

# Thermo-electrical modelling of the ATLAS ITk Strip Detector

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## Abstract

In this paper we discuss the use of linked thermal and electrical network models to predict the behaviour of a complex silicon detector system. We use the silicon strip detector for the ATLAS Phase-II upgrade to demonstrate the application of such a model and its performance. With this example, a thermo-electrical model is used to test design choices, validate specifications, predict key operational parameters such as cooling system requirements, and optimize operational aspects like the temperature profile over the lifetime of the experiment. The model can reveal insights into the interplay of conditions and components in the silicon module, and it is a valuable tool for estimating the headroom to thermal runaway, all with very moderate computational effort.

*Keywords:* Silicon detector, Thermal runaway, Thermal management, Cooling

## 1. Introduction

The temperatures in silicon detector systems are critically important to their performance. Fundamentally, the leakage current of a silicon sensor has a pronounced temperature dependence

$$I \propto T_S^2 e^{-T_A/T_S}, \quad (1)$$

where  $T_S$  is the sensor temperature and  $T_A \simeq 7000$  K [1]. Leakage currents in the silicon sensor can become particularly significant after irradiation, and the heat generated by these leakage currents, together with the heat from front-end electronic components on the detector, needs to be removed by a cooling system. The capability of the cooling system to remove this heat is limited by the temperature of the local cold sink (typically a circulated fluid) and the thermal impedance of the heat path between the source (electronics and sensor) and the sink. Due to the strong growth of leakage power with temperature, there is a critical temperature  $T_{\text{crit}}$  above which the heat cannot be removed quickly enough, and the detector becomes thermally unstable ('thermal runaway')<sup>1</sup>. Understanding the thermal behaviour and the headroom to thermal runaway is crucial for the design of a silicon detector system. Even before the limit of thermal stability is reached, temperatures in silicon detector systems have a major impact on key system parameters such as power supply capacity and cable dimensions, necessitating an accurate estimate.

In addition to the silicon, there can be aspects of the front-end electronics that have a temperature dependence. In the strip system for the ATLAS Phase-II upgrade [2], which is the subject of this case study, there are two additional temperature-dependent heat sources. The first is a radiation damage effect in the readout electronics, which leads to an increase in the digital power of the chip whose magnitude depends on the total ionisation dose (TID) and the temperature of the chip [2]. This phenomenon was first observed in the ATLAS IBL [3]. The other temperature dependence of a power source stems from the converter chip (FEAST [4, 5]) used in the on-detector DC-DC converter system supplying power to the front-end electronics. The FEAST chip has an efficiency that decreases at higher temperatures; its efficiency also depends on the magnitude of the load current.

<sup>1</sup>In a real detector system, the resulting growth of sensor temperature would be arrested by overcurrent limits in the power supplies, resulting in a reduction of the bias voltage. At the same time, the increased current leads to an increase of the noise, such that the overall result is a degradation of the S/N performance of the system.

31 In principle, the temperatures in the system for a given set of operational parameters (power density,  
32 thermal conductivities, etc.) can be predicted using finite element analysis (FEA) to an accuracy that is  
33 limited only by the quality of the input parameters. However, this is a time-consuming process and can be  
34 prohibitively difficult if a number of local heat sources depend non-linearly on temperature. A simplification  
35 to this problem that allows for an analytical solution in the case of a simple heat source topology has been  
36 developed in [6]. Here we develop this method further to include several temperature-dependent non-linear  
37 heat sources in the front-end electronics. The resulting set of equations cannot be solved analytically anymore,  
38 but the solution can be found with little effort using numerical problem solvers. This enables us to predict  
39 with some confidence the temperatures and power requirements in the ATLAS strip system throughout  
40 Phase-II operation. The results from this prediction have been used throughout the ATLAS strip project to  
41 consistently dimension the different systems (cooling, power, services, etc.), including an appropriate margin  
42 due to the inclusion of a common set of safety factors. This method can be easily adapted to any other  
43 system by adjusting the model to the system-specific geometries and parameters.

44 *1.1. The ATLAS strip system*

45 The strip system for the ATLAS Phase-II upgrade consists of two parts: the barrel system, comprised of  
46 four concentric cylindrical barrels, and two endcaps consisting of six disks each.

47 The basic unit of the detector is the silicon-strip module. Modules are composed of sensors with hybrids  
48 on top, which host the front-end chips as well as circuitry to convert the supply voltage to the chip voltages.  
49 The modules are glued onto both sides of a composite sandwich (local support) that contains two parallel  
50 thin-wall titanium cooling pipes embedded in carbon foam (Allcomp K9) between two facesheets of ultra-  
51 high-modulus carbon fibre (3 layers of K13C2U/EX1515) co-cured together with a Kapton/copper low-mass  
52 tape. During operation, cooling will be achieved by evaporating CO<sub>2</sub> in the cooling pipes with a final target  
53 temperature no higher than −35 °C anywhere along the local support.

54 In the barrel, the local support is called a stave, onto which 14 modules made with square sensors  
55 (96.85 × 96.72 mm<sup>2</sup>) are loaded. Two types of module are used in the barrel: modules with ‘short-strip’  
56 sensors of length 24.1 mm, and ‘long-strip’ sensor modules with 48.2 mm strip lengths. Short-strip modules  
57 are used in the two innermost barrel layers, and long-strip modules in the two outermost barrel layers, mainly  
58 for hit occupancy considerations. A model of the short-strip barrel stave geometry is shown in Fig. 1.

59 The geometry of the barrel stave is uniform along its length, with the exception of the end region of the  
60 stave, where an End-Of-Substructure (EOS) card is mounted on both surfaces. The EOS card shares part of  
61 its heat path with the adjacent module; underneath this module (hereafter referred to as an ‘EOS module’),  
62 the thermal path is degraded by the presence of electrically-insulating ceramic pipe sections. The thermal  
63 and electrical properties of an EOS module are sufficiently different from other modules along the length of  
64 the stave (‘normal modules’) to warrant separate treatment in the thermo-electrical model of the barrel.

65 The endcap system consists of two endcaps composed of 6 disks each. Each disk contains 32 ‘petals,’ the  
66 local substructure depicted in Fig. 2. Both sides of the petal are loaded with 6 modules, each with a distinct  
67 design, located at increasing radius from the beam pipe and labelled R0 through R5 (where ‘R’ stands for  
68 ring). Each endcap module consists of one or two wedge-shaped silicon sensors and a varying number of  
69 front-end chips and DC-DC converters. The EOS card is located adjacent to the R5 module, but the cooling  
70 pipes run directly underneath it without a shared heat path, in contrast to the barrel EOS card. Because  
71 of the unique geometry of each module in a petal, each of the six different types of module is modelled  
72 separately in the thermo-electrical model.

73 *1.2. Radiation environment*

74 A key input to the thermo-electrical calculation is the radiation environment of the strip system, as several  
75 inputs depend on radiation damage effects. The sensor leakage current can be parametrized as a function  
76 of the fluence expressed in 1 MeV neutron-equivalents, and the TID effect on the digital chip current will  
77 be described as a function of the total ionizing dose rate (more details on its dependencies can be found in  
78 Section 7).

