

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_{crit} above which the heat cannot be removed quickly enough, and the detector becomes thermally unstable ('thermal runaway')¹. 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]) 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.

¹In 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.

In principle, the temperatures in the system for a given set of operational parameters (power density, thermal conductivities, etc.) can be predicted using finite element analysis (FEA) to an accuracy that is limited only by the quality of the input parameters. However, this is a time-consuming process and can be prohibitively difficult if a number of local heat sources depend non-linearly on temperature. A simplification to this problem that allows for an analytical solution in the case of a simple heat source topology has been developed in [5]. Here we develop this method further to include several temperature-dependent non-linear heat sources in the front-end electronics. The resulting set of equations cannot be solved analytically anymore, but the solution can be found with little effort using numerical problem solvers. This enables us to predict with some confidence the temperatures and power requirements in the ATLAS strip system throughout Phase-II operation. The results from this prediction have been used throughout the ATLAS strip project to consistently dimension the different systems (cooling, power, services, etc.), including an appropriate margin due to the inclusion of a common set of safety factors. This method can be easily adapted to any other system by adjusting the model to the system-specific geometries and parameters.

1.1. The ATLAS strip system

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

In the barrel, the detector modules are made of square sensors ($96.85 \times 96.72 \text{ mm}^2$) with hybrids on top, which host the front-end chips (ABC130 [6] and HCC [2]) as well as circuitry to convert the supply voltage of larger than 10 V to the chip voltage of 1.5 V, controlled by the FEAST chip. The modules are glued onto both sides of a composite sandwich (local support) that contains two parallel thin-wall titanium cooling pipes embedded in carbon foam (Allcomp K9) between two facesheets of ultra-high-modulus carbon fibre (3 layers of K13C2U/EX1515) co-cured together with a Kapton/copper low-mass tape. A model of this geometry is shown in Fig. 1. During final operation, cooling will be achieved by evaporating CO₂ in the cooling pipes with a final target temperature no higher than -35°C anywhere along the stave.

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

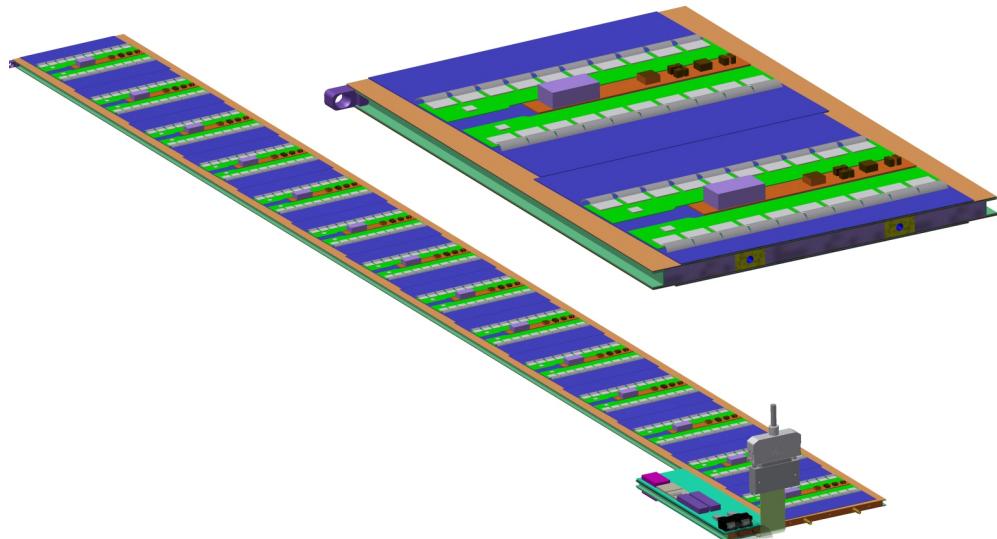


Figure 1: Strip barrel local support geometry. On the left, a complete stave is shown (EOS card in the foreground). The right picture shows a cross-section of the stave with the two cooling pipes visible inside the core.

