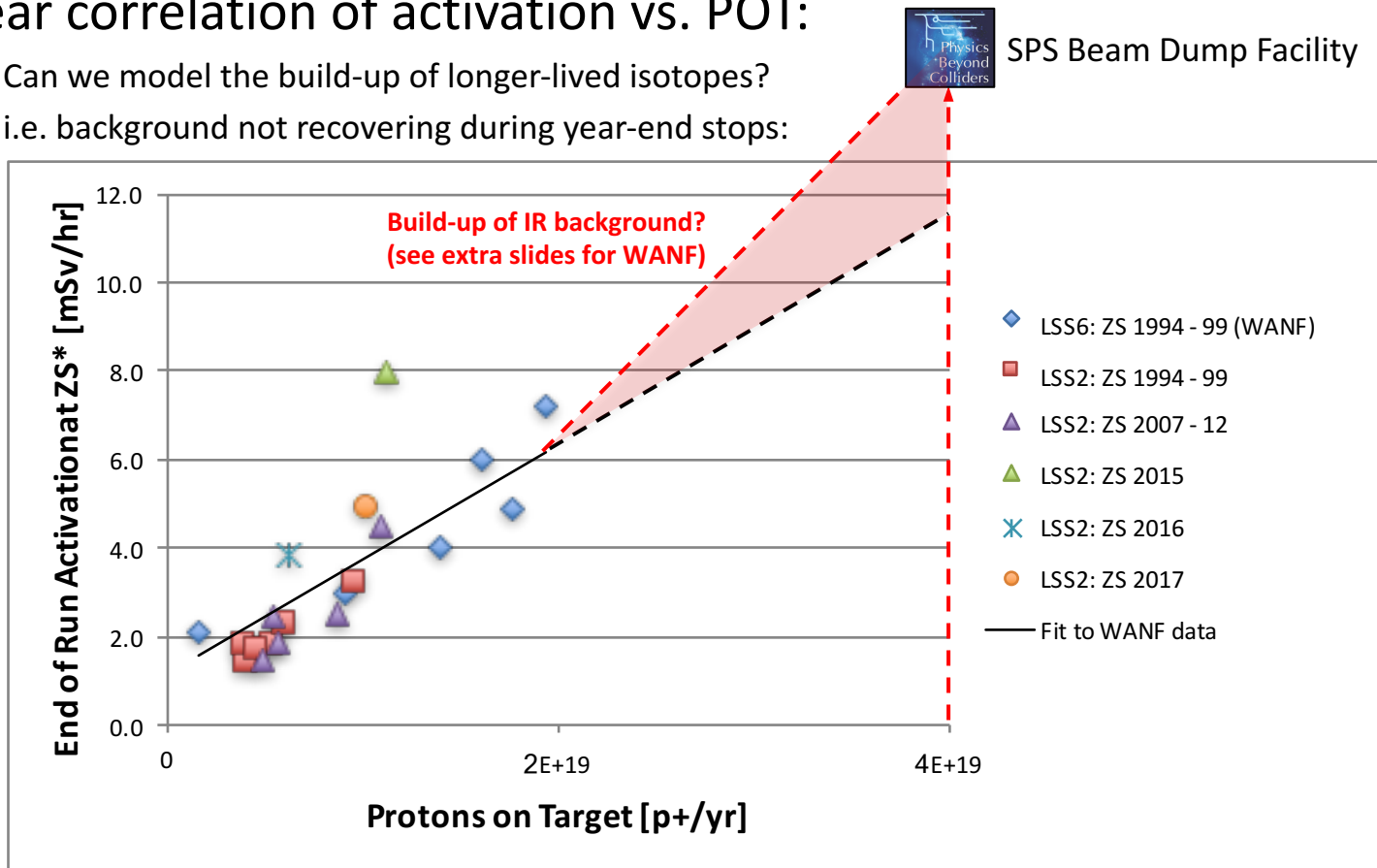
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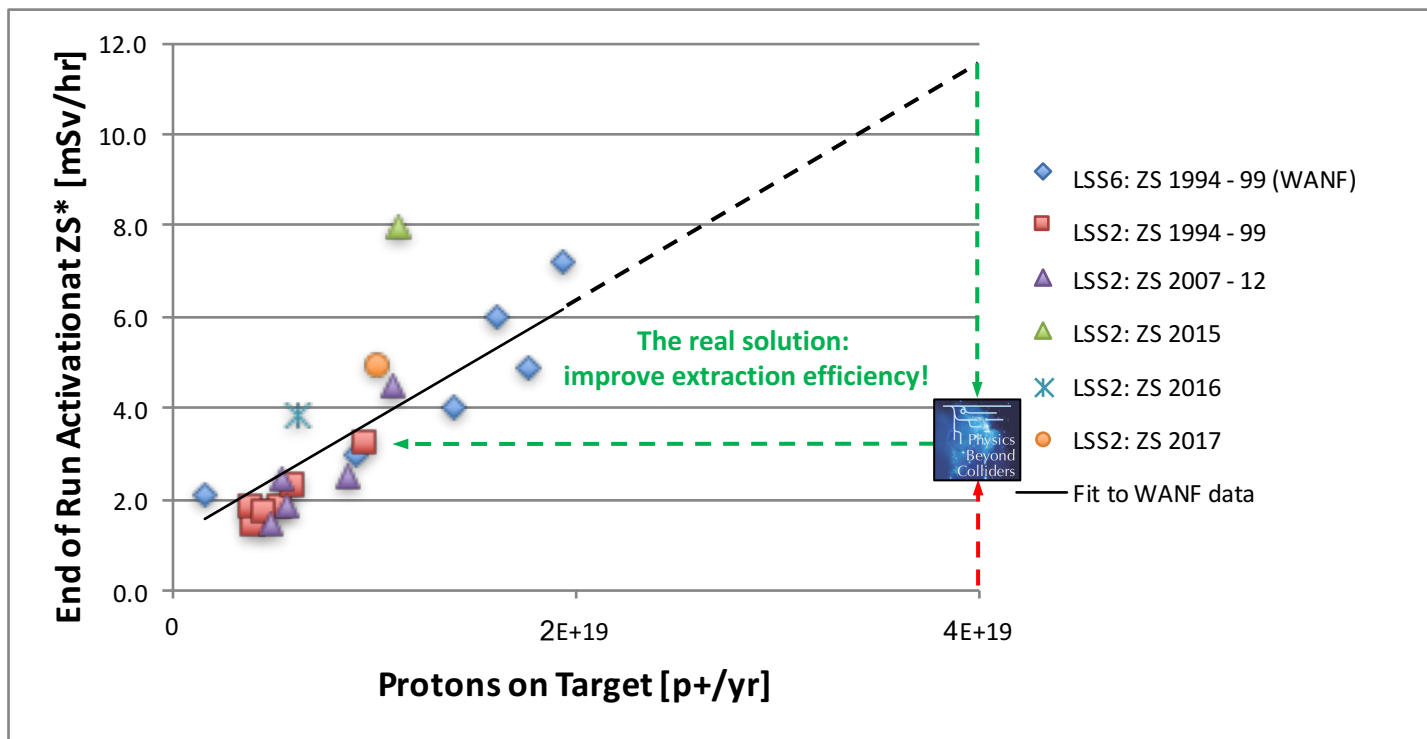
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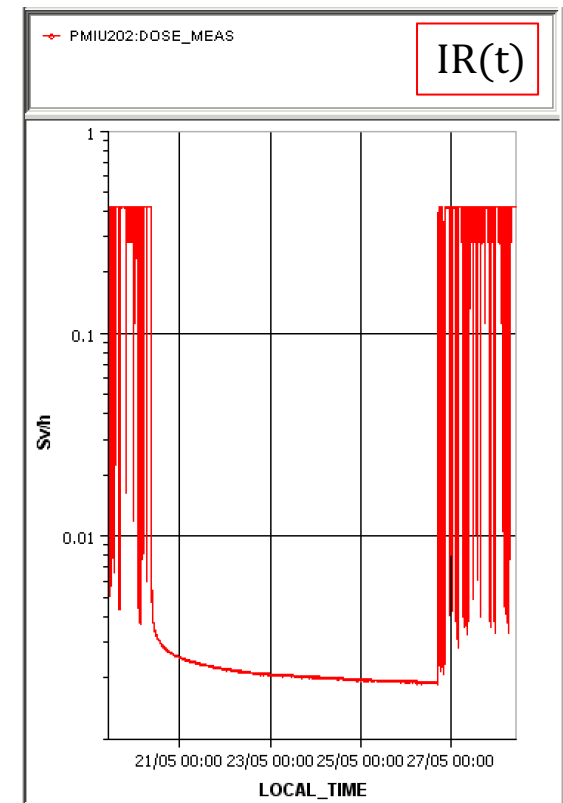
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- The effective half-life describes the non-linear time dependence of IR generated by the mixture of different radionuclides produced during the initial irradiation and in the resulting chains of radioactive decay
- $t_{\text{eff},1/2}$  increases towards  $\infty$  (stability) at an exponentially slower rate, which is what we might expect physically:  $\lim_{t \rightarrow \infty} t_{\text{eff},\frac{1}{2}} = \infty$  and  $\lim_{t \rightarrow \infty} \frac{\partial t_{\text{eff},\frac{1}{2}}}{\partial t} = 0$

# Effective half-life for different IR(t) models

- An example: difference between measured half-life next to ZS Tank 2 and different empirical models during a week long Technical Stop:

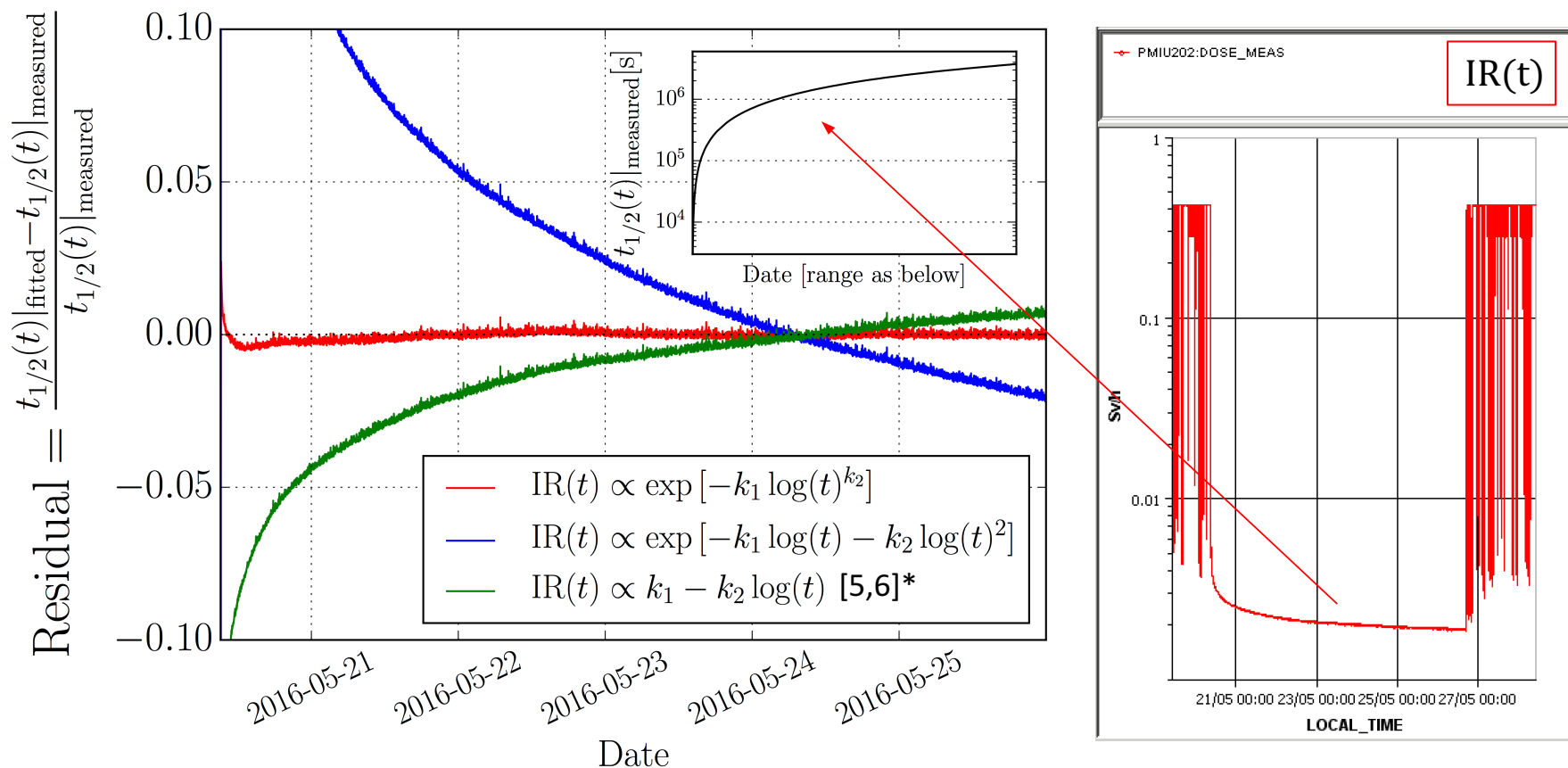


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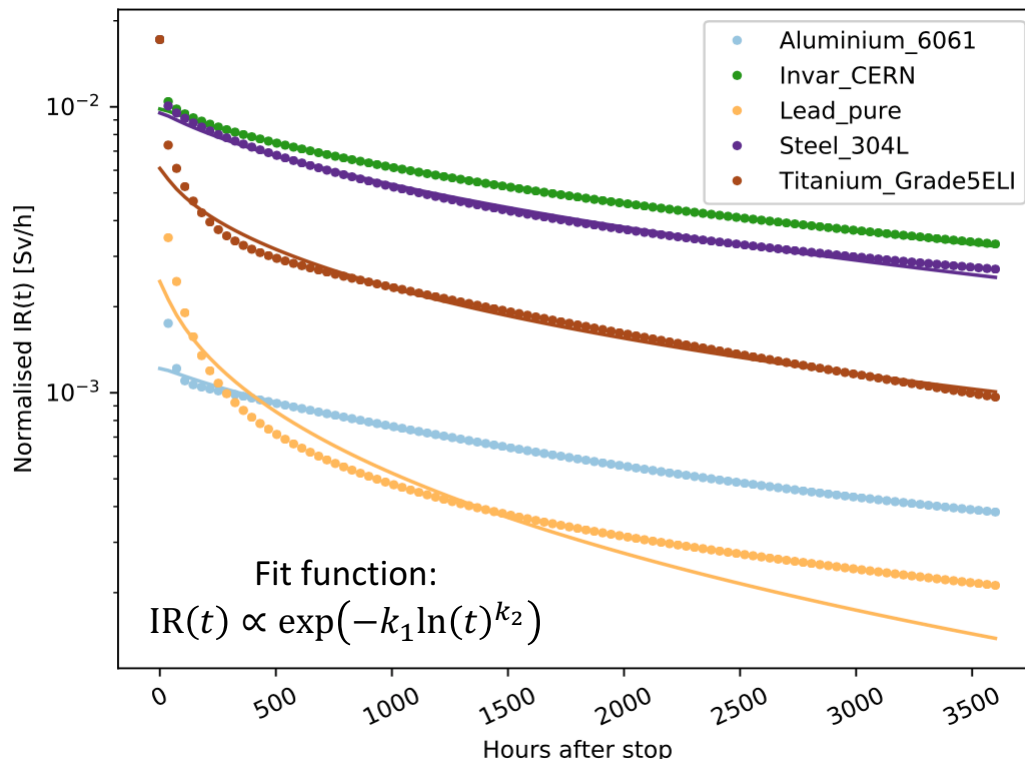
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# IR(t) vs. numerical computation, e.g. ActiWiz