79 Predictions for both of these quantities have been generated for each point in the ITk using the FLUKA  
80 particle transport code and the PYTHIA8 event generator (Fig. 3) [7]. In the barrel system, both of these  
81 distributions display a strong dependence on  $r$  but a weak  $z$ -dependence. Accordingly, we make the simplifying  
82 assumption that modules within the same barrel layer have identical fluence and TID, and model four

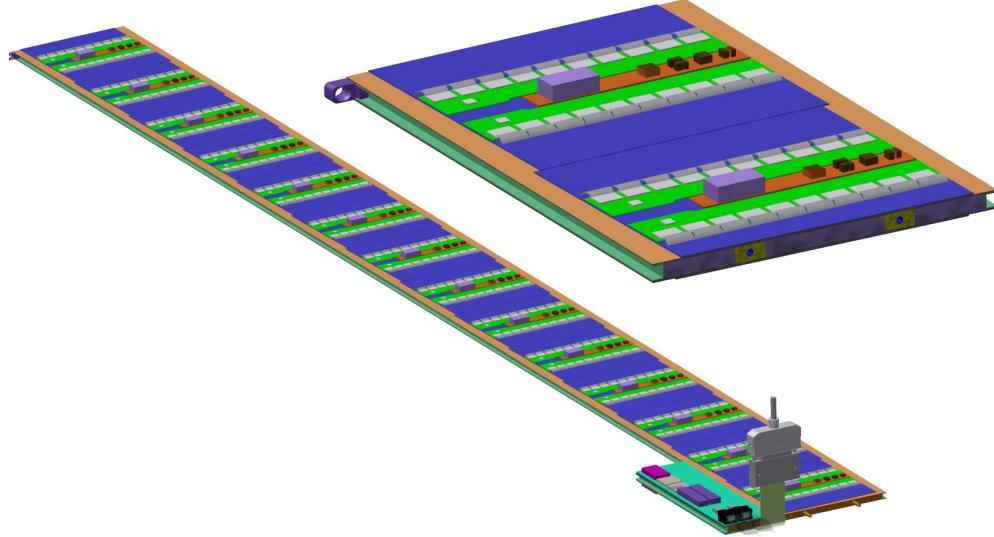


Figure 1: Strip barrel local support geometry. On the left, a complete stave is shown (EOS card in the foreground). The right picture shows a cross-section of the stave with the two cooling pipes visible inside the core.

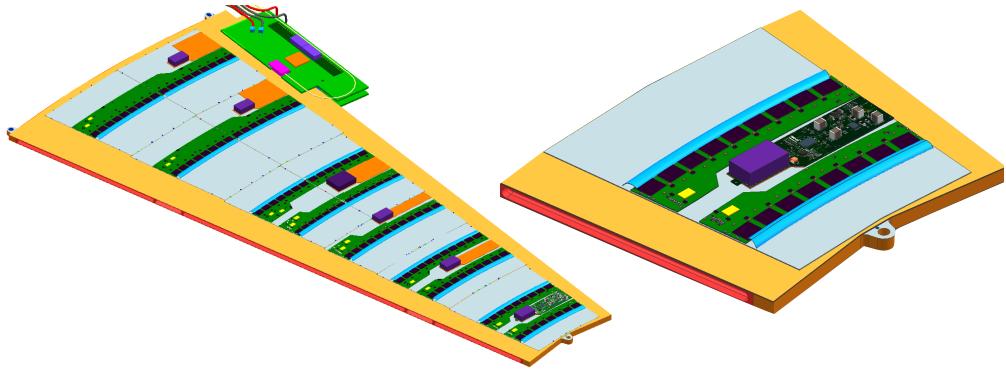


Figure 2: The geometry of the endcap strip petal, featuring 6 distinct module designs. A close-up of the R0 module is shown on the right.

83 different radiation profiles (one for each barrel layer). In the endcaps, the radiation levels vary significantly  
 84 over the length of the petals and from disk to disk; therefore, we model each disk and ring position separately  
 85 (36 in total).

## 86 2. The electrical model

87 The electrical model consists of low-voltage (LV) and high-voltage (HV) circuits, depicted in Fig. 4.  
 88 The LV current (supplied at 11 V) is used to power the hybrid controller chips (HCCs) [2], ATLAS Binary  
 89 Chips (ABCs) [8] and Autonomous Monitoring and Control chip (AMAC) located on PCBs that are glued  
 90 directly onto the surface of the sensor. These chips require between 1.5 and 3.3 V, which are provided  
 91 by the temperature-dependent FEAST DC-DC converter<sup>2</sup> and an LDO (low-dropout) regulator (labelled  
 92 linPOL12V in Fig. 4). The number of chips and converters on each module vary according to the design

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<sup>2</sup> In the final version of the ITk strip system the FEAST chip is meant to be replaced by a chip with similar functionality and behaviour, called the bPOL12V chip; we use the term FEAST throughout this paper.

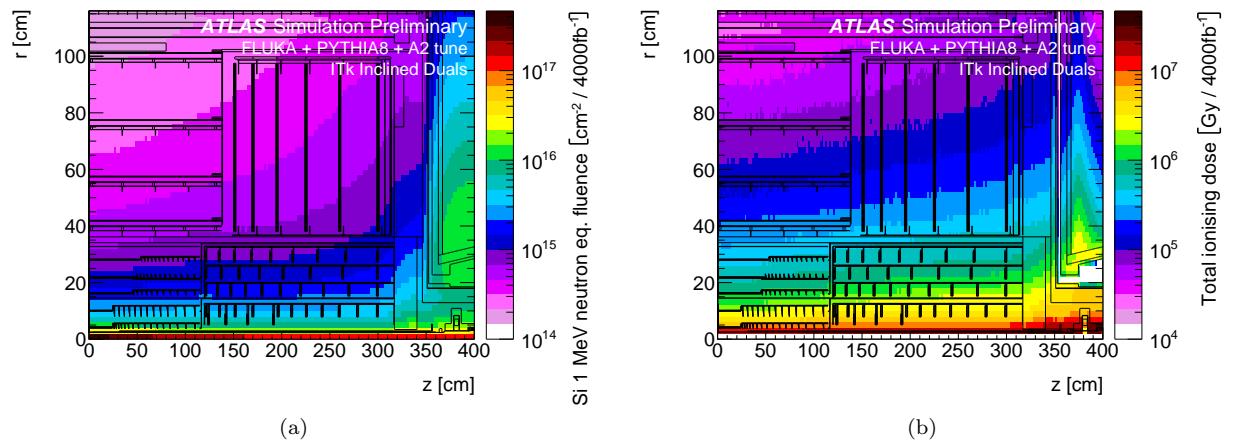


Figure 3: The ATLAS ITk radiation environment. (a) 1 MeV neutron equivalent fluence and (b) total ionizing dose. Both plots are for an integrated luminosity of  $4000 \text{ fb}^{-1}$  [7]. The figures use a rainbow colour gradient, with violet indicating the lowest values and red the highest values of fluence or ionizing dose.

of each different module type (barrel short-strip and long-strip modules, and six different endcap module designs). A barrel or endcap module contains 10–28 ABC chips, 1–4 HCCs, and 1–2 of each of the other components (linPOL12V/FEAST/AMAC).

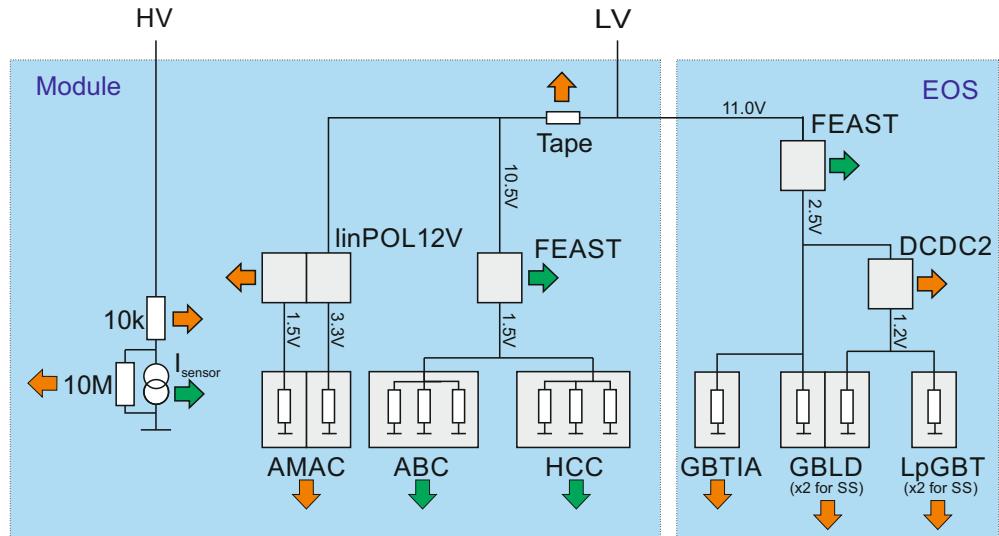


Figure 4: The electrical model of the ITk Strip barrel and endcap modules. Green arrows represent temperature-dependent heat sources, while orange arrows are temperature-independent. Grey squares are chips.