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

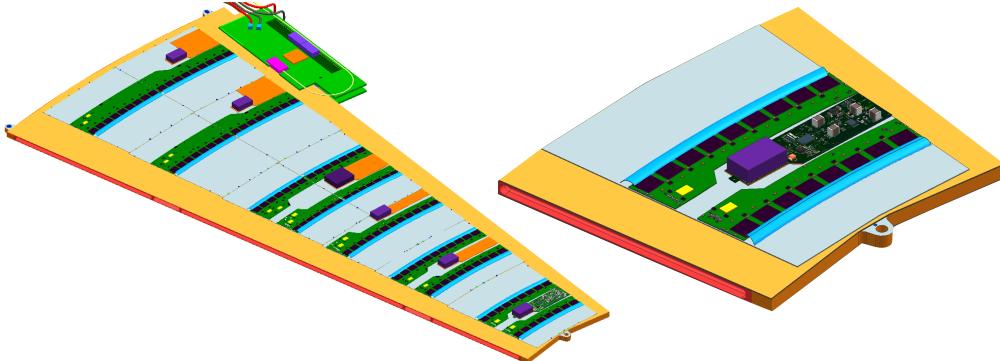


Figure 2: The geometry of the endcap strip petal, featuring 6 distinct module designs. A close-up of the R0 module is shown on the right.

70 1.2. Radiation environment

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

76 Predictions for both of these quantities have been generated for each point in the ITk using the FLUKA
 77 particle transport code and the PYTHIA8 event generator (Fig. 3) [7]. In the barrel system, both of these
 78 distributions display a strong dependence on r but a weak z -dependence. Accordingly, we make the simplifying
 79 assumption that modules within the same barrel layer have identical fluence and TID, and model four
 80 different radiation profiles (one for each barrel layer). In the endcaps, the radiation levels vary significantly
 81 over the length of the petals and from disk to disk; therefore, we model each disk and ring position separately
 82 (36 in total).

83 2. The electrical model

84 The electrical model consists of low-voltage (LV) and high-voltage (HV) circuits, depicted in Fig. 4. The
 85 LV current (supplied at 11 V) is used to power the hybrid controller chips (HCCs), ATLAS Binary Chips
 86 (ABCs) and Autonomous Monitoring and Control chip (AMAC) located on PCBs that are glued directly onto
 87 the surface of the sensor. These chips require between 1.5 and 3.3 V, which are provided by the temperature-
 88 dependent FEAST DC-DC converter (labeled bPOL12V in Fig. 4) and an LDO (low-dropout) regulator
 89 (labeled bPOL12). The number of chips and converters on each module vary according to the design of each
 90 different module type (barrel short-strip and long-strip modules, and six different endcap module designs).
 91 A barrel or endcap module contains 10–28 ABC chips, 1–4 HCCs, and 1–2 of each of the other components
 92 (linPOL12V/bPOL12V/AMAC).

93 The LV current is also delivered to the EOS card to power various data transfer components: the Gigabit
 94 Laser Driver (GBLD), low power GigaBit Transceiver (LpGBT) and Gigabit Trans-impedance Amplifier
 95 (GBTIA) chips. A FEAST identical to the one used on the module is used to step the voltage down from