- Cool-down computed numerically with *ActiWiz* [7] after bombardment of different materials with 400 GeV p+ for 200 days, POT =  $1 \times 10^{19}$ :

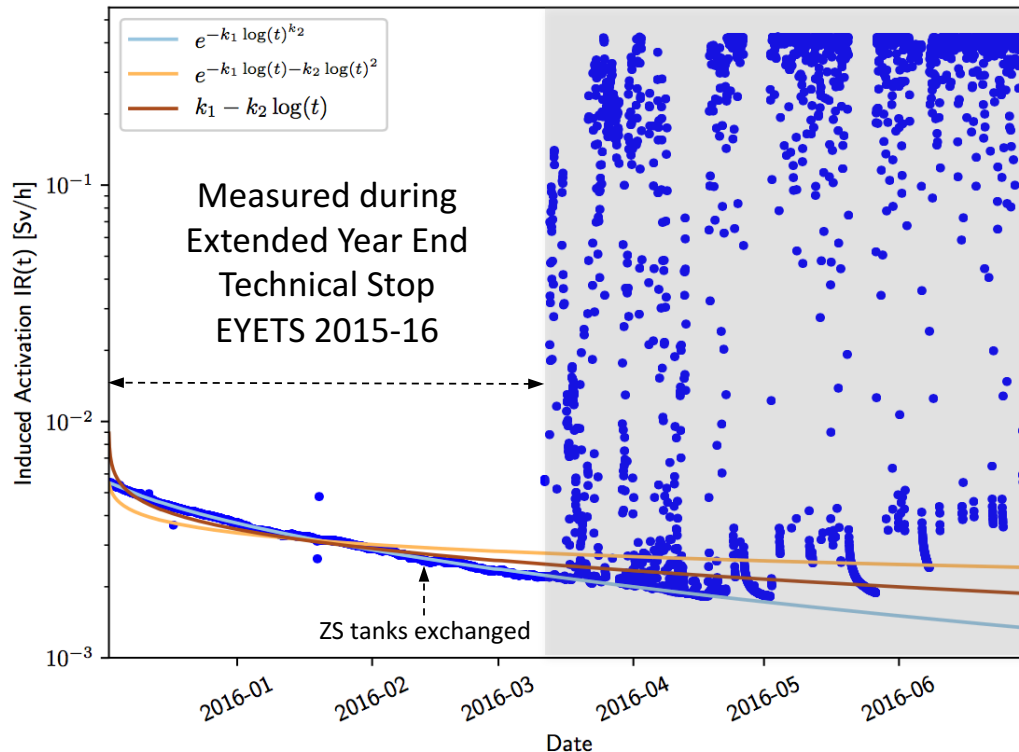


Material	Fit constants: $IR(t) \propto \exp(-k_1 \ln(t)^{k_2})$	
	$k_1$ [ $\times 10^{-4}$ ]	$k_2$
Aluminium	0.19	5.25
Invar	0.25	5.09
Steel 304L	0.45	4.89
Titanium	7.78	3.68
Lead	13.10	3.66

- Fit to *ActiWiz* data is generally poorer when change in IR is too large over the range of interest, i.e. the empirical model has its limits

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<b>Measured ZS Tank 2 EYETS15-16</b>	<b>0.39</b>	<b>4.91</b>

- Measured fit constants consistent with material composition of the ZS:
  - SS 304L vacuum tank and, SS 304L or Invar anode support [8]

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- The problem is time-discretised taking  $n$  bins of length  $\Delta t = 30$  mins:
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$$\text{IR}(t_n) = G \sum_{i=1}^n \underbrace{N_{L,n+1-i} P_{ex,n+1-i}}_{L_{BLM,n+1-i} [\text{Gy}]} \exp \left( -k_1 \ln \left( (i-1) \Delta t \right)^{k_2} \right)$$

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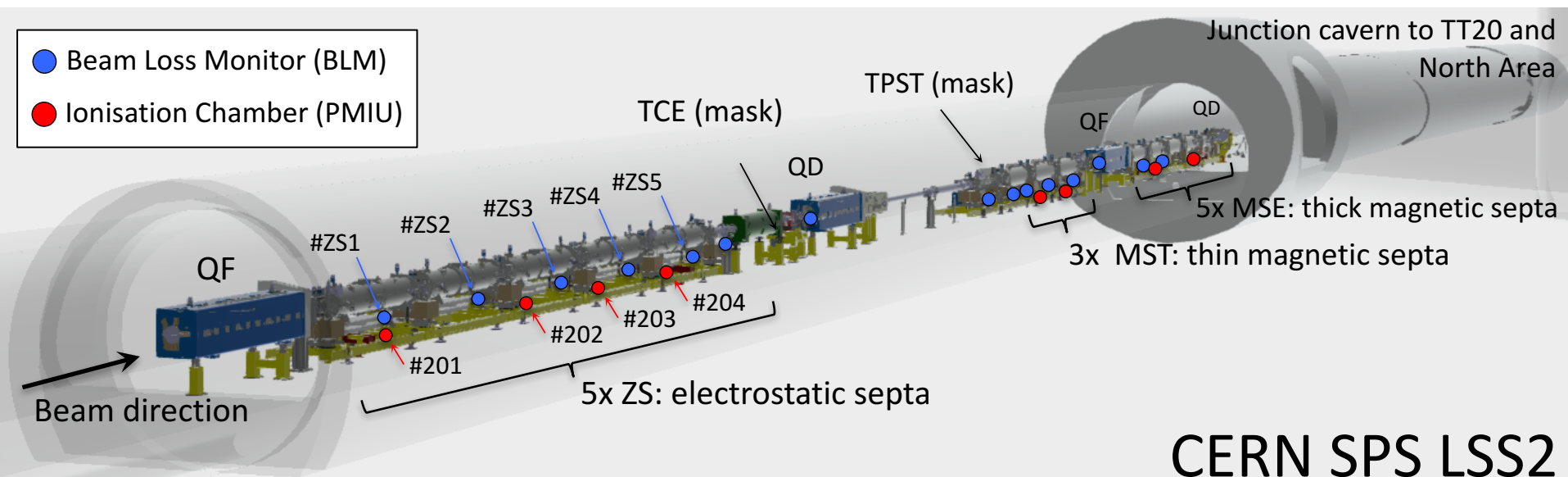
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- $P_{ex}$  is the number of extracted protons and  $N_L$  is the specific loss (per proton)
- To account for the changing extraction efficiency observed last year we linked the IR measured on a given PMIU detector to the prompt loss  $L_{BLM}$  measured at the nearest BLM: we fit to measured values of  $L_{BLM}$ :

$$L_{BLM} [\text{Gy}] = N_L \left[ \frac{\text{Gy}}{\text{p}^+} \right] P_{ex} [\text{p}^+]$$

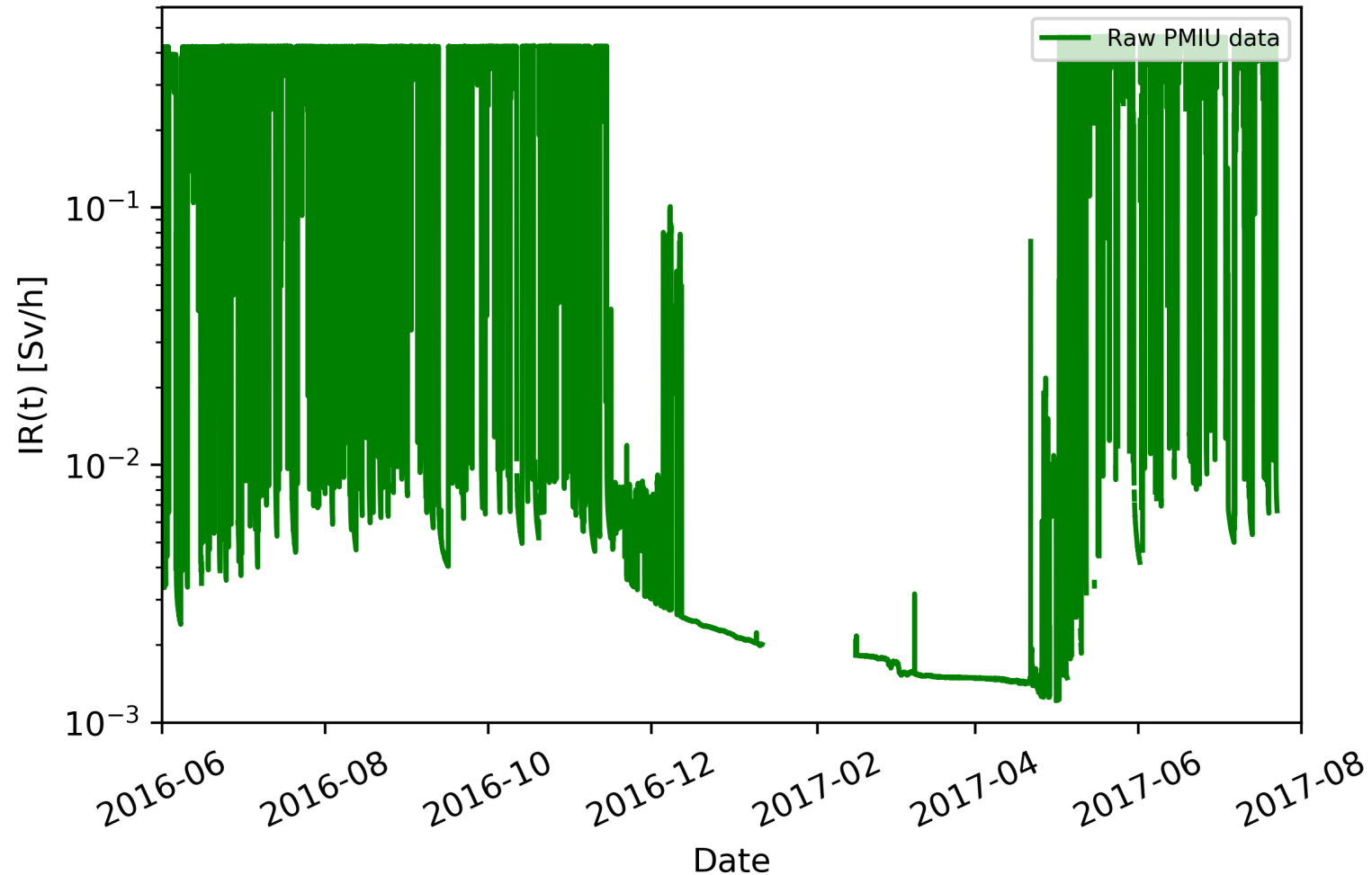
# A predictive model for IR(t) (2)

- We have a network of radiation monitors in LSS2:
  - BLMs for prompt extraction dose measurements:  $L_{BLM}(t)$
  - PMIUs for residual, induced radioactive (without beam):  $IR(t)$
  - We use the ring BCT to normalise to the number of protons extracted:  $P_{ex}(t)$
- All the data is logged by the CERN Accelerator Logging Service:
  - We use an API (via pyTimber) to access all time series data in CALS before preparing aligning, binning and filtering the data for fitting



CERN SPS LSS2

# Data preparation: an example\* PMIU202

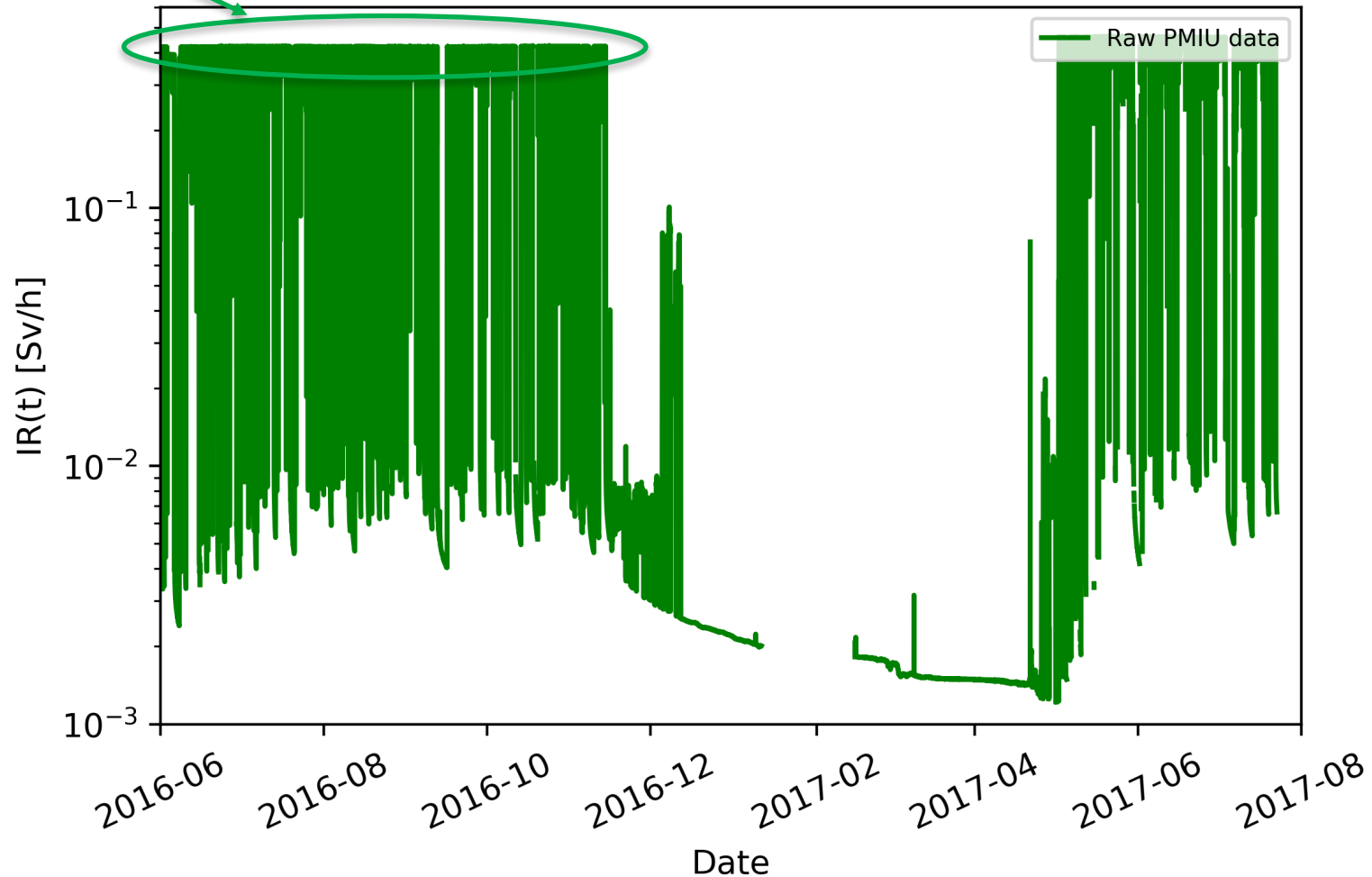


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# Data preparation: an example\* PMIU202