The LV current is also delivered to the EOS card to power various data transfer components: the Gigabit Laser Driver (GBLD), low power GigaBit Transceiver (LpGBT) and Gigabit Trans-impedance Amplifier (GBTIA) chips. A FEAST identical to the one used on the module is used to step the voltage down from 11 V to 2.5 V, and an additional DC-DC converter ('DCDC2') brings the voltage down further for some components. The short-strip barrel staves contain two GBLD and LpGBT chips.

The bus tape, which carries both LV and HV currents, has a small ohmic resistance, which impacts the module in two ways. First, the tape itself will generate some heat according to the amount of current passing through it; this source of heat is accounted for in the model, however the contribution to the total module

104 power is negligible. Second, due to the voltage loss along the traces, there is a slight reduction in voltage  
105 supplied to successive modules along the substructure. The treatment of this effect is slightly different in the  
106 barrel and endcap models: in the barrel, the voltage delivered to every module is averaged to 10.5 V; in the  
107 endcap, the  $\Delta V$  is estimated based on the calculated expected power loss along the tape for each module  
108 and varies between 10.8 and 11 V. In both the barrel and endcap systems, the impact of using a different  
109 treatment is small.

110 Finally, the HV current provides the bias voltage on the silicon sensors. An HV multiplexer switch  
111 (HVMUX) can be used to disconnect the sensor from the bias line (it requires a  $10\text{ M}\Omega$  resistor parallel to  
112 the sensor in order to function). Two HV filters with an effective resistance of  $10\text{ k}\Omega$  are situated in series  
113 with the sensor. The nominal operating voltage of the sensor is expected to be 500V, but the system is  
114 designed to handle a bias voltage of up to 700V.

### 115 3. The thermal model

116 The thermal network consists of heat sources (some of which are temperature-dependent) and thermal  
117 resistances. The latter are given by the properties of the mechanical design (heat conductivities of the  
118 materials) and the geometry of the heat path. The geometry is generally 3-dimensional, but it is the strategy  
119 of the simple network models to lump the 3D behaviour into one thermal resistance parameter. In the models  
120 discussed here, we have used a granularity corresponding to single detector modules for which the thermal  
121 resistance has been modelled. The temperatures in the model are then given for the nodes in the network in  
122 analogy to the potentials in an electrical network<sup>3</sup>.

123 The complexity of the thermal network used in this study, depicted in Fig. 5, is given by the variety of  
124 temperature-dependent heat sources in the ATLAS strip system: the digital power for each type of chip, the  
125 FEAST chip providing the on-detector DC-DC conversion, and the sensor leakage power. In the ATLAS  
126 ITk strip modules, all of these components are located on top of the sensors, such that the heat generated  
127 in them flows through the sensor into the support structure, the stave (barrel) or petal (endcap) core with  
128 the embedded cooling pipe. In the network model, the heat flow from these sources combines and travels  
129 through a common impedance  $R_M$  to the sink at a temperature  $T_C$ . For each of the temperature-dependent  
130 heat sources (ABC, HCC, FEAST and the sensor) we have added a resistance from the common temperature  
131  $T_{\text{mod}}$  to allow for a finite and different heat path for each of them. Finally, the EOS card adjacent to the last  
132 module on the barrel stave or endcap petal is modeled as an additional source of heat with an independent  
133 impedance for its unique thermal path.

134 This is a more complex thermal network than the one studied in Ref. [6], for which an analytical solution  
135 for the determination of thermal stability is given. In particular, because of the non-linear temperature  
136 dependence of some of the heat sources, it is not possible in the present case to solve the set of equations  
137 describing the model analytically. However, the set of equations is still sufficiently small to solve numerically  
138 using any modern mathematical software. We do not expect the numerical method to introduce a loss of  
139 precision on the scale of the model approximations.

### 140 4. Obtaining thermal impedances using FEA

141 The cooling path between the sources dissipating electrical power and the cooling fluid is 3-dimensional  
142 and includes components with orthotropic thermal conductivity. Hence the prediction of temperature at any  
143 node of the model requires a 3D thermal FEA [9, 10]. However, the thermal conductivities of the components  
144 along the path are approximately constant, so that the temperature rise  $\Delta T_i$  above the coolant temperature  
145 of any node  $i$  ( $i = \text{ABC, HCC, AMAC, FEAST, tape, RHV, or sensor}$ ) in the thermal network model is  
146 adequately described by a linear sum of contributions from individual sources, i.e:

$$\Delta T_i \equiv T_i - T_C = R_i P_i + (R_C + R_M) \sum_j P_j, \quad (2)$$

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<sup>3</sup>The similarity of the electrical and thermal networks we are using in our approach has been exploited before: historically, Fourier's description of heat conduction pre-dated and inspired Ohm's work on electrical resistive networks. Here we followed the opposite direction.

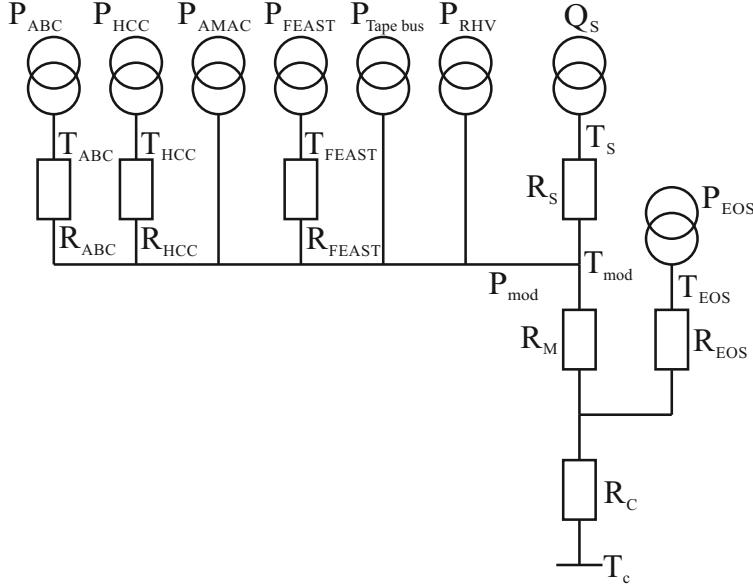


Figure 5: Thermal network model.

where the index  $j$  runs over all powered nodes. (We have momentarily disregarded the contribution from the EOS card.) The expression includes an impedance term  $R_{CM} \equiv R_C + R_M$ , the common thermal pathway shared by the components. We omit the sensor thermal impedance in the definition of  $R_{CM}$  because its contribution is negligible (roughly  $0.02\text{ }^{\circ}\text{C/W}$ ).

In order to extract the thermal impedances for the thermal network model, the finite element model is run multiple times, with each heat source (or group of similar sources) switched on in turn with a representative amount of heat. In each of these cases, the temperature is calculated for all nodes in the thermal network model (Fig. 5). The temperature of a node is here taken as the average of the temperatures for all the grid points in the FEA model within the volume of the object corresponding to the node<sup>4</sup>. The thermal impedances are then obtained from a fit of Eq. 2 using the temperature data for all nodes for all cases of heat injection.

Because of the nature of the network, the fitted value for the common impedance  $R_{CM}$  is determined by the observed temperature rises of components where no heat is injected. The linearity of this relationship is illustrated in Fig. 6. The value of each component-specific impedance is determined from the temperature rise observed when heat is injected into that component. The linear approximation of the model reasonably describes the FEA simulation, and the level of disagreement, discussed below, is taken as an uncertainty that is assessed as a safety factor as described in Section 7.2.

For the barrel, this procedure is performed for both an EOS module and a normal module, for short-strip and long-strip module types and for 6 different heat sources (24 module FEA simulations in total). In all cases, the agreement of the network temperatures using the thermal impedances from the fit with the data from FEA is better than  $0.5\text{ }^{\circ}\text{C}$  for all nodes. The thermal impedance from the sensor to the sink ( $R_{CM}$ ) is consistently between  $1.1$  and  $1.4\text{ }^{\circ}\text{C/W}$ , but higher values (between  $10$  and  $20\text{ }^{\circ}\text{C/W}$ ) are found for other impedances in the network ( $R_{HCC}$  and  $R_{FEAST}$ ), mostly because these are for components with a small footprint constituting a bottleneck for the heat flow. A 10% uncertainty is assigned to the thermal impedances in the barrel modules for the purposes of assessing safety factors.