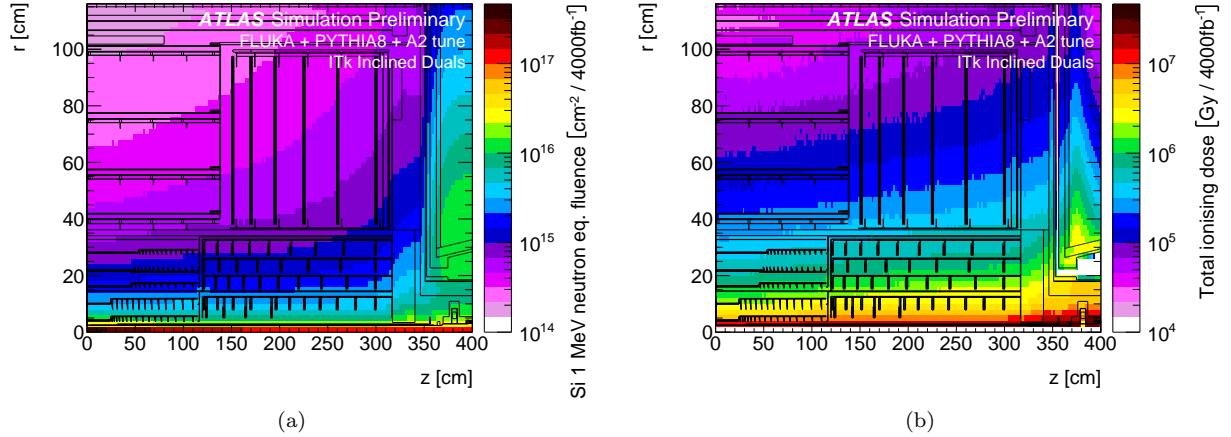


Figure 3: The ATLAS ITk radiation environment. (a) 1 MeV neutron equivalent fluence and (b) total ionizing dose. Both plots are for an integrated luminosity of 4000 fb^{-1} [7].

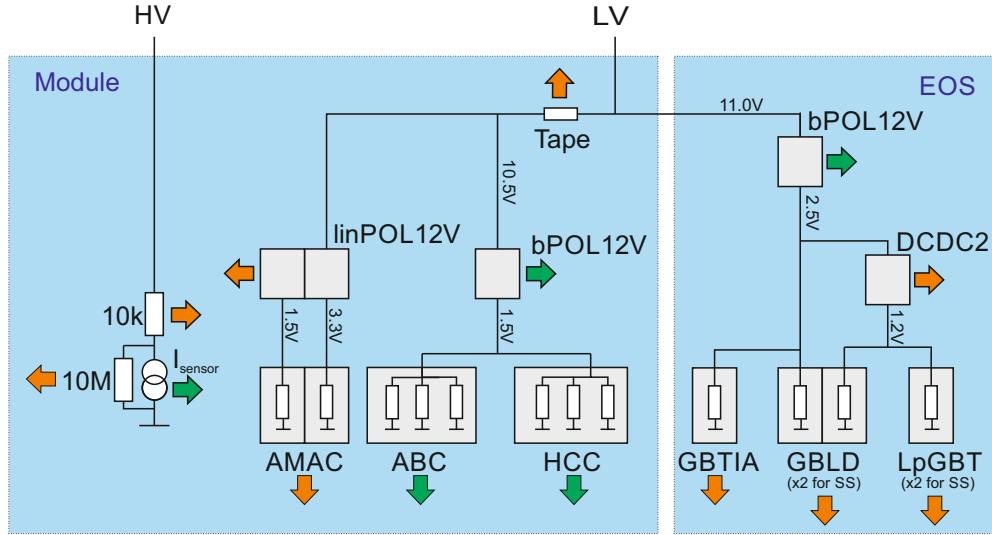


Figure 4: The electrical model of the ITk Strip barrel and endcap modules. Green arrows represent temperature-dependent heat sources, while orange arrows are temperature-independent. Grey squares are chips.

96 11 V to 2.5 V, and an additional LDO regulator brings the voltage down further for some components. The
97 short-strip barrel staves contain two GBLD and LpGBT chips.

98 The bus tape, which carries both LV and HV currents, has a small ohmic resistance, which impacts the
99 module in two ways. First, the tape itself will generate some heat according to the amount of current passing
100 through it; this source of heat is accounted for in the model, however the contribution to the total module
101 power is negligible. Second, due to the voltage loss along the traces, there is a slight reduction in voltage
102 supplied to successive modules along the substructure. The treatment of this effect is slightly different in the
103 barrel and endcap models: in the barrel, the voltage delivered to every module is averaged to 10.5 V; in the
104 endcap, the ΔV is estimated based on the calculated expected power loss along the tape for each module
105 and varies between 10.8 and 11 V. In both the barrel and endcap systems, the impact of using a different
106 treatment is small.