Beam ON = saturation from prompt losses

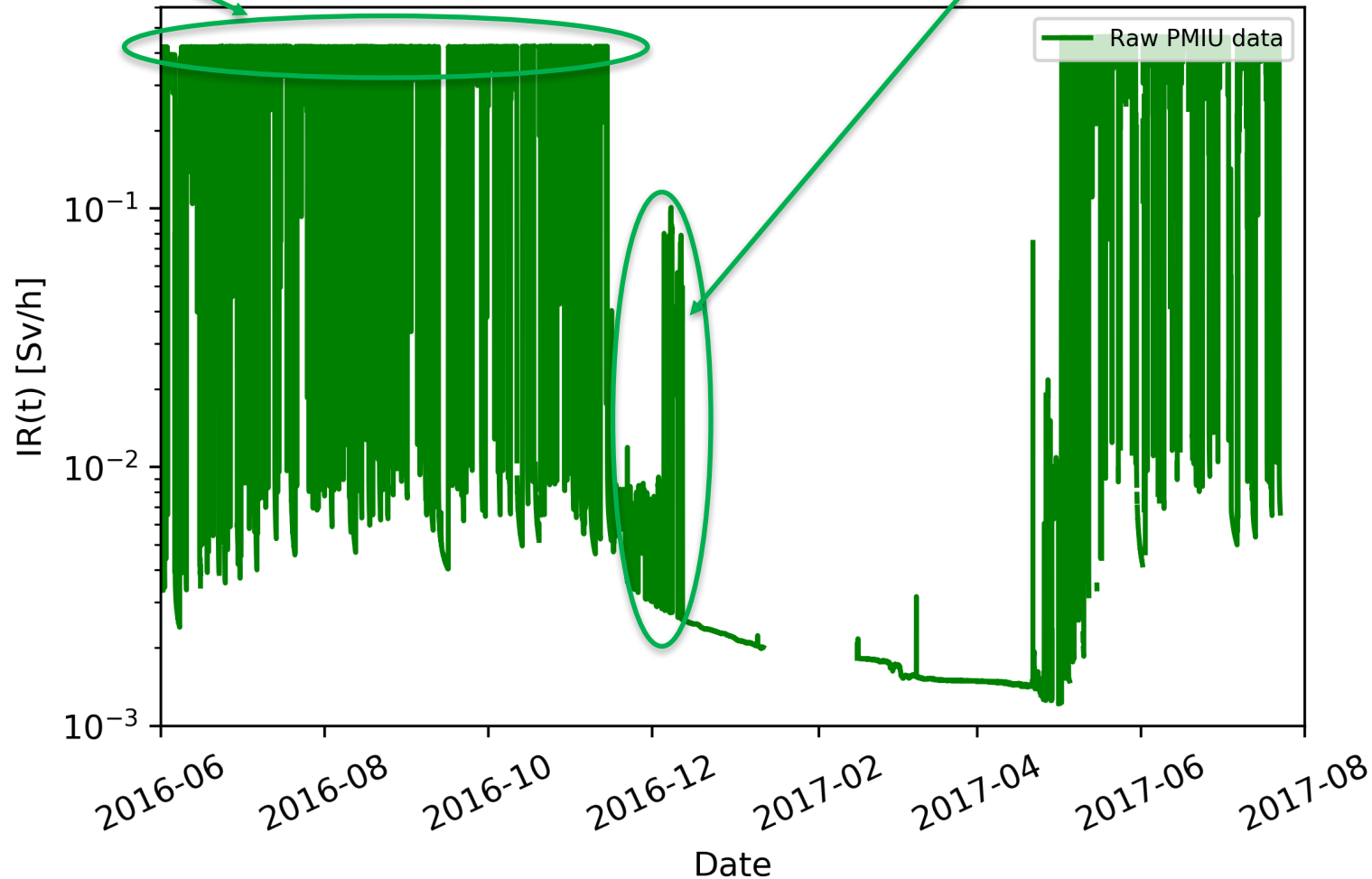


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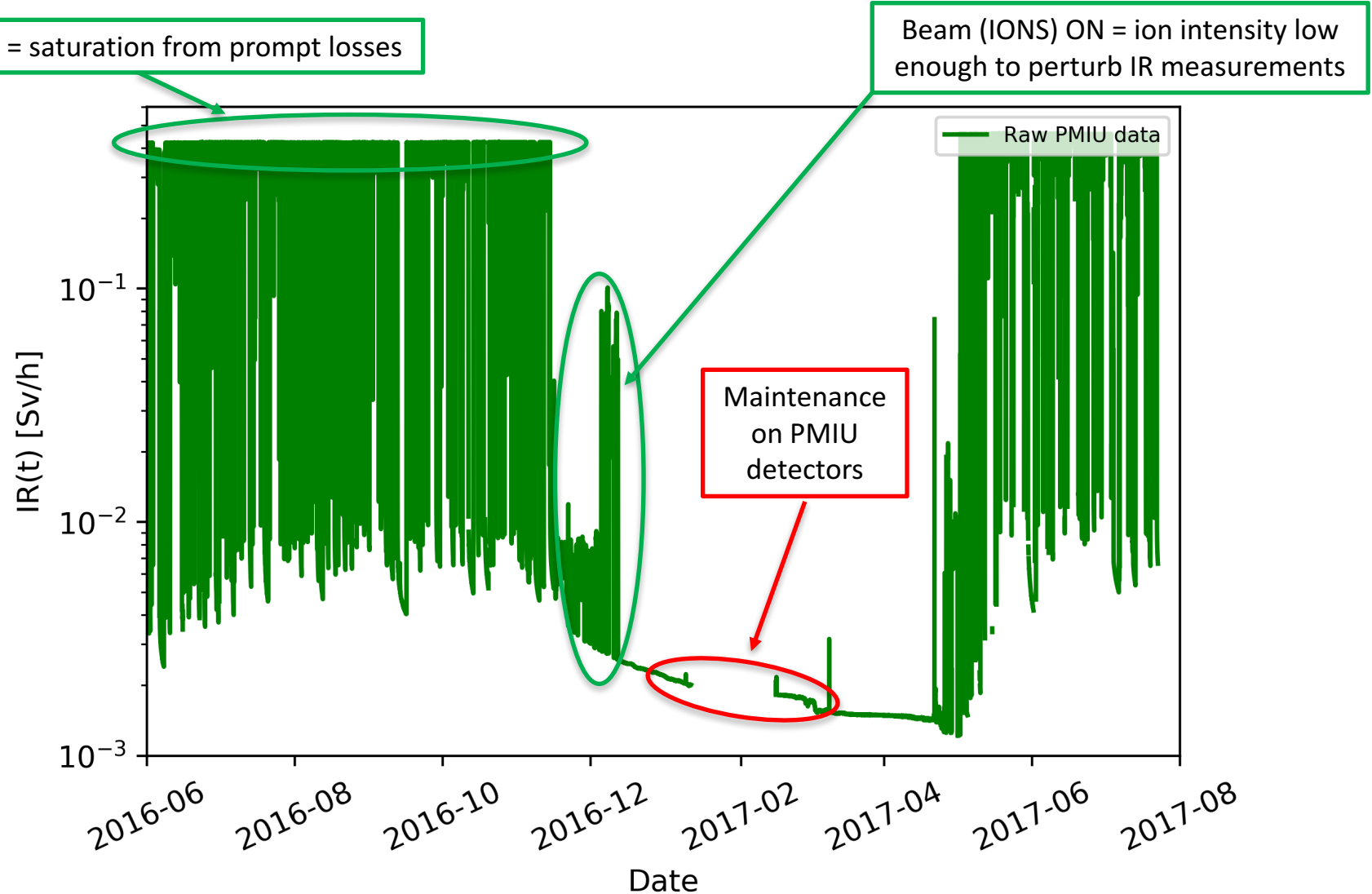
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Beam (IONS) ON = ion intensity low enough to perturb IR measurements