For the endcap, the procedure to determine the thermal impedances for each of the 6 module types used simulations of an endcap petal with 3 different heat sources (requiring 3 FEA simulations of a full petal).

<sup>4</sup>This is particularly interesting in the case of the sensor, which fills a large volume, with a potentially large range of temperatures. In Ref. [6] the analytic model parameters were extracted from the maximum sensor temperature predicted by FEA. Further comparisons after the publication of that paper indicated that the thermal stability limit is predicted more accurately if the average sensor temperature is used.

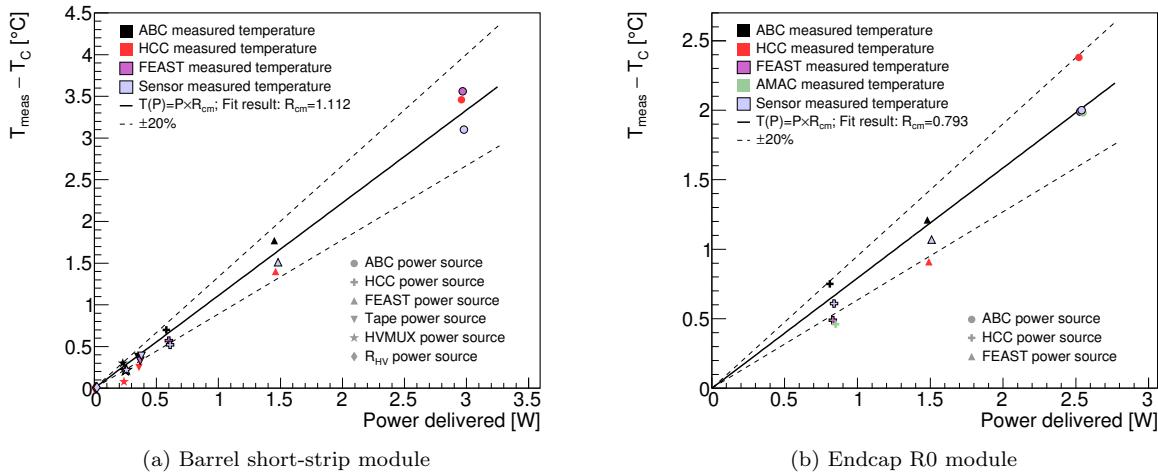


Figure 6: The relationship between the temperature rise observed in the FEA for a specific component and the heat injected in another component. The slope of the fitted line is the estimate for  $R_{CM}$ . (a) The fit for a short-strip barrel module adjacent to the EOS card. (b) The fit for the endcap R0 module. For each data point marker, the source of power is indicated by the shape, and the measured component is indicated by the colour (black: ABC, red: HCC, magenta: FEAST, green: AMAC, light blue: silicon sensor). The dashed lines represent a  $\pm 20\%$  error band on the fit for  $R_{CM}$ .

$R_{CM}$  ranges from 0.6 to 1.4 °C/W, with other nodes between 5 and 20 °C/W. Because the location of powered components is more irregular on an endcap module, the difference between the predicted temperatures of the linear network and the FEA can reach up to 1.2 °C for key temperature-dependent nodes. Therefore, a 20% uncertainty is assigned to the thermal impedances in the endcap modules.

There are two recognised departures from linearity of the thermal path: the rise in thermal conductivity of the silicon sensor with decreasing temperature, and the rise in heat transfer coefficient (HTC) of the evaporating CO<sub>2</sub> coolant with increasing thermal flux. The FEA models are run using mean values for these quantities appropriate to the operating conditions, and the thermo-electrical model results are insensitive to the variations expected in practice. However, if this level of realism is required and if reliable parametrizations for these dependencies can be obtained, then the inclusion of such variations in the model is possible.

## 5. Other model inputs

The three temperature-dependent elements of the thermo-electrical model—the radiation-induced digital current increase in the front-end chips, the efficiency of the FEAST-controlled DC-DC converter, and the sensor leakage current—are described in this section. Each effect is studied experimentally and fit with functional forms in order to accurately represent them in the model. The uncertainty in the experimental data, and in our modelling assumptions, are estimated here and considered in the evaluation of safety factors, described in detail in Section 7.2.

### 5.1. DC-DC converter

The DC-DC converter, which features an air-core inductor and is controlled by the FEAST chip, supplies a low-voltage (1.5 V) current to the ABC130 and HCC front-end chips on the module. The efficiency of the FEAST depends on its temperature as well as the output (load) current load delivered to the front-end chips. To correctly model the FEAST efficiency, experimental measurements have been performed to characterize the dependence and fitted with a functional form<sup>5</sup>.

<sup>5</sup> Because of the nearly identical functionality and behaviour expected of the bPOL12V chip that will replace the FEAST in the final version of the ITk strip system, the FEAST data and the resulting fit shown here is expected to accurately reflect the bPOL12V case.

197 For the measurement, the FEAST power board was glued to an aluminum cold plate, cooled with CO<sub>2</sub>,  
 198 and powered with the nominal working input and output voltages (11 V input, 1.5 V output). The temperature  
 199 of the FEAST was measured with an NTC (negative temperature coefficient) thermistor and a PTAT  
 200 (proportional to absolute temperature) sensor residing on the FEAST for a range of load currents up to the  
 201 maximum design current of 4A.

202 The data was then fit with a function with sufficient parameters to ensure reasonable agreement; the  
 203 choice of functional form has no physical interpretation. Fig. 7 depicts the FEAST efficiency data and the  
 204 parametrized fit used in the model. The parametrization fits the data with an accuracy better than 1%;  
 205 this uncertainty in the FEAST efficiency modelling is small compared to other uncertainty sources, and is  
 206 therefore neglected in our model.

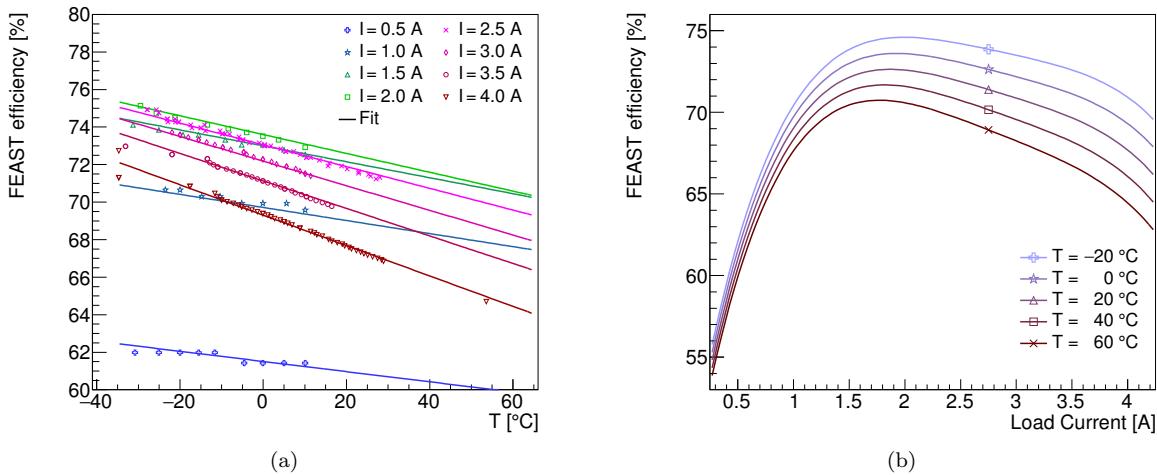


Figure 7: The FEAST efficiency model based on experimental data. (a) The experimental data points characterizing the FEAST efficiency are plotted as markers and coded by colour and marker style for load current. The data is compared to the analytic fit, evaluated in curves of equal current. (b) The same analytic fit, presented as a function of current load for curves of equal temperature.

### 207 5.2. Digital current increase of chips using 130 nm CMOS technology

208 The ABC and HCC chips, designed using IBM 130 nm CMOS 8RF technology, are known to suffer  
 209 from an increase in digital current when subjected to a high-radiation environment [2]. This phenomenon,  
 210 known as the ‘TID bump,’ is well-studied [11, 12] and has a characteristic shape whereby the effect reaches  
 211 a maximum as a function of the accumulated dose and then gradually diminishes (see Fig. 8).