107 Finally, the HV current provides the bias voltage on the silicon sensors. An HV multiplexer switch
108 (HVMUX) can be used to disconnect the sensor from the bias line (it requires a $10 \text{ M}\Omega$ resistor parallel to

109 the sensor in order to function). Two HV filters with an effective resistance of $10\text{ k}\Omega$ are situated in series
 110 with the sensor. The nominal operating voltage of the sensor is expected to be 500V, but the system is
 111 designed to handle a bias voltage of up to 700V.

112 3. The thermal model

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

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

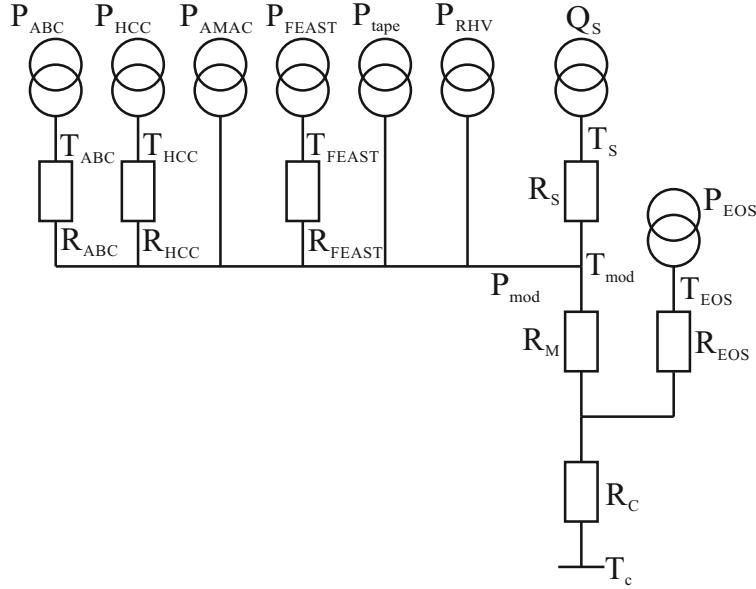


Figure 5: Thermal network model.

131 This is a more complex thermal network than the one studied in Ref. [5], for which an analytical solution
 132 for the determination of thermal stability is given. In particular, because of the non-linear temperature
 133 dependence of some of the heat sources, it is not possible in the present case to solve the set of equations
 134 describing the model analytically. However, the set of equations is still sufficiently small to solve numerically
 135 using functional programming languages such as Mathematica (used in the barrel model) or Python (used
 136 in the endcap system).

¹³⁷ **4. Obtaining thermal impedances using FEA**

¹³⁸ The cooling path between the sources dissipating electrical power and the cooling fluid is 3-dimensional
¹³⁹ and includes components with orthotropic thermal conductivity. Hence the prediction of temperature at any
¹⁴⁰ node of the model requires a 3D thermal FEA [8, 9]. However, the thermal conductivities of the components
¹⁴¹ along the path are approximately constant, so that the temperature rise ΔT_i above the coolant temperature
¹⁴² of any node i ($i = \text{ABC, HCC, AMAC, FEAST, tape, RHV, or sensor}$) in the thermal network model is
¹⁴³ adequately described by a linear sum of contributions from individual sources, i.e:

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

¹⁴⁴ where the index j runs over all powered nodes. (We have momentarily ignored the contribution from the
¹⁴⁵ EOS card.)

¹⁴⁶ In order to extract the thermal impedances for the thermal network model, the finite element model is run
¹⁴⁷ multiple times, with each heat source (or group of similar sources) switched on in turn with a representative
¹⁴⁸ amount of heat. In each of these cases, the temperature is calculated for all nodes in the thermal network
¹⁴⁹ model (Fig. 5). The temperature of a node is here taken as the average of the temperatures for all the
¹⁵⁰ grid points in the FEA model within the volume of the object corresponding to the node². The thermal
¹⁵¹ impedances are then obtained from a fit of Eq. 2 using the temperature data for all nodes for all cases of
¹⁵² heat injection.