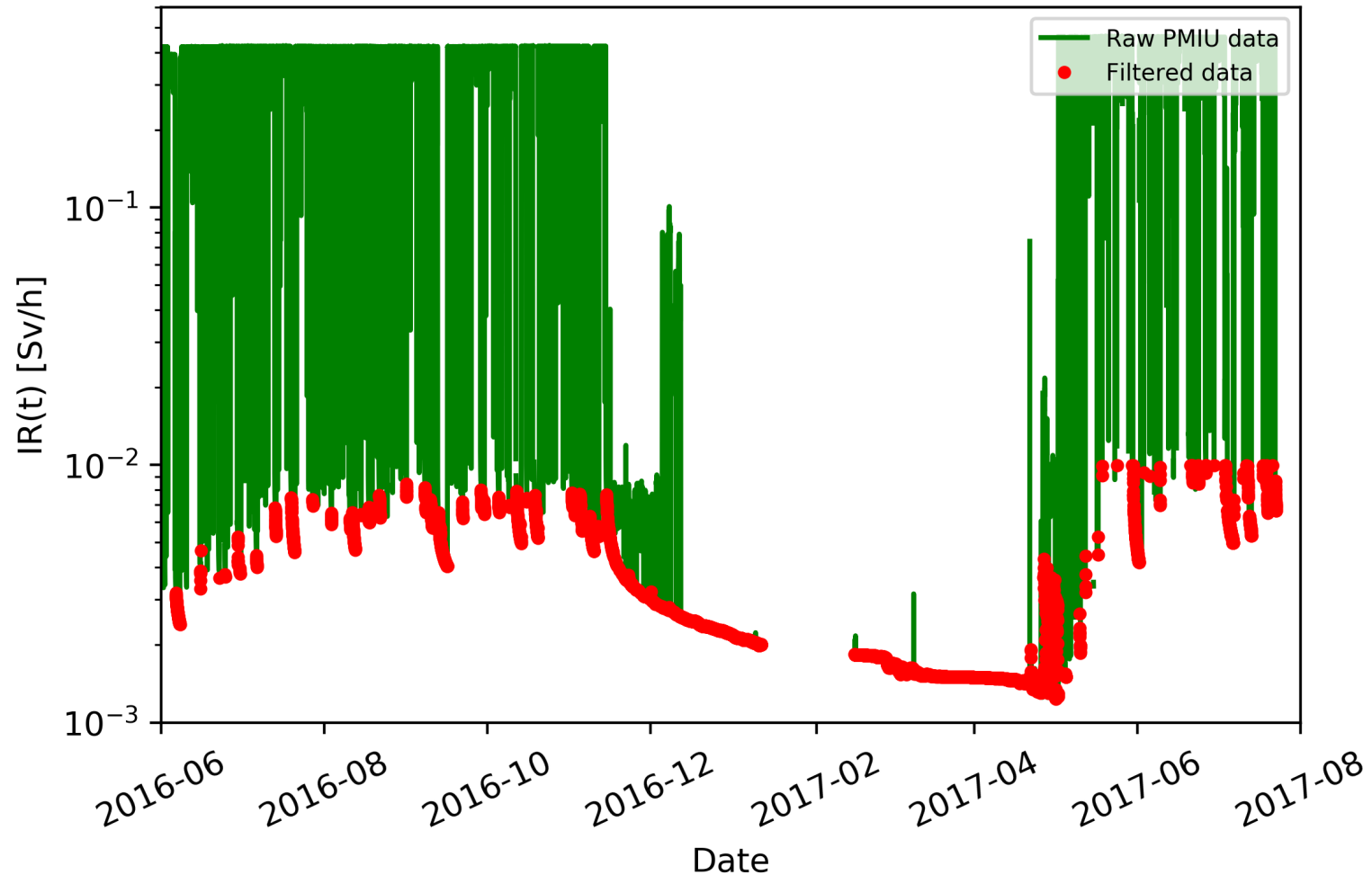
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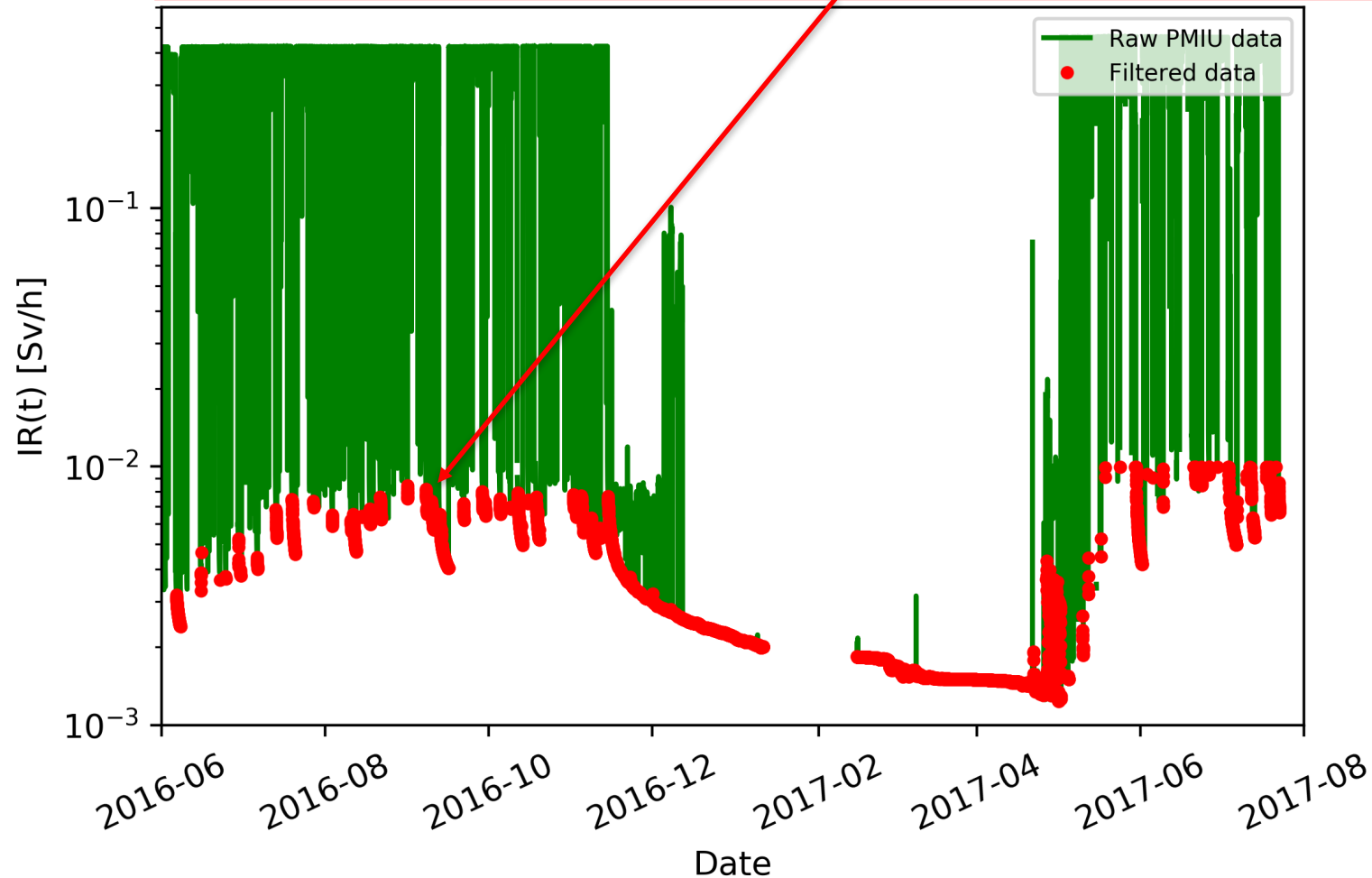
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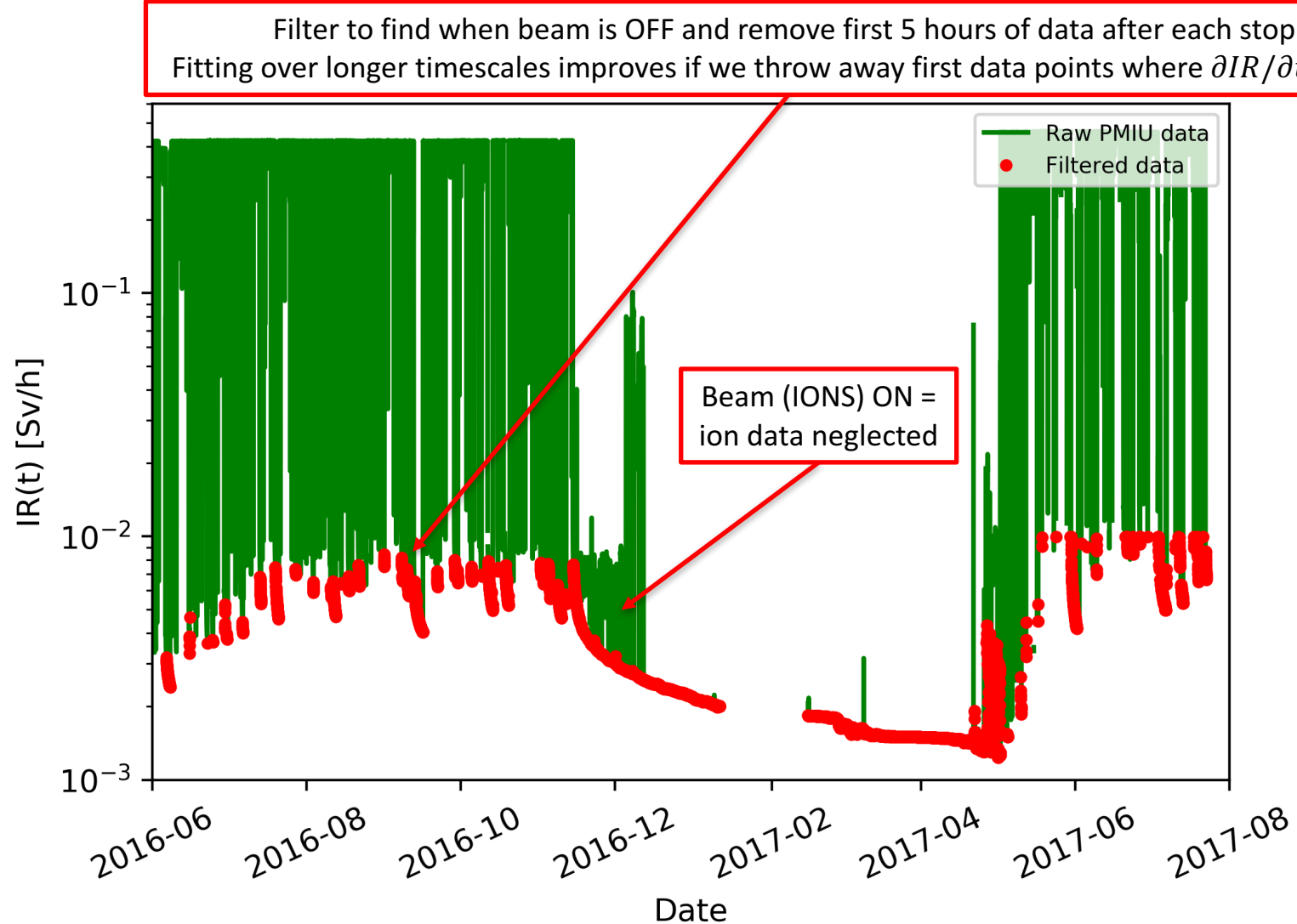
# Data preparation: an example\* PMIU202

Filter to find when beam is OFF and remove first 5 hours of data after each stop:  
Fitting over longer timescales improves if we throw away first data points where  $\partial IR/\partial t$  is large



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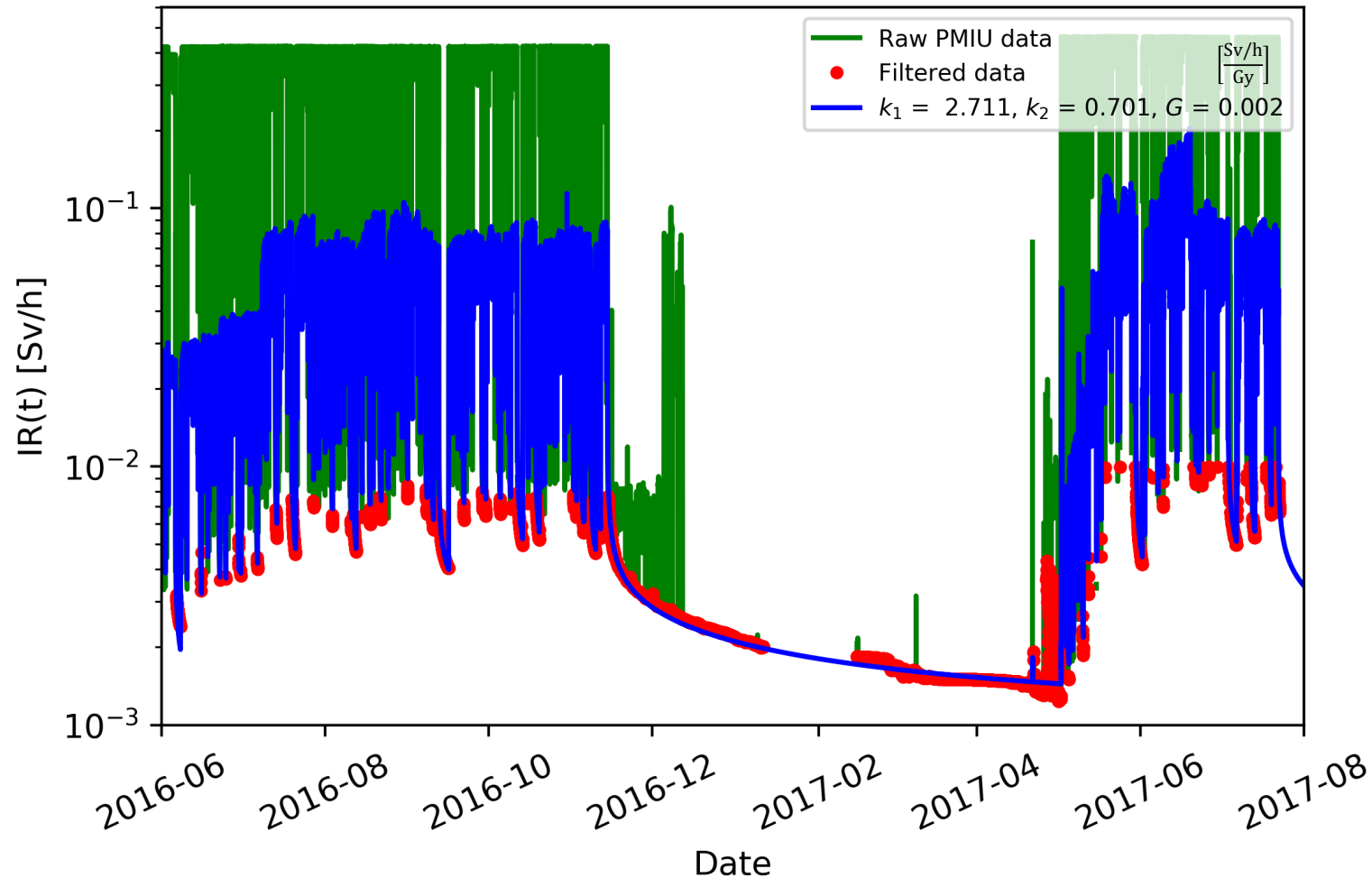
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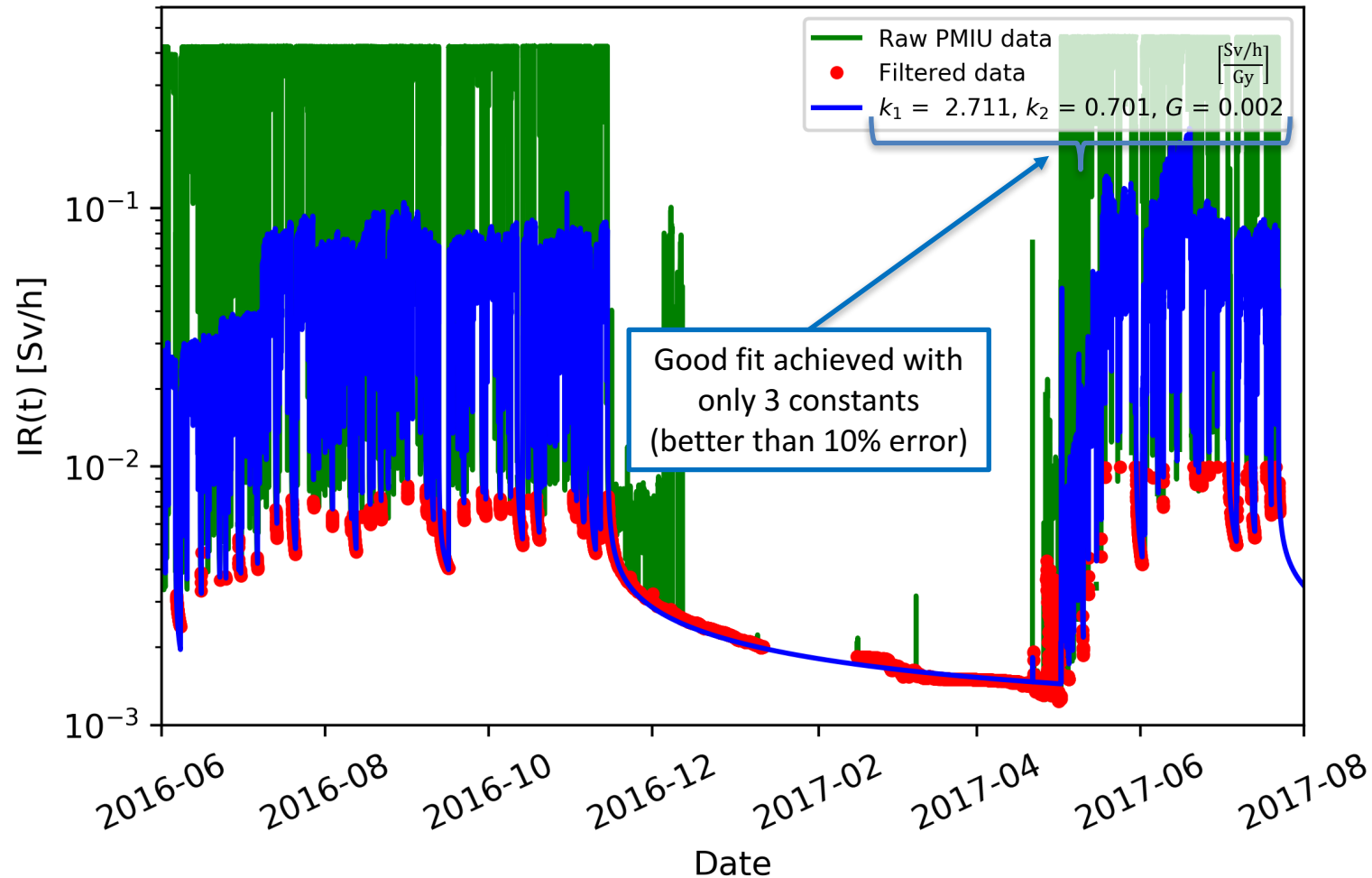
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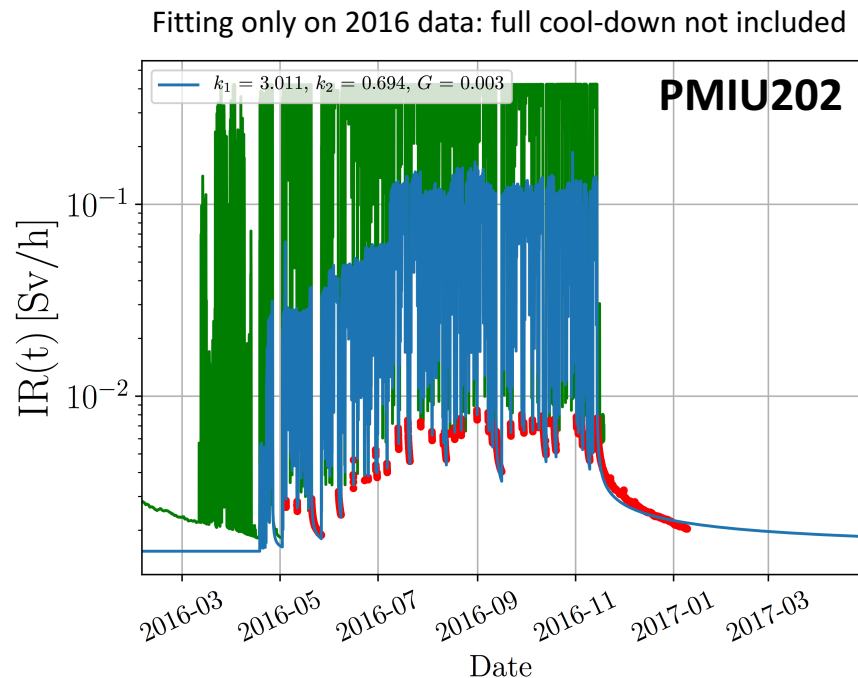
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# Fit constants for different detectors