212 In an effort to characterize the nature of the TID bump in the ABC and HCC chips empirically, many  
 213 irradiation campaigns have been conducted using a variety of radiation sources, testing the effect at different  
 214 temperatures and dose rates. The data collected from these studies was used to develop a model of the  
 215 TID bump that estimates the digital current increase given the total ionizing dose, the dose rate, and the  
 216 operating temperature of the chip. The data was fitted using a 5-parameter model which was guided by  
 217 assumptions on the underlying damage mechanisms, although the fit function had to be adjusted to achieve  
 218 a satisfactory match to the data. The result of this parametrization, which is depicted in Fig 8, is used as  
 219 an input to the thermo-electrical model in order to correctly model the ABC and HCC currents. The TID  
 220 bump is assumed to fully apply to the HCC digital current, and apply to 69% of the ABC digital current<sup>6</sup>.

221 The TID bump displays certain key features, which are reflected in the parametrization: first, the effect is  
 222 larger at colder temperatures and higher dose rates. This means it can be mitigated by operating the chips at

<sup>6</sup> The primary source of the additional digital current is leakage in a certain type of transistor on the chip, which is only used in parts of the ABC. We have introduced the fraction factor described in the text to allow flexibility to scale this leakage current for different chip designs.

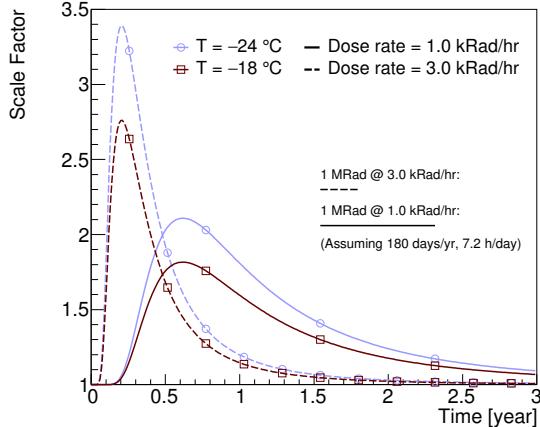


Figure 8: Parameterization of the impact of ionizing radiation on the magnitude of the front-end chip digital current (the TID bump), presented as a function of time elapsed during detector operation. The current is multiplied by a scale factor that is modeled as a function of total ionizing dose, dose rate, and temperature, using a fit to experimental data. The figure presents four scenarios using two different chip temperatures and dose rates. The relationship between time and total ionizing dose is calculated assuming the detector operates for 180 days per year, 7.2 hours per day. The time that it takes to reach 1 MRad of total ionizing radiation is indicated on the figure for both dose rates.

higher temperature (note that the dose rate is determined by the LHC operational conditions). Second, the figure illustrates how chips receiving different dose rates will reach their maximum digital current increase at different times. This feature is particularly important when modelling the total power consumed by the barrel and endcap systems. In both systems, the dose rate varies significantly depending on the position of the module in the detector. The effect means that the maximum system power will be smaller than the sum of the maximum power of each module, as each chip reaches its maximum at a different point in time (see Section 7.4).

The TID bump is an important source of uncertainty in our model. The experimental data exhibit a relatively large variation in the TID bump effect, in particular between different batches of the same type of chip delivered by the manufacturer, suggesting an unknown effect in the fabrication process. To estimate the uncertainty in the TID bump, the parametrized function is fit again using only the worst-performing data (defined as having the largest TID bump effect). This ‘pessimistic’ parametrization is used as a safety factor to estimate the detector performance in worst-case scenarios.

The irradiation of individual chips have typically been performed at constant dose rate and temperature. However, both of these parameters will vary as a function of time in the scenarios that we attempt to model. In our current parametrization, we use only the instantaneous value of these two parameters, thus neglecting any possible history of the TID effect for a given chip. We also ignore any short-term effects due to variations in the dose rate on the scale of hours or days. This approach is mandated by the lack of more varied experimental data and the absence of a good theoretical model for this effect. This probably constitutes the largest source of uncertainty in our model.

### 5.3. Radiation-dependent leakage current

The radiation-induced sensor leakage current can be parametrized as a function of the hadron fluence expressed in 1 MeV equivalent neutrons. We have used linear parametrizations obtained from fits to experimental data taken at 500 V and 700 V at  $-15^{\circ}\text{C}$ , and scale them to a given sensor temperature using Eq. 1.

## 6. Running the model

The thermo-electrical model constructs a profile of the sensor module operation conditions over the lifetime of the detector in the following manner. First, the total module power (including all components,

but excluding the sensor leakage power) and the sensor temperature assuming no leakage current ( $T_0$ ) are calculated using a reasonable set of initial component temperatures. The initial value for the module power is used to solve for the sensor power and temperature accounting for leakage current, using the thermal balance equation and the relationship from Eq. 1. Using this calculated sensor leakage current and temperature, the power and temperature of the module components are updated given the initial (year 0, month 0) startup parameters.

Next, the module conditions of the following month (year 0, month 1) are calculated. Using the component temperatures calculated from the previous month and the operational parameters (ionizing dose and dose rates) from the current month, the module total power (excluding sensor leakage) is again calculated, and subsequently the sensor temperature and leakage current are computed. Following this, the module component temperatures and power values are derived for this month. This process is repeated in one-month steps until the final year of operation, or until a real solution for the sensor temperature does not exist, indicating that thermal runaway conditions have been reached.

In the barrel subsystem, the above procedure is performed four separate times to represent the radiation conditions of the four barrel layers located at different radii from the beam axis<sup>7</sup> for both a normal and an EOS-type module. Thus, eight modules are simulated in total for the barrel (4 layers  $\times$  normal/EOS), and they are combined in their proper proportion to simulate the entire barrel system.

In the endcap subsystem, the total ionizing dose and dose rates vary significantly depending on the position of the module; furthermore, the design of each module on a petal differs significantly. Therefore, all 36 module types (6 rings  $\times$  6 disks) are simulated independently, and combined to represent the full endcap.

We have implemented this algorithm separately for the barrel and endcaps. In both cases, the calculation for a set of operating conditions over the full lifetime of the LHC, in steps of one month, takes between 5 and 10 minutes on a standard PC. The results represent the equivalent of simulating 1152 barrel modules and 5184 endcap modules, if the same time granularity would be required<sup>8</sup>. For comparison, the processing time for a single steady-state FEA simulation of an endcap petal is about 20 minutes (not taking into account the time needed to set the parameters of the simulation). Thus, the thermo-electrical model enables a quick turn-around for systematic studies of the parameter space that would not be possible using FEA alone.

## 7. Outputs of the thermo-electrical model

The thermo-electrical model provides a wide range of predictions for the operation of the strip system. A detailed discussion of all results is beyond the scope of this article; instead, we present here a subset of the results to demonstrate the capabilities and use of the thermo-electrical model for the design of the detector system.

### 7.1. Operational scenarios

To study the different aspects of our predictions for the operation of the ITk strip system throughout its lifetime, we performed the calculation of the system parameters over the expected 14 years of operation in monthly steps as outlined in Section 6. Time-dependent operational inputs to the calculation were taken from the expected performance of the HL-LHC (Fig. 9a) [13]. For the cooling, which can be adjusted during data taking using detector control systems, we studied flat (constant) coolant temperature profiles ranging from 0 °C to –35 °C, the lowest evaporation temperature achievable with the ITk evaporative CO<sub>2</sub> cooling system. We also studied a ‘ramp’ scenario in which the coolant temperature starts at 0 °C and is gradually lowered down to –35 °C over the course of 10 years (Fig. 9b).

### 7.2. Safety factors

To ensure the robustness of the system design against uncertainties in the assumptions used in the model, we also evaluate the model using a set of input parameters with some key inputs degraded. The set of safety factors used is given in Table 1. Each safety factor has been estimated individually based on experience, the

<sup>7</sup>The correct module type, short-strip in the inner two layers and long-strip for the outer two layers, is used for each layer.

<sup>8</sup>Recall from Section 4 that the thermo-electrical model requires 24 barrel module FEA simulations and 3 endcap petal FEA simulations to extract the thermal impedances, but this procedure is only performed once.