¹⁵³ Because of the nature of the network, the fitted value for the common impedance $R_{CM} = R_C + R_M$ is
¹⁵⁴ determined by the observed temperature rises of components where no heat is injected. The linearity of
¹⁵⁵ this relationship is illustrated in Fig. 6. The value of each component-specific impedance is determined from
¹⁵⁶ the temperature rise observed when heat is injected into that component. The linear approximation of the
¹⁵⁷ model reasonably describes the FEA simulation, and the level of disagreement, discussed below, is taken as
¹⁵⁸ an uncertainty that is assessed as a safety factor as described in Section 7.2.

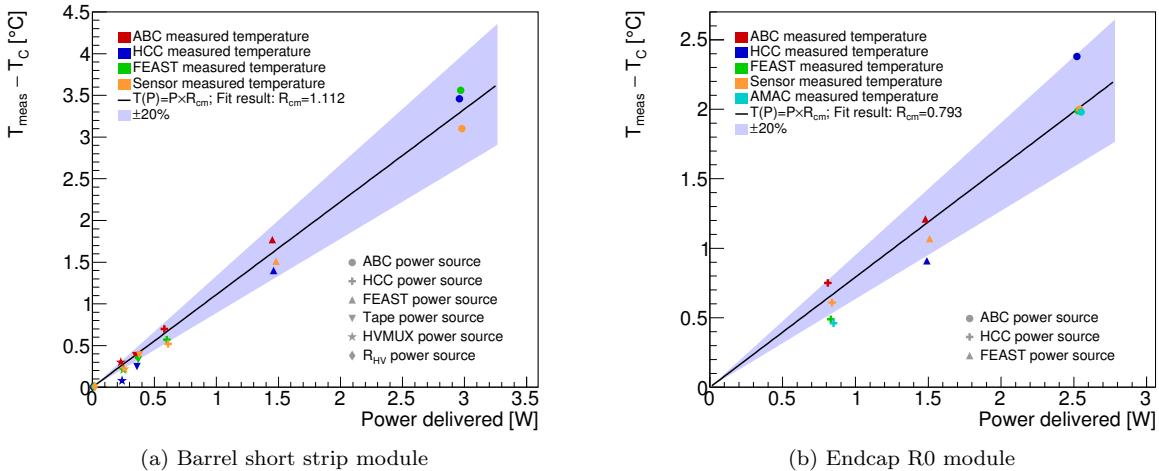


Figure 6: The relationship between the temperature rise observed in the FEA for a specific component and the heat injected in another component. The slope of the fitted line is the estimate for R_{CM} . (a) The fit for a short-strip barrel module adjacent to the EOS card. (b) The fit for the endcap R0 module. For each data point marker, the source of power is indicated by the shape, and the measured component is indicated by the color. The blue band represents a $\pm 20\%$ error band on the fit for R_{CM} .

²This is particularly interesting in the case of the sensor, which fills a large volume, with a potentially large range of temperatures. In Ref. [5] the analytic model parameters were extracted from the maximum sensor temperature predicted by FEA, whereas our subsequent studies have shown that the thermal stability limit is predicted more accurately if the average sensor temperature is used.

159 For a barrel module, the agreement of the network temperatures using the thermal impedances from the
160 fit with the data from FEA is better than 0.5°C for all nodes. This procedure is performed for both an EOS
161 module and a normal module. The thermal impedance from the sensor to the sink (R_{CM}) is consistently
162 between 1.1 and $1.4^{\circ}\text{C}/\text{W}$, but higher values (between 10 and $20^{\circ}\text{C}/\text{W}$) are found for other impedances in the
163 network (R_{HCC} and R_{FEAST}), mostly because these are for components with a small footprint constituting a
164 bottleneck for the heat flow. A 10% uncertainty is assigned to the thermal impedances in the barrel modules
165 for the purposes of assessing safety factors.