PMIU #	Paired with BLM.ZS #	Fit constants $IR(t)^*$		
		$k_1$	$k_2$	$G \left[ \frac{Sv/h}{Gy} \right]$
1	201	3.14	0.69	0.0062
<b>2</b>	<b>202</b>	<b>3.01</b>	<b>0.69</b>	<b>0.0027</b>
3	203	2.73	0.66	0.0007
4	204	2.44	0.76	0.0008
average	average	3.31	0.67	0.0016

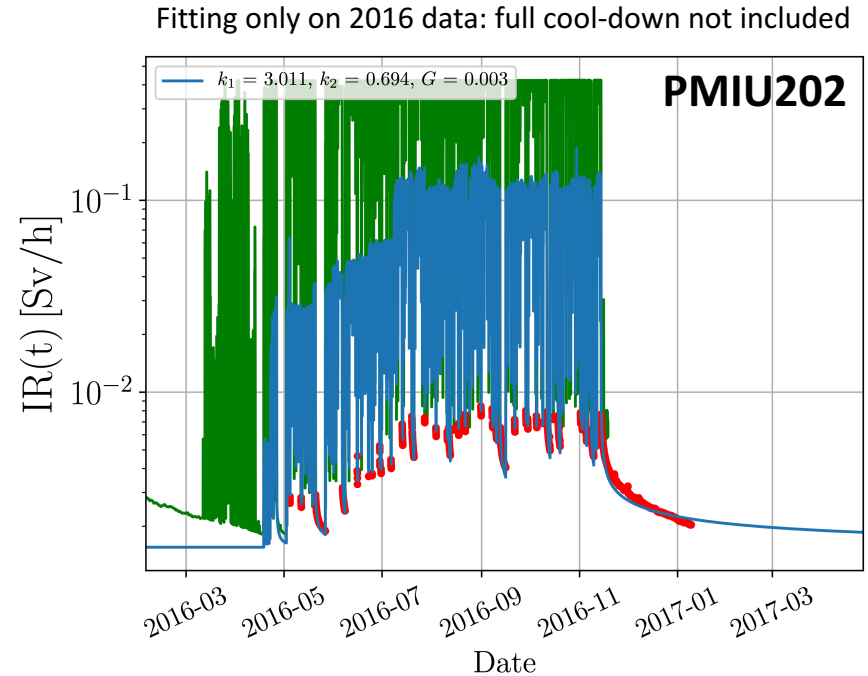
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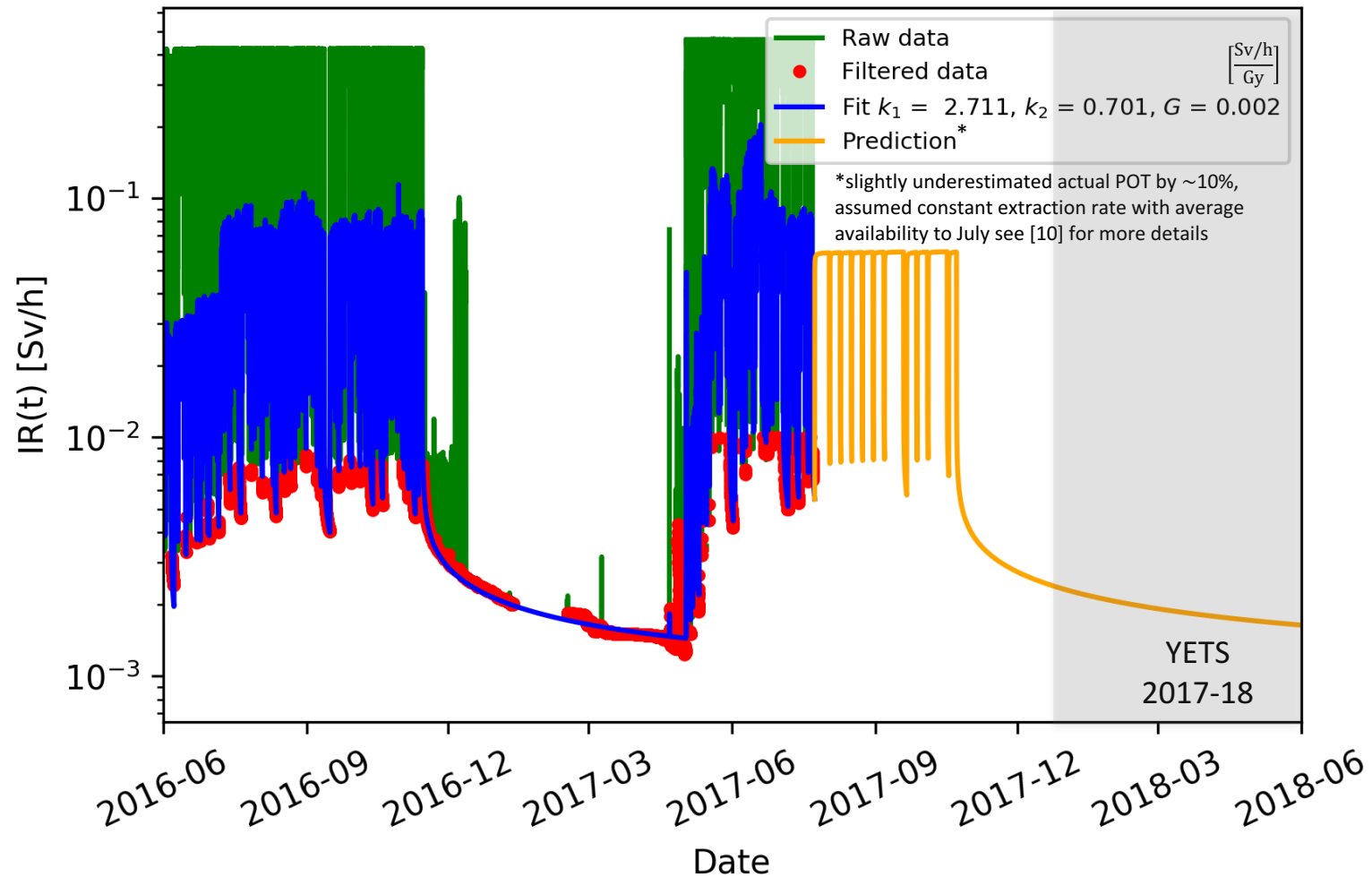
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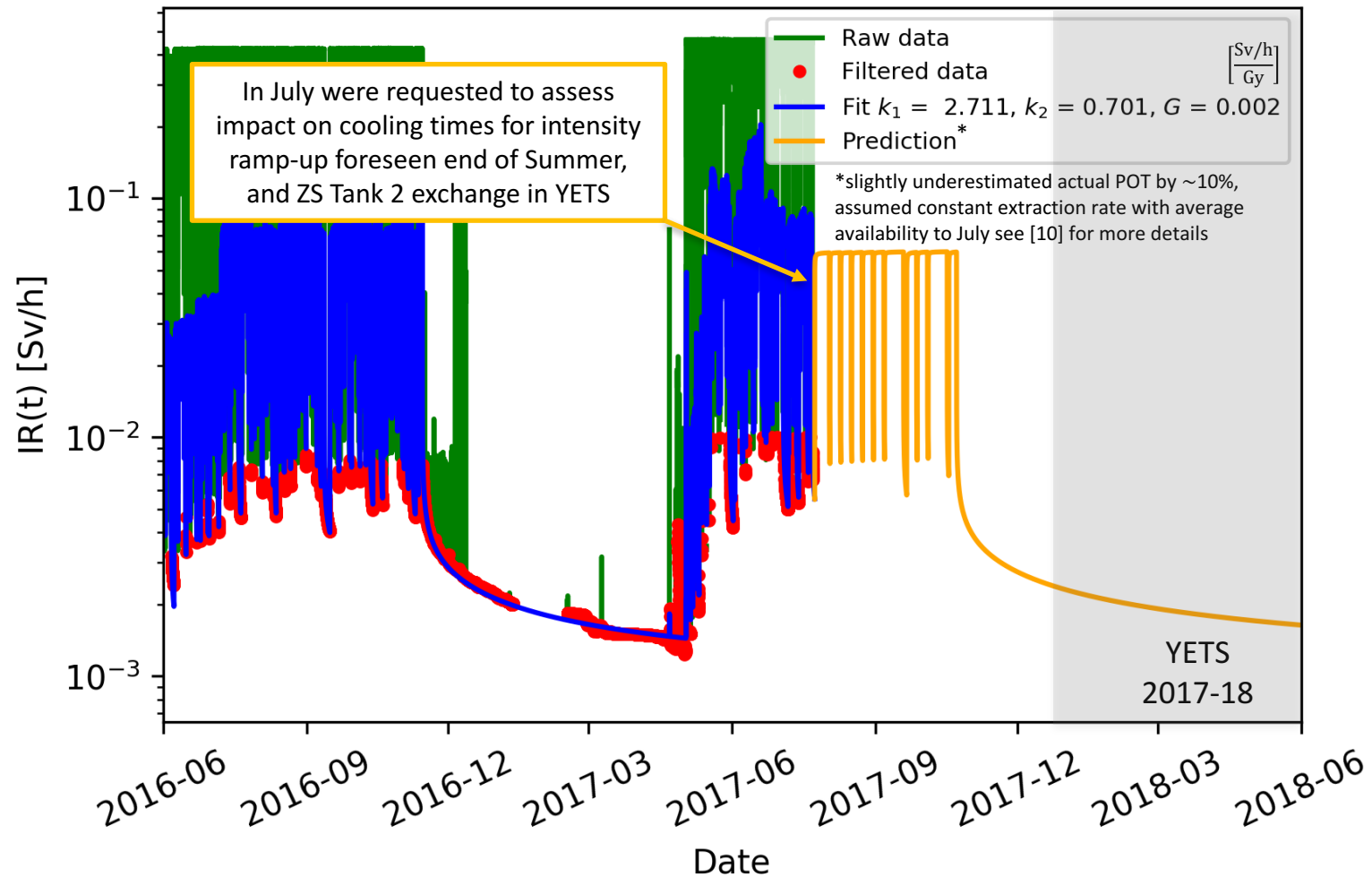


- Variation in  $G$  is due to different relative positions of BLMs and PMIUs
- Difference in  $k_1$  and  $k_2$  indicate spatial difference in the decay rate:
  - Local differences of materials composing equipment or being activated?
- Predictive power of model was also tested using fit constants with older data, dating back to 2011: agreement to about 10% was found [9]