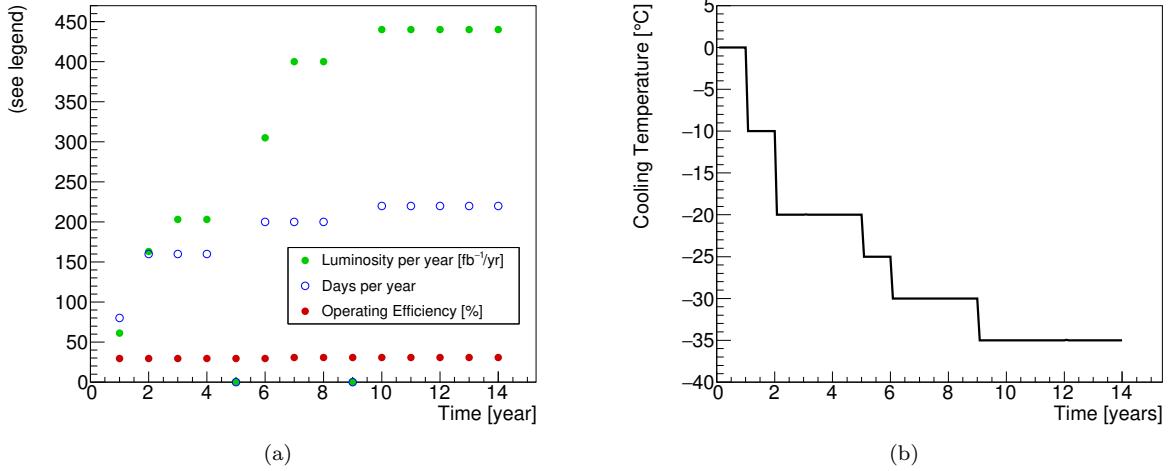


Figure 9: (a) Expected HL-LHC performance and (b) ‘cooling ramp’ scenario for the coolant temperature. Year-long shutdowns of the LHC are anticipated in years 5 and 9.

complexity of the system aspect described by the parameter, and from available data or the absence of such data. Note that the model can be evaluated with all the safety factors listed in Table 1 used together, a situation that is unlikely to occur in the real system, to provide a worst-case estimate for the performance of the ITk strip system. The individual effects of the different safety factors are demonstrated in Fig. 10.

Table 1: Safety factors.

Safety factor	Value	Reason
Fluence	50%	Accuracy of fluence calculations and uncertainties in material distributions
Thermal impedance	10% barrel, 20% endcap	Local support build tolerances, thermal network assumptions
Digital current	20%	Final chip performance and parametrization of TID effect
Analog current	5%	Final chip performance
Tape electrical impedance	10%	Electrical tape manufacturing tolerances
Bias voltage	700 V	Increased bias voltage from nominal 500 V to maintain S/N
TID parametrization	Nominal/Pessimistic	Different data sets for fit of TID bump

It is important to note that combining multiple safety factors can have a compounding effect on the system. As an example, the effect of an increased bias voltage combined with a larger digital current will result in a much higher sensor leakage current at the detector end-of-life than either situation occurring individually. The analytical model presented here allows for scenarios like these to be examined quickly and effectively.

### 7.3. Module properties

Several module properties predicted by the thermo-electrical model are shown in Figs. 11 and 12 for the barrel system. The different radiation-dependent effects occur on different timescales. The maximum in the digital chip power due to the TID effect occurs relatively early (in year 1 to 4), although the bump has a long tail, particularly in the outer layers of the barrel. The sensor leakage power, on the other hand, grows towards the end of the lifetime of the ITk. If the leakage current continued to increase in the case of further irradiation, or if the cooling temperature were raised, this growth would ultimately lead to thermal runaway. Due to the radial dependence of the radiation environment, the radiation-induced effects are most pronounced in the innermost barrel layers.

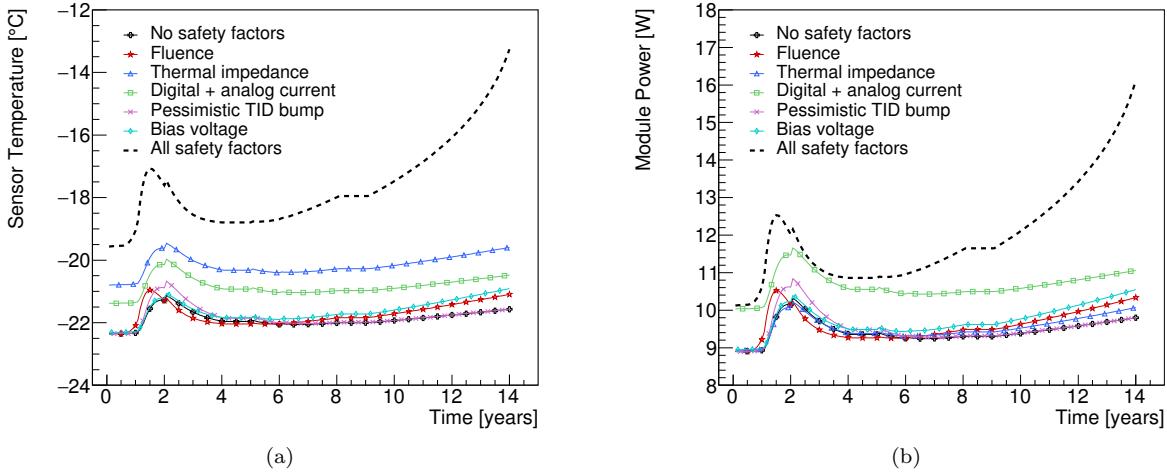


Figure 10: Comparing the impact of different safety factors on (a) the sensor temperature and (b) the module power for the endcap R3-type module, using a flat cooling scenario ( $-30\text{ }^{\circ}\text{C}$ ). The dotted line depicts the effect of all safety factors applied at once.

#### 314 7.4. System properties

315 One of the key concerns for the design of the strip system is thermal stability of the system. If the cooling  
 316 temperature is too high to limit the leakage power from the radiation-damaged sensors to a level where  
 317 the heat can still be removed, the system is unstable (it goes into ‘thermal runaway’). To find the cooling  
 318 temperature  $T_C$  at which this condition is reached, we make repeated simulations of the ITk strip system  
 319 using the thermo-electrical model, with each simulation representing the full 14-year operation of the ITk at  
 320 a fixed  $T_C$ . Between simulations,  $T_C$  is increased in steps of  $5\text{ }^{\circ}\text{C}$  until the model finds thermal runaway.  
 321 In the numeric evaluation of the thermo-electrical model this manifests itself in the absence of a solution to  
 322 the system of equations. In the endcap strip system, this occurs at a cooling temperature of  $-15\text{ }^{\circ}\text{C}$  under  
 323 nominal conditions (i.e. with no safety factors applied); in this scenario, thermal runaway would be reached  
 324 in the 12<sup>th</sup> year of operation. With all safety factors applied, thermal runaway would occur at a cooling  
 325 temperature of  $-25\text{ }^{\circ}\text{C}$  (in year 10). In the barrel system, where the radiation environment is slightly less  
 326 intense, the conditions for thermal runaway occur at the same cooling temperatures, but a few years later  
 327 than in the endcaps: in the final year of operation and a cooling temperature of  $-15\text{ }^{\circ}\text{C}$  under nominal  
 328 conditions, and at  $-25\text{ }^{\circ}\text{C}$  (in year 13) with safety factors applied. As the design cooling temperature of  
 329 the ITk cooling system is  $-35\text{ }^{\circ}\text{C}$ , we have confidence that the ITk strip system has a sufficient margin for  
 330 thermal stability.

331 Beyond the issue of stability, the thermo-electrical model delivers predictions for the development of  
 332 current and power requirements for the overall system. Some of the predictions are shown in Fig. 13. Again,  
 333 the different timescales of the various radiation-induced effects are visible; ignoring this time dependence  
 334 could lead to over-specification of some system aspects. Taking Fig. 13b as an example, the average module  
 335 power (indicated by the thick black line) is 6.6 W in the beginning of operation and reaches 8.0 W at the TID  
 336 bump, in the second year. If we naively summed the maxima of the TID bumps, neglecting time dependence,  
 337 we would arrive at 8.6 W, overestimating the power at the TID bump by 7.5% and overestimating the effect  
 338 of the TID bump by 43%. The difference, multiplied in the endcap by 4608 modules, amounts to nearly  
 339 3 kW of power and impacts the specifications of e.g. the cooling system.