166 For the endcap modules, the procedure to determine the thermal impedances is performed for each of the
167 6 module types. R_{CM} ranges from 0.6 to $1.4^{\circ}\text{C}/\text{W}$, with other nodes between 5 and $20^{\circ}\text{C}/\text{W}$. Because the
168 location of powered components is more irregular on an endcap module, the difference between the predicted
169 temperatures of the linear network and the FEA can reach up to 1.2°C for key temperature-dependent
170 nodes. Therefore, a 20% uncertainty is assigned to the thermal impedances in the endcap modules.

171 There are two recognised departures from linearity of the thermal path: the rise in thermal conductivity
172 of the silicon sensor with decreasing temperature, and the rise in heat transfer coefficient (HTC) of the
173 evaporating CO_2 coolant with increasing thermal flux. The FEA models are run using mean values for these
174 quantities appropriate to the operating conditions, and the thermo-electrical model results are insensitive to
175 the variations expected in practice. However, if this level of realism is required and if reliable parametrizations
176 for these dependencies can be obtained, then the inclusion of such variations in the model is possible.

177 5. Other model inputs

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

184 5.1. DC-DC converter

185 The DC-DC converter, controlled by the FEAST chip, supplies a low-voltage (1.5 V) current to the
186 ABC130 and HCC front-end chips on the module. The efficiency of the FEAST depends on its temperature
187 as well as the output (load) current load delivered to the front-end chips. To correctly model the FEAST
188 efficiency, experimental measurements have been performed to characterize the dependence and fitted with
189 a functional form.

190 For the measurement, the FEAST power board was glued to an aluminum cold plate, cooled with CO_2 , and
191 powered with the nominal working input and output voltages (11 V input, 1.5 V output). The temperature
192 of the FEAST was measured with an NTC (negative temperature coefficient) thermistor and a PTAT (point
193 to absolute temperature) sensor residing on the FEAST for a range of load currents up to the maximum
194 design current of 4A.

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

200 5.2. Digital current increase of chips using 130 nm CMOS technology

201 The ABC and HCC chips, designed using IBM 130 nm CMOS 8RF technology, are known to suffer from
202 an increase in digital current when subjected to a high-radiation environment [2]. This phenomenon, known
203 as the “TID bump,” is well-studied [10, 11] and has a characteristic shape whereby the effect reaches a
204 maximum as a function of the accumulated dose and then gradually diminishes (see Fig. 8).

205 In an effort to characterize the nature of the TID bump in the ABC and HCC chips empirically, many
206 irradiation campaigns have been conducted using a variety of radiation sources, testing the effect at different
207 temperatures and dose rates. The data collected from these studies was used to develop a model of the
208 TID bump that estimates the digital current increase given the total ionizing dose, the dose rate, and the