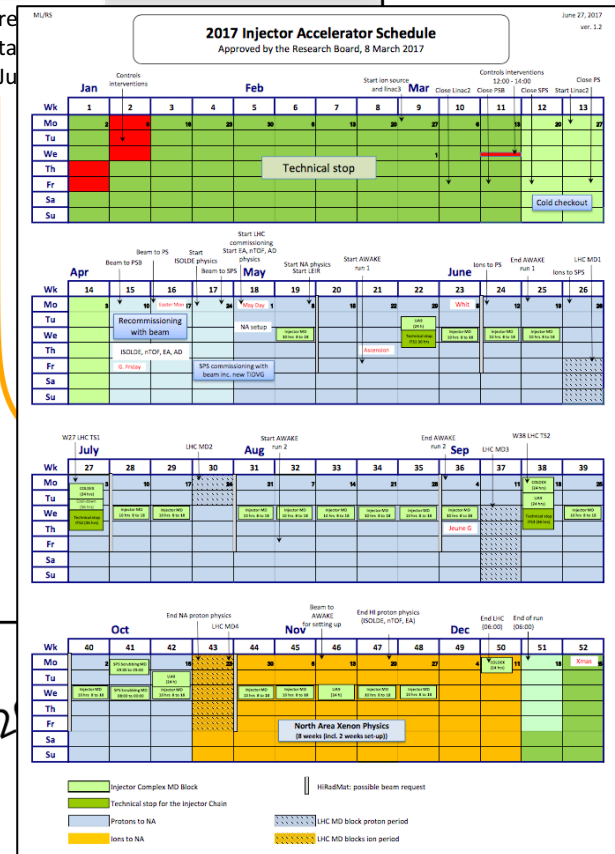
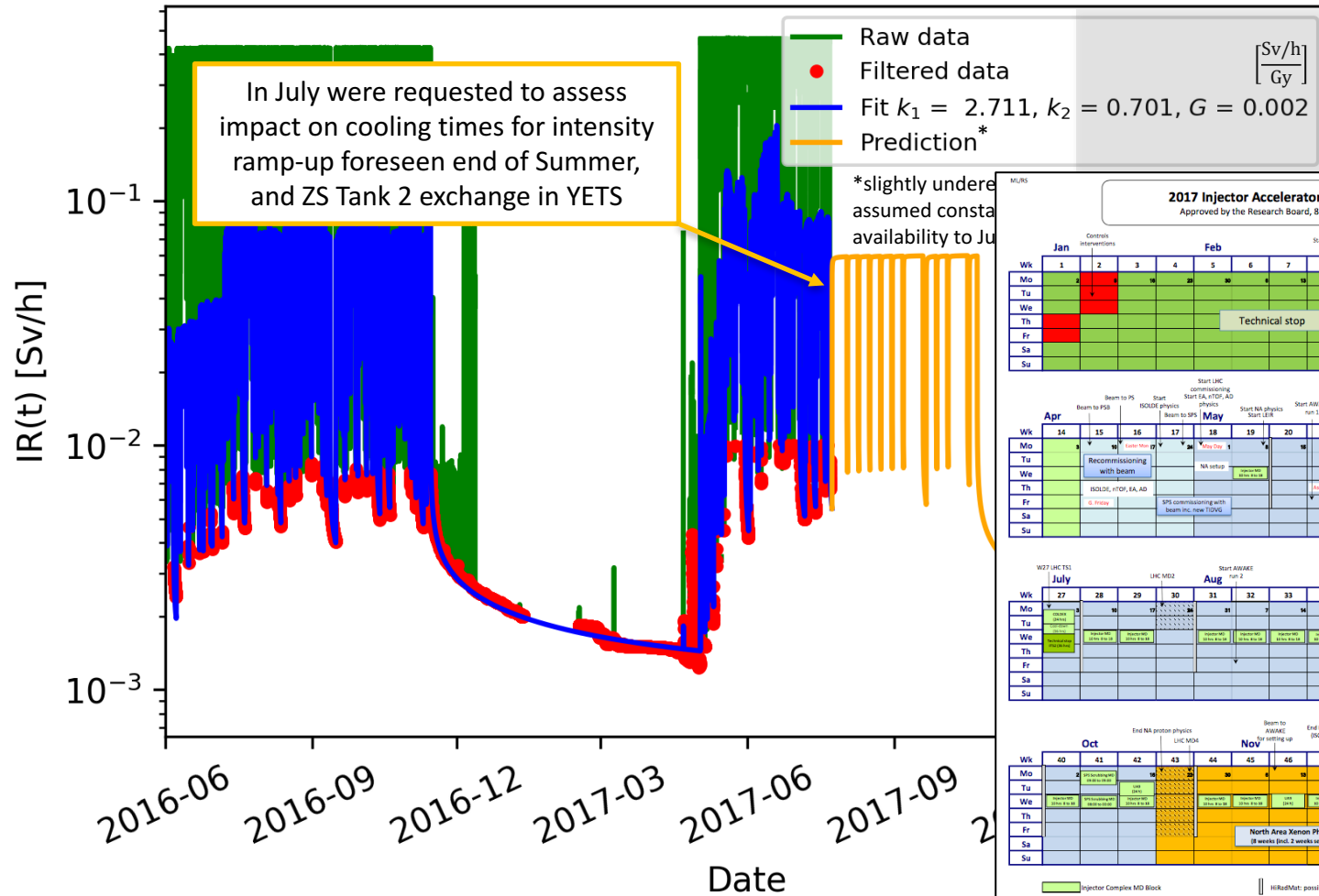
# Predictions for IR in YETS 2017



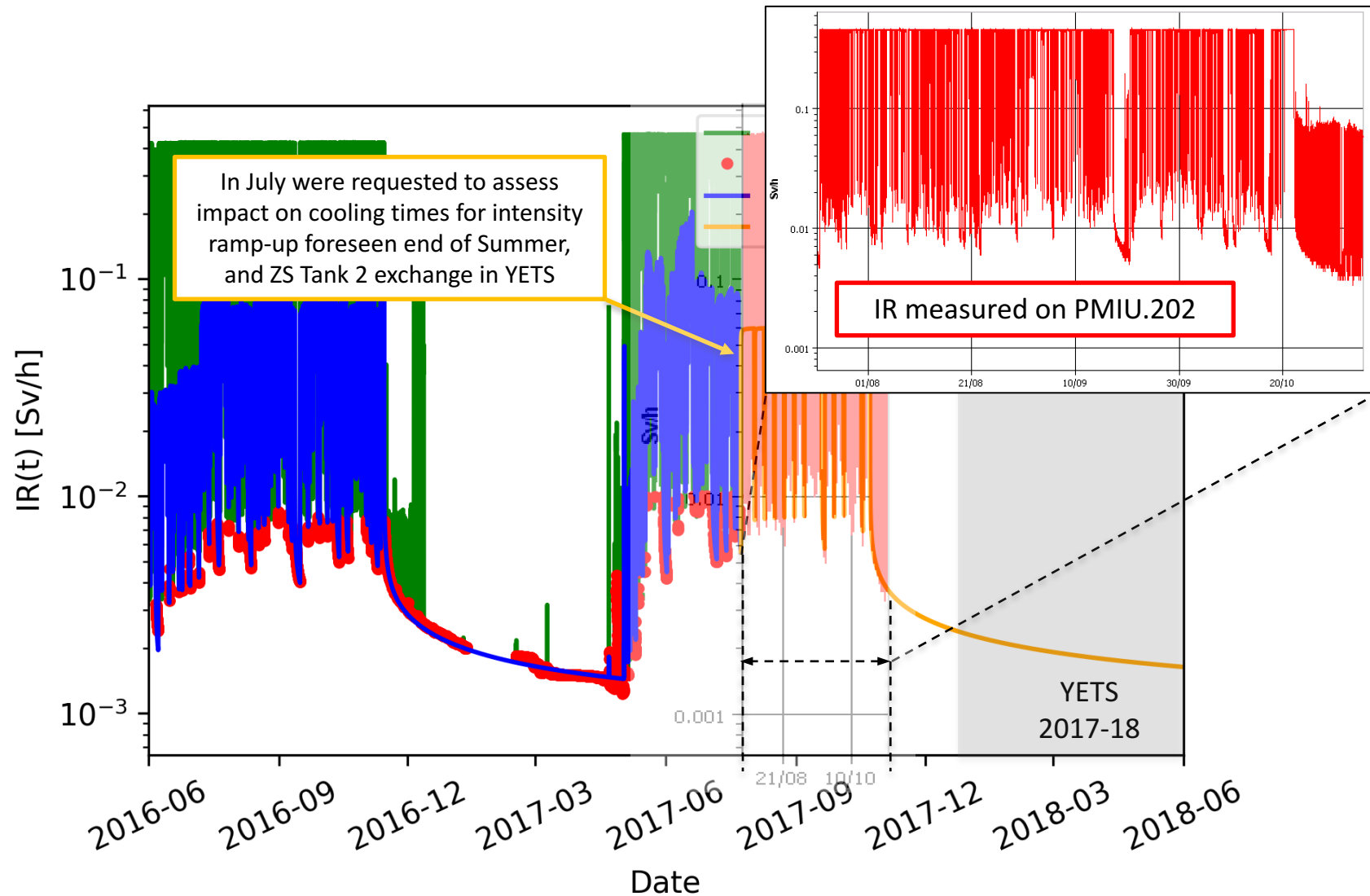
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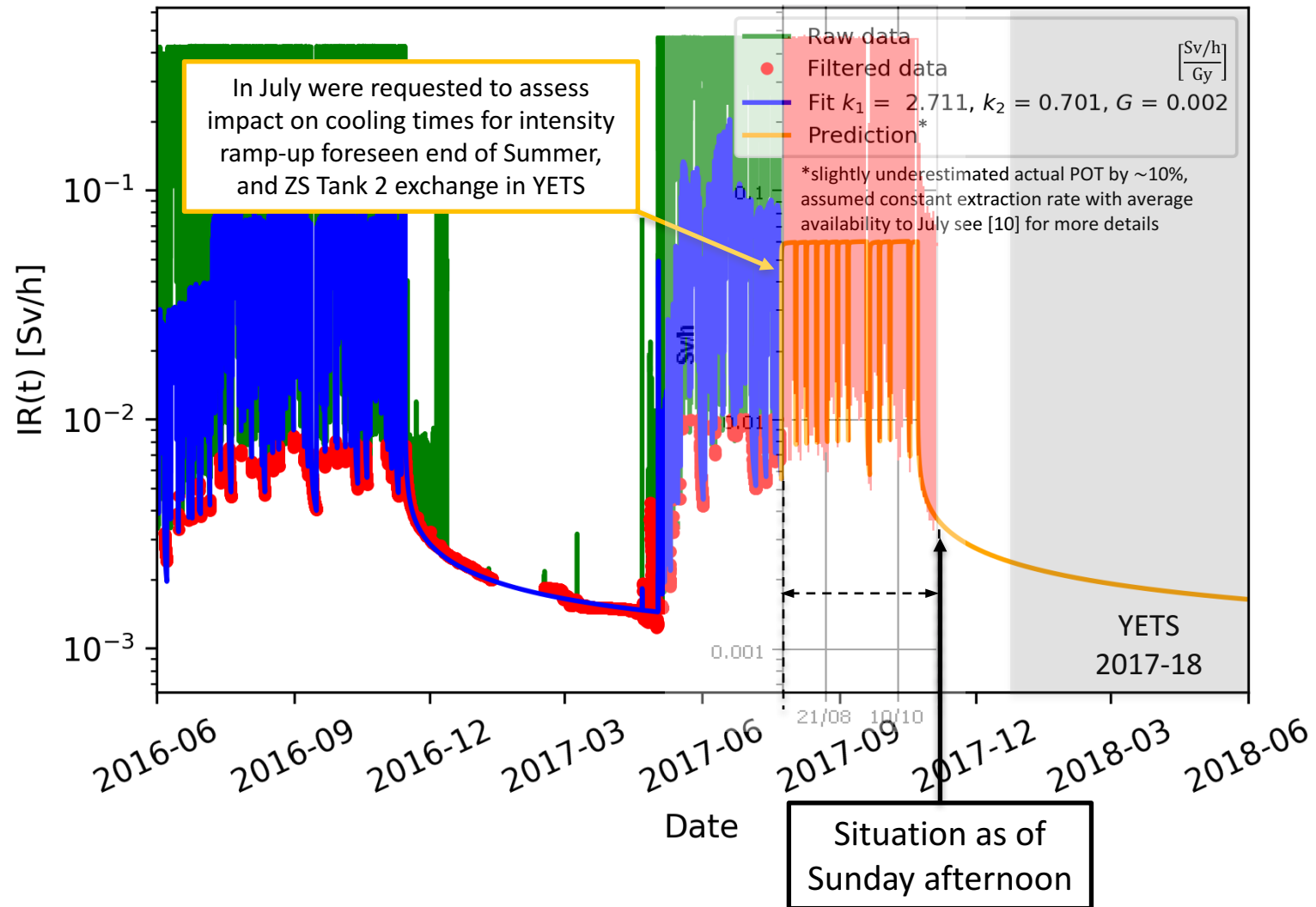
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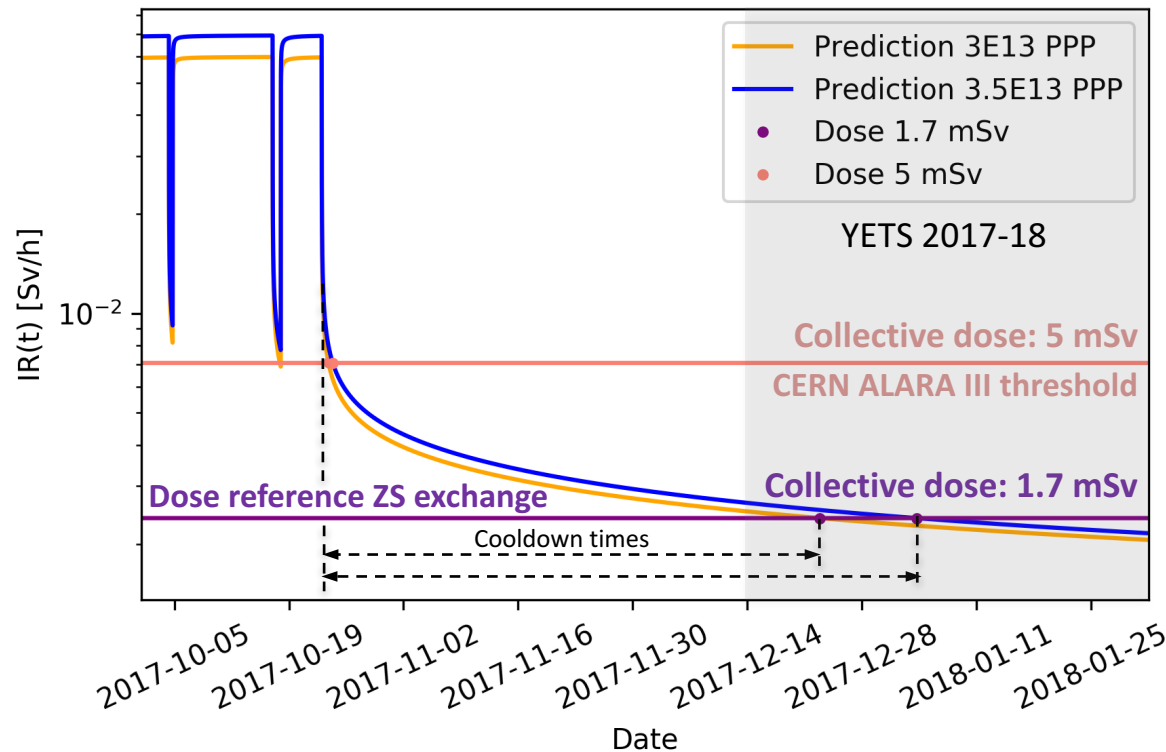
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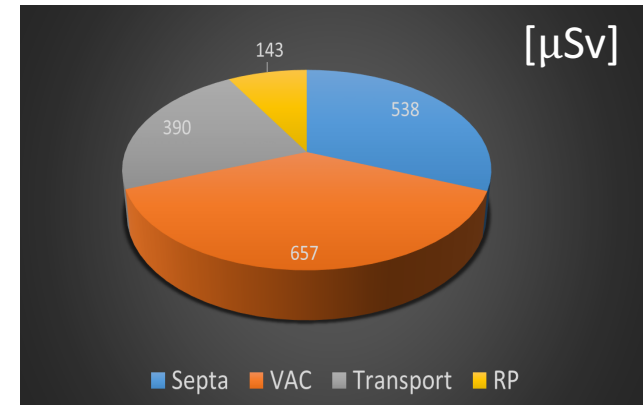


# Cooldown predictions in YETS 2017



IR dose rate measured on PMIU.202 during exchange of a ZS tank (ZS.21639) combined with collective dose taken in February 2016 (100 days after stop) used as a reference:

Work dose estimation = 3.8 mSv  
 Collective dose taken = 1.7 mSv  
**PMIU202 measured = 2.3 mSv/h**

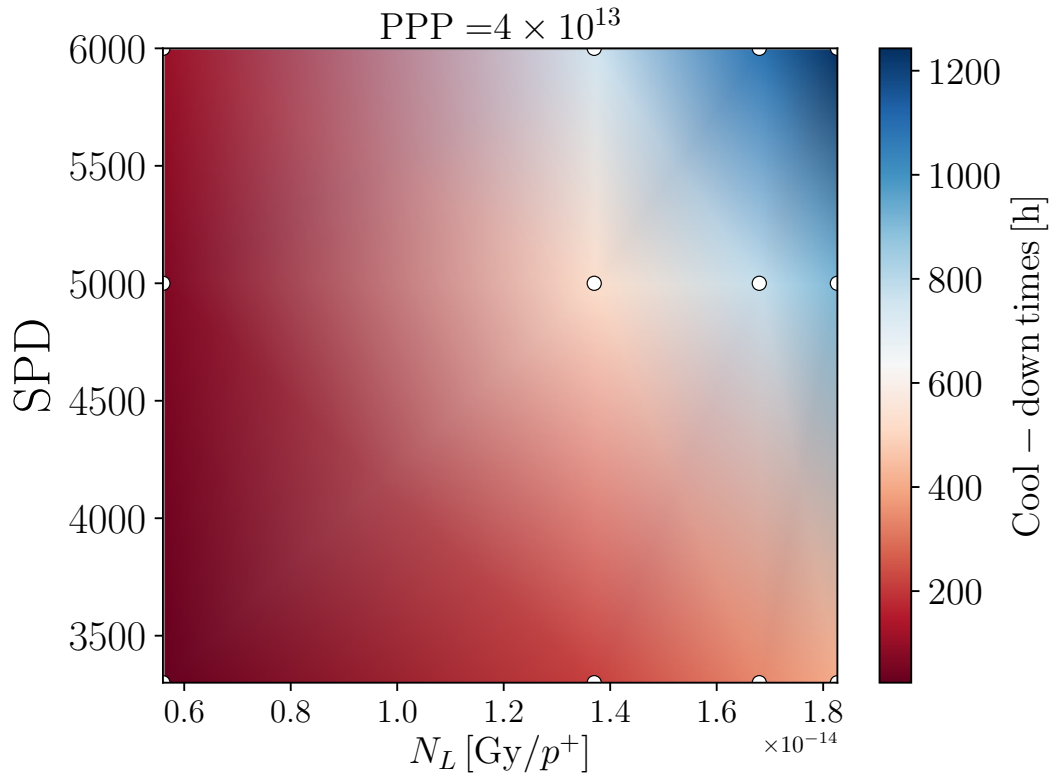


(E)YETS	Protons per pulse (av.) [ppp]	Spills per day (av.) [spd]	$N_L$ (av.) [Gy/ $10^{14}$ p <sup>+</sup> ]	Protons extracted [ $\times 10^{19}$ ]	Cooldown time for given dose	
					5 mSv	1.7 mSv
2017 – 18 (predicted)	3.0E13	2700	< 1.5	1.2	20 h	60 d 21 h
	3.5E13	2700		1.3	1 d 8 h	72 d 16 h
2015 – 16 (measured)	2.7E13	3300	1.8	1.6	5 d 19 h	100 d (reference intervention, above)



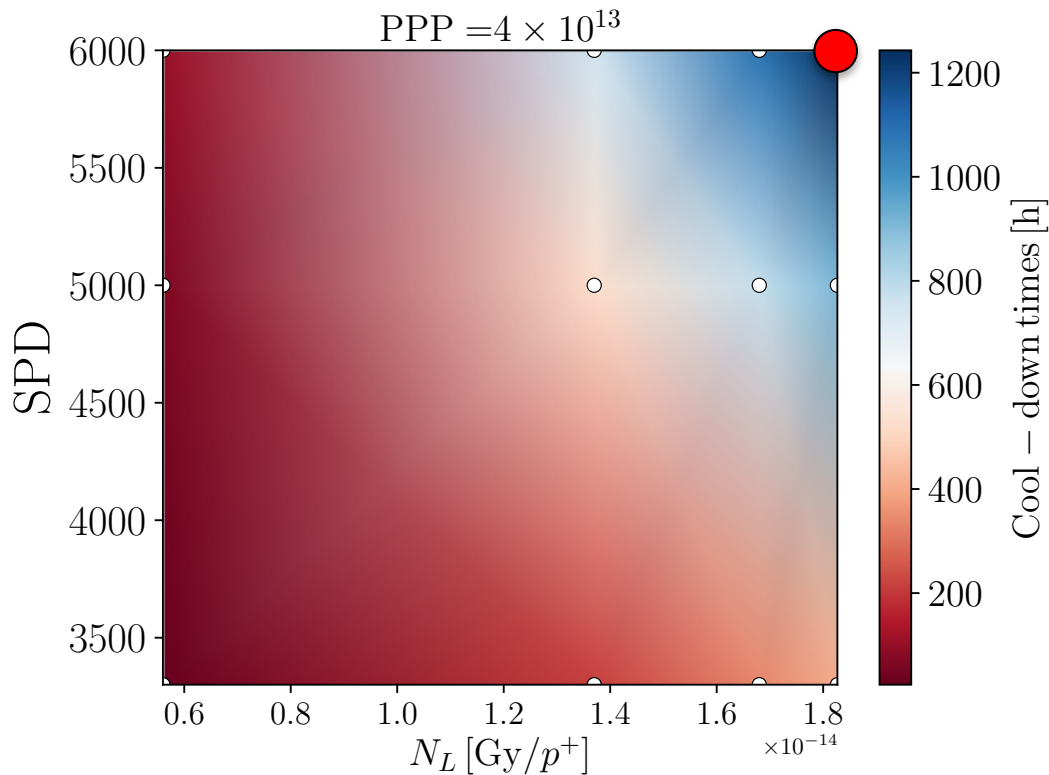
# Cool-down predictions for SPS BDF


- Predicted cool-down time at the end of a year of SPS BDF operation [9]:
  - For a 5 mSv intervention based on the exchange of ZS Tank 2 carried out February 2016
  - PMIU202 is paired with the specific loss measured on the BLM next to ZS Tank 2
  - This neglects the build-up of longer living isotopes over many years of operation!



# Cool-down predictions for SPS BDF

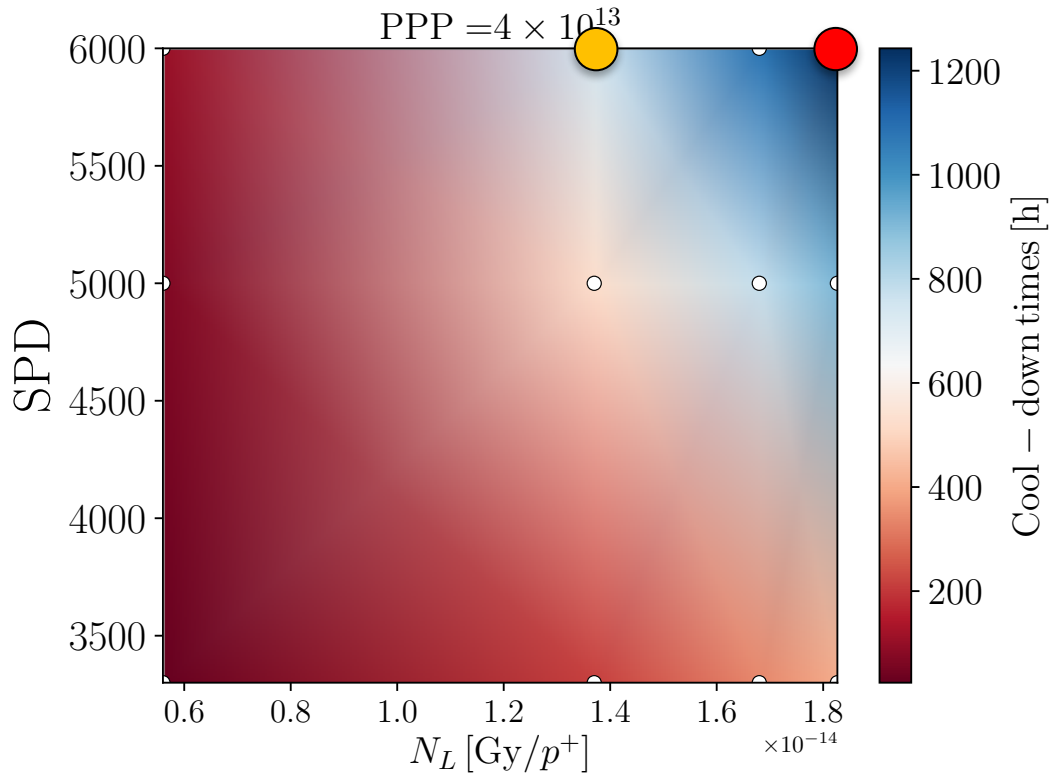
- Predicted cool-down time at the end of a year of SPS BDF operation [9]:
  - For a 5 mSv intervention based on the exchange of ZS Tank 2 carried out February 2016
  - PMIU202 is paired with the specific loss measured on the BLM next to ZS Tank 2
  - This neglects the build-up of longer living isotopes over many years of operation!





Key	Extraction efficiency (specific loss) $N_L$ [Gy/ $10^{14} p^+$ ]	Cool-down time
	As in 2015: 1.8	~7 weeks

# Cool-down predictions for SPS BDF

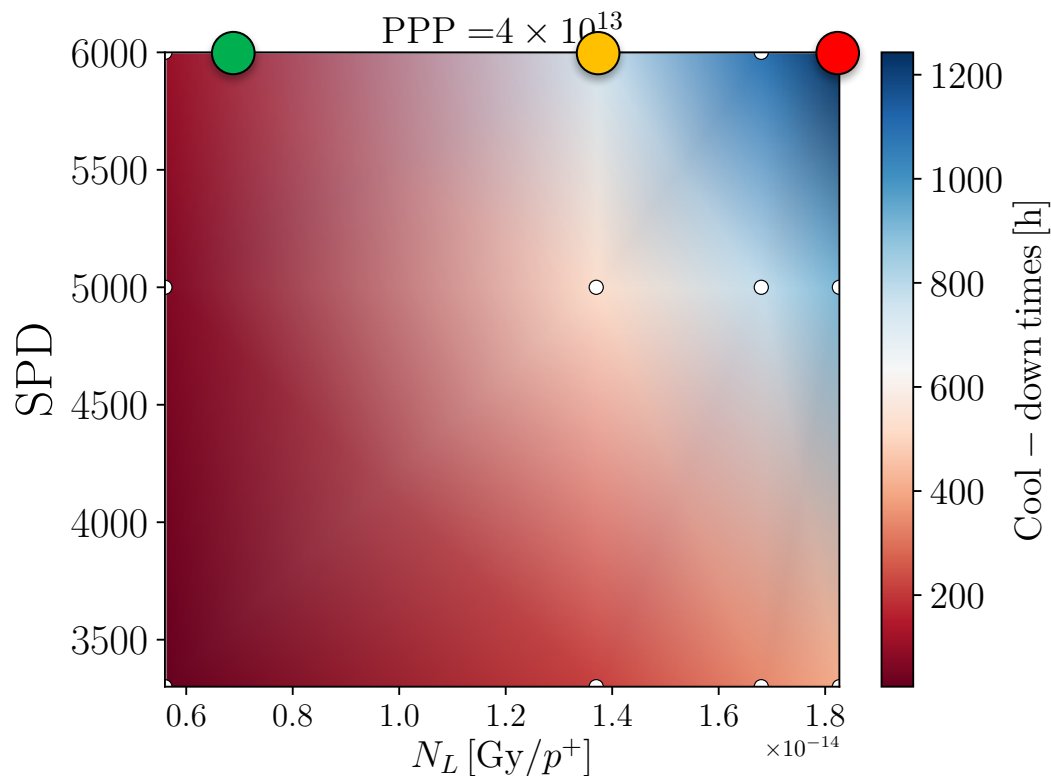
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	As in 2017: 1.4	~4 weeks

# Cool-down predictions for SPS BDF

- Predicted cool-down time at the end of a year of SPS BDF operation [9]:
  - For a 5 mSv intervention based on the exchange of ZS Tank 2 carried out February 2016
  - PMIU202 is paired with the specific loss measured on the BLM next to ZS Tank 2
  - This neglects the build-up of longer living isotopes over many years of operation!



Key	Extraction efficiency (specific loss) $N_L$ [Gy/ $10^{14} p^+$ ]	Cool-down time
● (Red)	As in 2015: 1.8	~7 weeks
● (Yellow)	As in 2017: 1.4	~4 weeks
● (Green)	< 2017/2: 0.7	< 1 week

We found the cooldown times to be roughly quadratic with intensity or specific loss:

$$t_{\text{cooldown, 5 mSv}} \sim N_L^2, SPD^2, PPP^2$$

# Conclusions

- An empirical relationship has been exploited to model the build-up and cool-down of slow extraction induced radioactivity of the SPS
- A software toolkit was developed in python to exploit the empirical relationship and to fit logged data:
  - Only 3 fit constants are required to give reasonable predictions over the timescales fitted (~10%)
  - Cool-down times for a simple case study (ZS exchange) were predicted for YETS 2017 and for future SPS BDF operation
  - As longer time periods are fitted, one expects the prediction of the build-up of longer-lived isotopes will improve: to be studied
- The cool-down following a year of operation scale quadratically with intensity or extraction inefficiency:
$$t_{\text{cooldown}} \sim N_L^2, \text{SPD}^2, \text{PPP}^2$$
- The model offers the possibility to predict cool-down times online and alert the operations team to anomalies: development possible

# Thank you!

- Any questions?