340 The predictions from this model are now used throughout the strip project to consistently size the power  
 341 supply and cooling systems. Including safety factors in the predictions gives us some confidence that the  
 342 designs are robust; by using commonly agreed safety factors, we ensure a consistent use of safety factors  
 343 throughout the project and prevent safety factor creep.

344 Because of the different timescales for the peak power due to the TID effect and the radiation-induced  
 345 sensor leakage, there is room to optimize the cooling temperature profile to minimize the total power in the  
 346 strip system while avoiding thermal runaway. The thermo-electrical model is a powerful tool to plan such an

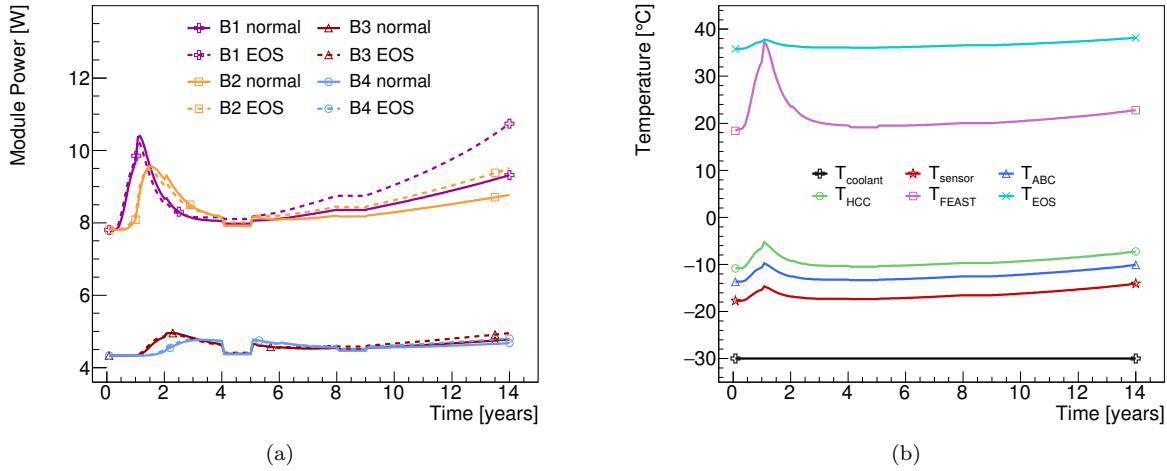


Figure 11: Examples of barrel module performance predictions for a flat cooling scenario ( $-30^{\circ}\text{C}$ ) including safety factors. (a) Power per module. (b) Temperatures for different nodes of an end-of-stave barrel module in the innermost barrel. The discontinuities in year 5 and 9 are due to anticipated year-long shutdowns of the LHC.

347 optimized cooling profile. In fact, the cooling ‘ramp’ scenario introduced in Section 7.1 is the result of such  
 348 an optimization. In this scenario, depicted in Fig. 14, the cooling temperature begins at a relatively high  
 349 value ( $0^{\circ}\text{C}$ ) to minimize the impact of the TID bump in the first two years of operation, thus avoiding a  
 350 peak in the module power (see Fig. 14a). In subsequent years,  $T_C$  is steadily decreased to maintain a sensor  
 351 current at or below about 1 mA, as illustrated in Fig. 14b, in the interest of both minimizing the module  
 352 power and avoiding thermal runaway.

## 353 8. Model performance verification

354 The accuracy of the predictions of the thermo-electrical model is affected by two major factors: the quality  
 355 of the input parameters, and the error introduced by reducing the complex 3D geometry into a linear thermal  
 356 impedance network. The former has been discussed throughout this paper where the different inputs have  
 357 been presented. For the latter, we have studied the agreement of predictions from the network model with  
 358 the more accurate results obtained from FEA for selected states of the system.

359 To verify the level of this agreement, we have calculated the sensor temperature curve for a barrel EOS-  
 360 type module up to thermal runaway, both in the full FEA and in the network model. For this exercise, we do  
 361 not vary any of the input parameters in the model other than the sensor leakage power with its temperature  
 362 dependence. The resistor values in the network model are the same as used throughout for our model,  
 363 obtained as described in Section 4. For the power from the various electronics components, the FEAST  
 364 efficiency and the TID scale factor we have used representative nominal values.

365 Because the variable model inputs are kept constant for this study, we can reduce the complex thermal  
 366 network to its Thévenin equivalent, which is identical to the network studied in Ref. [6], and use the analytical  
 367 expressions given there. The reduced network is described by the base temperature  $T_0$ , defined as the sum  
 368 of the coolant temperature and the temperature rise due to the front-end electronics alone, and the total  
 369 thermal impedance  $R_t$  from the sensor to the coolant. Using the nominal resistances and representative power  
 370 numbers from the module,  $T_0 = -21.9^{\circ}\text{C}$  and  $R_t = 1.132 \text{ K/W}$  in the network model, compared to  $-22.4^{\circ}\text{C}$   
 371 and  $1.147 \text{ K/W}$  obtained directly from the FEA. The comparison of the predicted sensor temperatures for  
 372 both cases is shown in Fig. 15. Despite a large temperature variation of about  $10^{\circ}\text{C}$  across the sensor,  
 373 the network model runaway prediction agrees well with the FEA<sup>9</sup>. This gives us confidence that the use

<sup>9</sup>The critical temperature here is  $-12.4^{\circ}\text{C}$ , which is higher than the numbers given in Section 7.4, because the study here

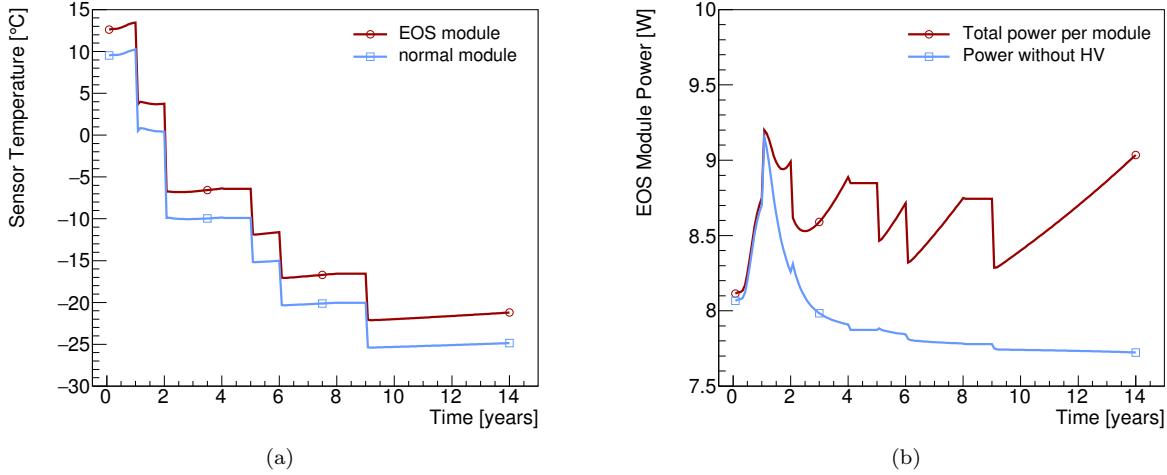


Figure 12: Examples of barrel module performance predictions for the ramp cooling scenario including safety factors. (a) Sensor temperature in the innermost barrel modules. (b) Power in an end-of-stave barrel module in the innermost layer.

374 of a thermal network model is not likely to significantly degrade the predictions beyond the uncertainties  
375 introduced by other inputs to the model.

## 376 9. Conclusions

377 We have developed a model of the ATLAS ITk strip system that is based on the interplay between a  
378 thermal and an electrical network model. The set of equations in the model can be numerically solved using  
379 standard data analysis software in a short time, allowing for a quick turn-around for systematic studies of the  
380 system performance. The complexity of these networks is given by the number of interconnected components  
381 between the networks, many of which have a non-linear dependence on the temperature or electrical power.  
382 This approach can be easily adopted for any other silicon detector system.