# References

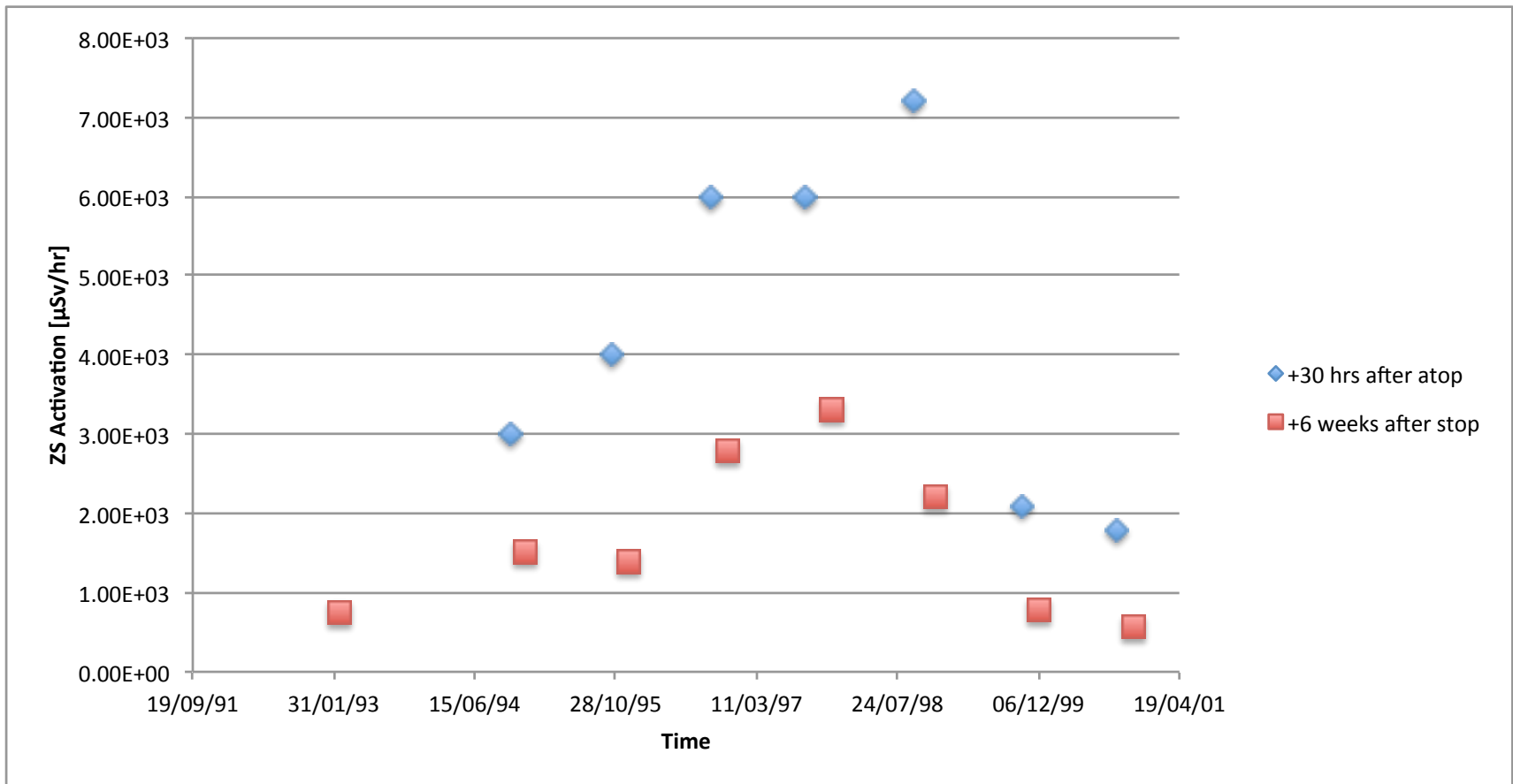
- [1] R.L. Keizer et al., *High intensity running and radiation problems during the 1995 physics run*, SL-Note-96-05-MS, Geneva, CERN
- [2] R.L. Keizer et al., *High intensity running and radiation problems*, SL-Note-95-14-BT, Geneva, CERN, 1995
- [3] G. Ferioli and R.L. Keizer, *Analysis of the induced radioactivity in the SPS extraction channels during 1994*, SL-Note-95-06-BT, 1995
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- [6] G.R. Stevenson, *Activation at accelerators*, CERN-TIS-RP-90-10-CF, Jahrestagung 1990, Fachverband für Strahlenschutz, Göttingen, Germany, pp. 109-117, 1990.
- [7] H. Vincke and C. Theis, *ActiWiz – optimizing your nuclide inventory at proton accelerators with a computer code*, Progress in Nuclear Science and Technology, vol 4., pp. 228-232, 2014
- [8] J. Borburgh, *Activation reduction due to different septum tank materials*, SLAWG meeting #7, 28 September 2016, Geneva, CERN: <https://indico.cern.ch/event/570165/>
- [9] M.A. Fraser et al., *Modelling the radioactivity induced by slow extraction losses in the CERN SPS*, Proc. IPAC'17, Copenhagen, Denmark, paper TUPIK086, 2017
- [10] A.C. Araujo Martinez and M. A. Fraser, *Modelling slow extraction induced radioactivity in SPS LSS2*, CERN-STUDENTS-Note-2017-035, Geneva, CERN, 2017.

# Extra slides



# Build-up of IR in LSS6 during WANF

- Surveys made before, during and after the WANF period in LSS6 show the heating and cooling of LSS6. Measurements made 30 hours and again a few weeks later after operation was halted. Background recovers a year or two after WANF ceased.



# Deriving the effective half-life of IR(t)

- Differentiating (using the chain rule) the empirical ansatz:

$$IR(t) \propto \exp(-k_1 \ln(t)^{k_2}),$$

one can write:

$$\frac{dIR(t)}{dt} = -\frac{k_1 k_2 \ln(t)^{k_2-1}}{t} IR(t)$$

- This a linear ODE which resembles our exponential law for radioactive decay with a time-dependent effective decay constant:

$$\frac{dIR(t)}{dt} = -\lambda(t)IR(t)$$

where by inspection the effective half-life can be written:

$$t_{1/2}(t) = \frac{\ln(2)}{\lambda(t)} = \frac{t}{k_1 k_2 \ln(t)^{k_2-1}} \ln(2)$$

# ALARA III at CERN

<https://edms.cern.ch/document/1296520/1>

## Waiving of the ALARA committee meeting

### Circumstances

- Repetitive intervention
  - A **procedure** has been worked out under which circumstances a waiving of the ALARA committee meeting could be possible
  - Generic DIMRs should be worked out and approved a priori in an ALARA committee meeting.
- Urgent maintenance/repair
  - 'Urgent ALARA committee' decision
  - No 'formal (physical)' ALARA committee meeting required
  - Generic DIMRs for standard maintenance/repair should be worked out and approved a priori in an ALARA committee meeting.

new

### CRITÈRE DE DOSE INDIVIDUELLE

**hard limits**

Équivalent de dose prévisionnel individuel ( $H_i$ ) pour l'intervention, ou pour l'ensemble des interventions de même nature lorsque celles-ci sont répétées plusieurs fois sur une année :

100  $\mu$ Sv

1 mSv

niveau I

niveau II

niveau III

### CRITÈRE DE DOSE COLLECTIVE

Équivalent de dose prévisionnel collective ( $H_c$ ) pour l'intervention, ou pour l'ensemble des interventions de même nature lorsque celles-ci sont répétées plusieurs fois sur une année :

500  $\mu$ Sv

**5 mSv**

~~10 mSv~~

niveau I

niveau II

niveau III

# History of ZS normalised losses: 2017

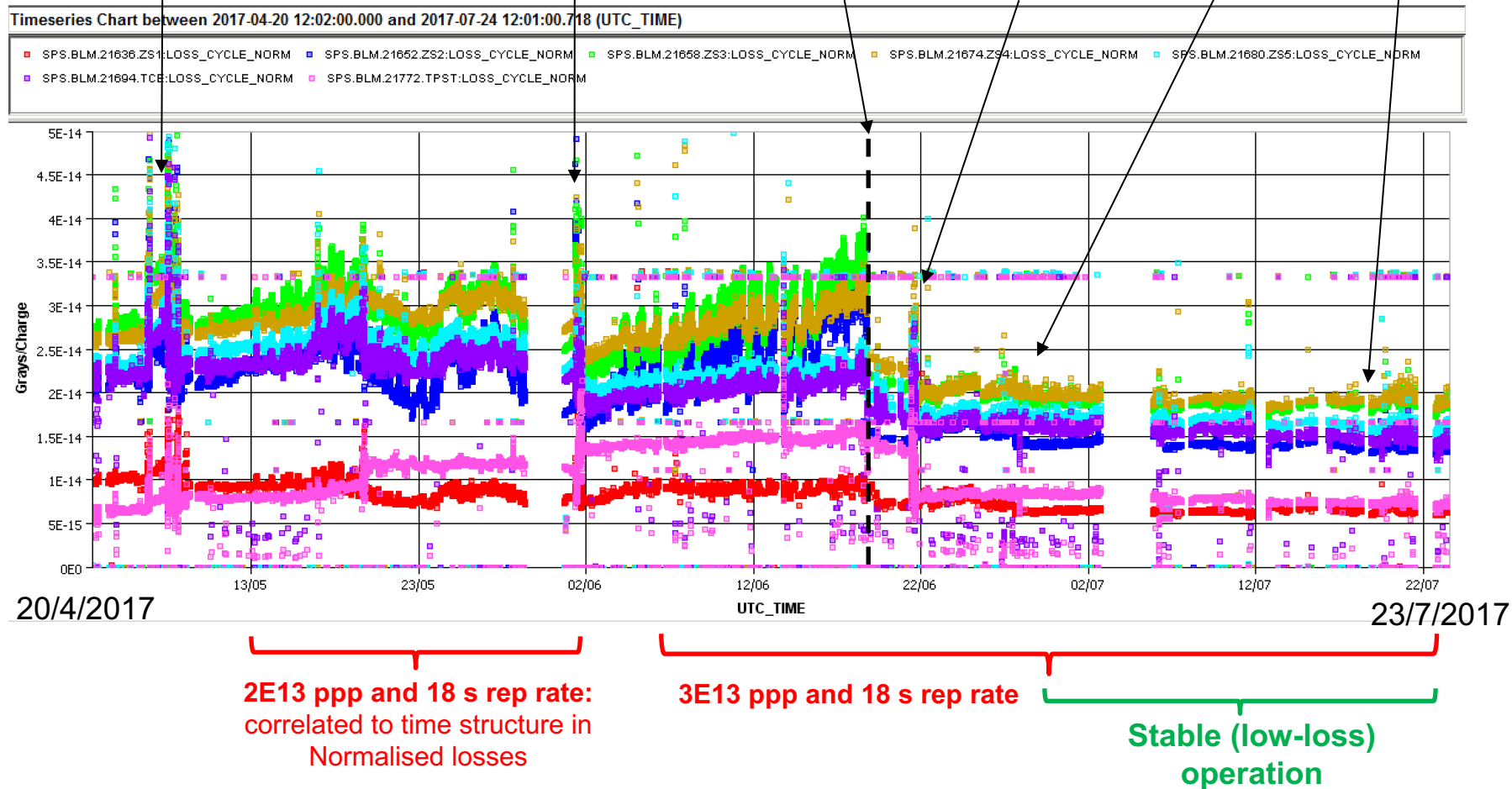
ZS re-alignment by  
TE-ABT: 9/5/2017

ZS re-alignment by  
TE-ABT 2/6/2017

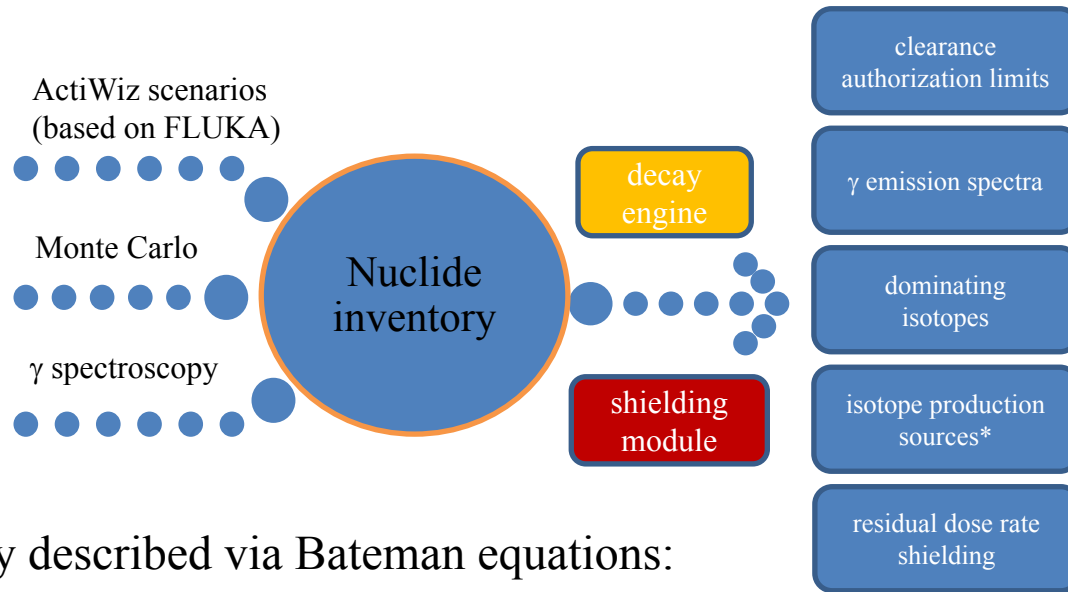
ZS Tank 2 cathode retracted by 2 mm  
All cathodes retracted a further 1.4 mm  
V = -220 kV -> -230 kV

ZS re-alignment by  
TE-ABT 22/6/2017

Gaps profiled:  
V = -220 kV



# ActiWiz: <http://actiwiz.web.cern.ch>



Production & decay described via Bateman equations:

$$\frac{dN_n}{dt} = P_n + (b_{n-1,n} \cdot \lambda_{n-1} \cdot N_{n-1}) - \lambda_n \cdot N_n$$

Laplace transform (L)

$$N_n(t) = \sum_{k=1}^m \sum_{i=1}^n \left[ \left( \prod_{j=i}^{n-1} \lambda_{j,j+1} \right) \sum_{j=i}^n \left( \underbrace{\frac{N_i^k e^{-\lambda_j(t_{k,irr} + t_{k,cool})}}{\prod_{\substack{p=i \\ p \neq j}}^n (\lambda_p - \lambda_j)}}_{\text{decay}} + \underbrace{\frac{P_i^k (1 - e^{-\lambda_j t_{k,irr}}) e^{-\lambda_j t_{k,cool}}}{\lambda_j \prod_{\substack{p=i \\ p \neq j}}^n (\lambda_p - \lambda_j)}}_{\text{build-up}} \right) \right]$$

- $N_n$  ... Number of isotope n
- $P_n$  ... Production rate of isotope n
- $\lambda_n$  ... Decay constant of isotope n
- $b_n$  ... Branching ratio from isotope n-1 into n
- $k$  ... index of irradiation/cooling cycle