383 In the case of the ATLAS strip system, several temperature-dependent heat sources had to be modeled.  
384 In addition to the sensor leakage current, these are the radiation-induced increase of the digital front-end  
385 power ('TID bump') and the efficiency of the DC-DC conversion system. The outputs of the model give  
386 us confidence that the ITk strip system will be thermally stable until the end of LHC Phase-II operation,  
387 even with the inclusion of safety factors on key inputs. Furthermore, the model provides information for  
388 benchmark system parameters like cooling, supply power and currents in power cables, which is used in the  
389 specification of these systems. The use of the model outputs throughout the strip project ensures consistent  
390 specifications, including a common strategy on safety factors. Using the thermo-electrical model, we can also  
391 propose an optimized cooling temperature 'ramp' scenario, which stabilizes leakage power throughout the  
392 lifetime of the experiment while minimizing the TID bump.

393 We have verified the performance of the thermal network model compared to a full FEA treatment, and we  
394 are confident that the level of disagreement is smaller than the uncertainty introduced by the model inputs.  
395 Among the inputs, the most likely source of uncertainty stems from the limitations in our understanding of  
396 the parametrization of the TID effect.

## 397 10. Acknowledgements

398 The evaluation of the thermo-electrical model depends critically on the input parameters to the model.  
399 To capture the whole of the system, these need to distill all that is known of the system, and we are therefore

ignores temperature effects such as the FEAST efficiency, which can only be modelled in the network model.

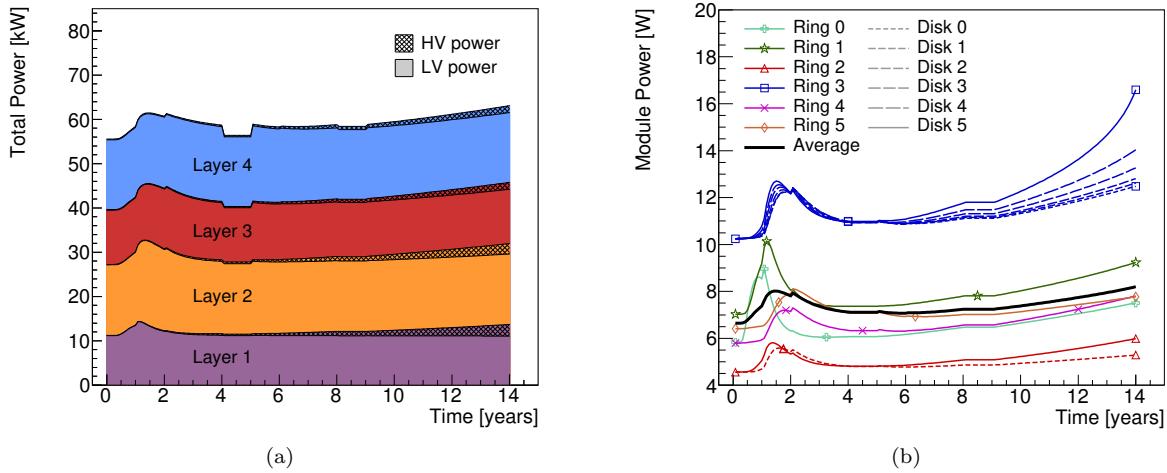


Figure 13: Examples of system performance predictions. (a) Barrel total power requirements. The plot shows the stacked power requirements for the four barrel layers (purple: layer 1, orange: layer 2, red: layer 3, blue: layer 4). Full colour indicates power from the front-end electronics, and hatched parts are contributions from HV power for the four barrels. The discontinuities in year 5 and 9 are due to anticipated year-long shutdowns of the LHC. (b) The power requirements for 12 of the 36 simulated endcap modules, labelled according to their ring type and disk position. (Some modules are omitted to improve the clarity of the figure.) The solid black line indicates the average module power. Both predictions use a scenario with flat  $-30^{\circ}\text{C}$  cooling and including all safety factors.

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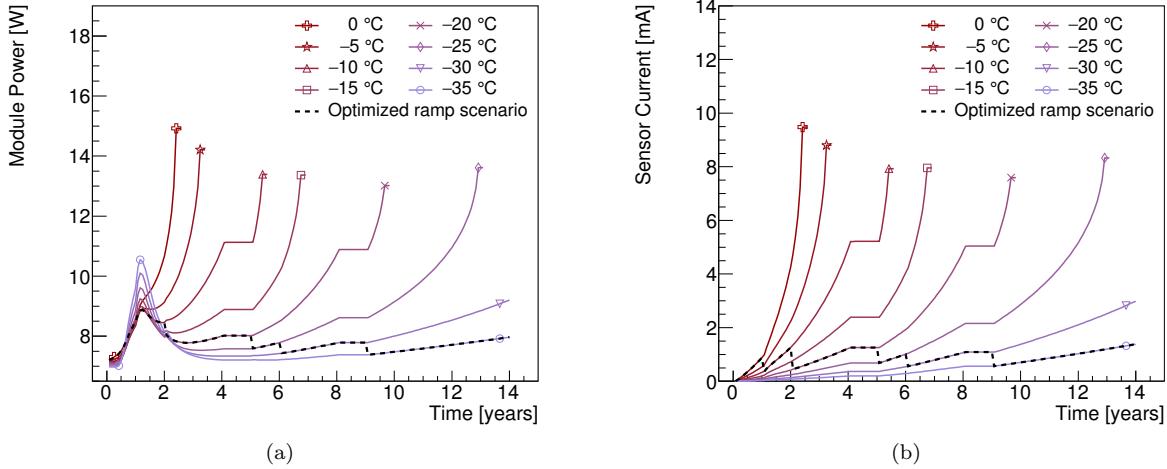


Figure 14: (a) Total power and (b) sensor leakage current of the endcap R1-type module for eight different flat cooling profiles, ranging from 0 °C to –35 °C, as well as the cooling ramp scenario specified in Fig. 9b (dashed curve). The curves that are discontinued before year 14 correspond to scenarios that have reached thermal runaway. The cooling ramp scenario has been selected to minimize the module power while keeping the sensor leakage current stable throughout the lifetime of the ITk. All safety factors are applied in these plots.

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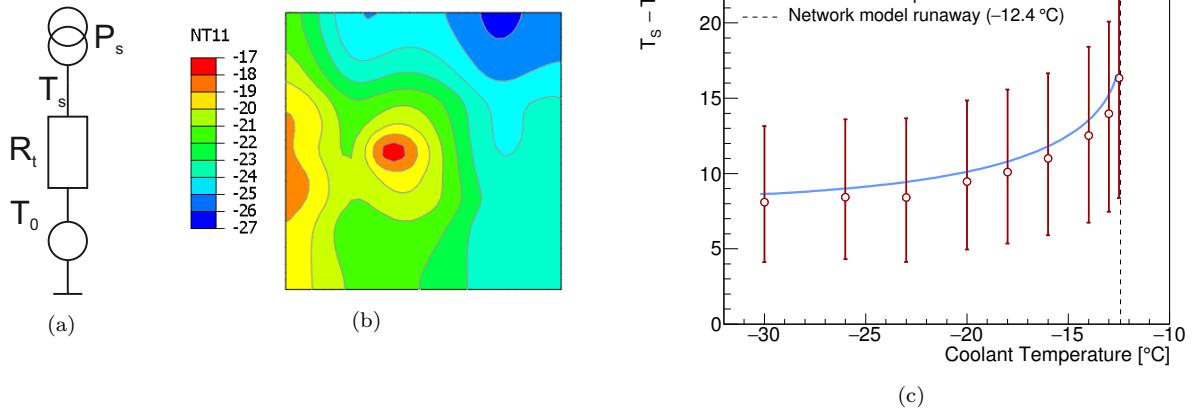


Figure 15: (a) Thévenin equivalent of the thermal network. (b) Result of sensor surface temperature calculations using FEA, assuming zero sensor power. The EOS card is to the left of the module, and the cooling pipes run from top to bottom about a quarter of the module width from each edge. The figure uses a rainbow colour gradient, with blue indicating the lowest temperatures and red the highest temperatures. (c) Difference of average sensor and coolant temperature, comparing FEA (red points) and the network model prediction (blue curve). The bars on the FEA data indicate minimum and maximum sensor temperature. The dotted vertical line indicates the critical temperature derived analytically using the network model ( $-12.4$  °C).