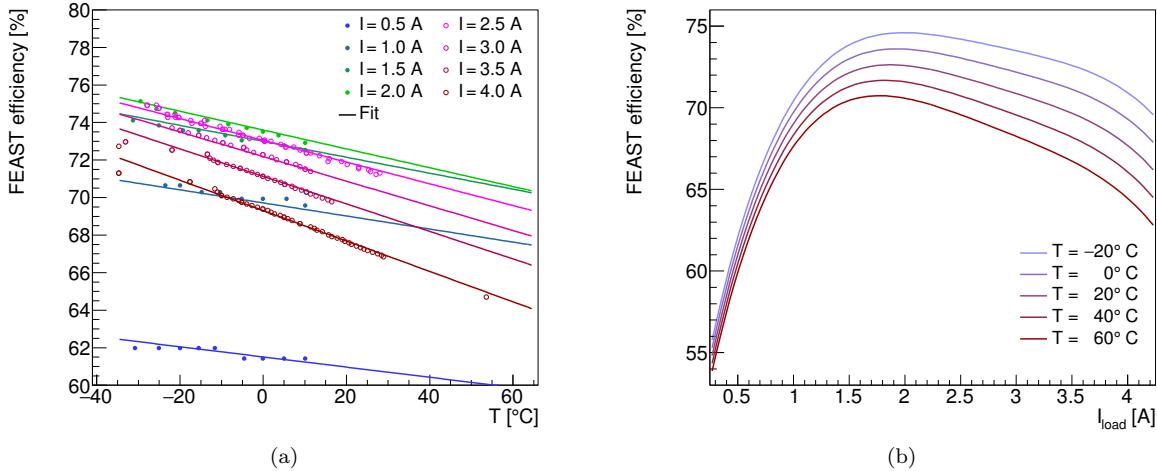


Figure 7: The FEAST efficiency model based on experimental data. (a) The experimental data points characterizing the FEAST efficiency are plotted as dots and color coded for load current. The data is compared to the analytic fit, evaluated in curves of equal current. (b) The same analytic fit, presented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

209 operating temperature of the chip. This parametrization, which is depicted in Fig 8, is used as an input
 210 to the thermo-electrical model in order to correctly model the ABC and HCC currents. The TID bump is
 211 assumed to fully apply to the HCC digital current, and apply to 69% of the ABC digital current (according
 212 to our understanding of its digital circuitry).

213 The TID bump displays certain key features, which are reflected in the parametrization: first, the effect is
 214 larger at colder temperatures and higher dose rates. This means it can be mitigated by operating the chips at
 215 higher temperature (note that the dose rate is determined by the LHC operational conditions). Second, the
 216 figure illustrates how chips receiving different dose rates will reach their maximum digital current increase
 217 at different times. This feature is particularly important when modelling the total power consumed by the
 218 barrel and endcap systems. In both systems, the dose rate varies significantly depending on the position of
 219 the module in the detector. The effect means that the maximum system power will be smaller than the sum
 220 of the maximum power of each module, as each chip reaches its maximum at a different point in time.

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

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

234 5.3. Radiation-dependent leakage current

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

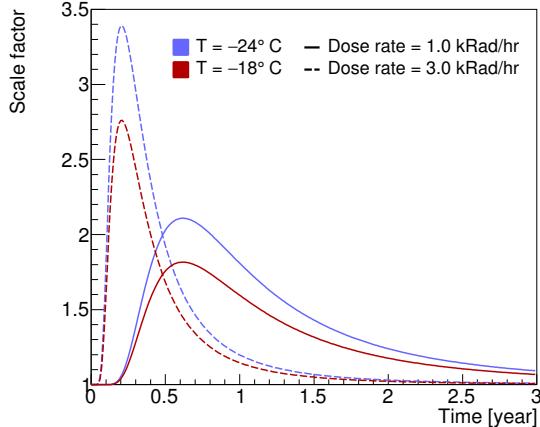


Figure 8: Parametrization of the impact of the total ionizing dose on the magnitude of the front-end chip digital current (the TID bump), presented as a function of time. The current is multiplied by a scale factor that is modeled as a function of total ionizing dose, dose rate, and temperature, based on experimental data.

239 6. Running the model

240 The thermo-electrical model constructs a profile of the sensor module operation conditions over the
 241 lifetime of the detector in the following manner. First, the total module power (including all components,
 242 but excluding the sensor leakage power) and the sensor temperature assuming no leakage current (T_0) are
 243 calculated using a reasonable set of initial component temperatures. The initial value for the module power is
 244 used to solve for the sensor power and temperature accounting for leakage current, using the thermal balance
 245 equation and the relationship from Eq. 1. Using this calculated sensor leakage current and temperature, the
 246 power and temperature of the module components are updated given the initial (year 0, month 0) startup
 247 parameters.

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

255 In the barrel subsystem, the above procedure is performed four separate times to represent the radiation
 256 conditions of the four barrel layers located at different radii from the beam axis³ for both a normal and an
 257 EOS-type module. Thus, eight modules are simulated in total for the barrel (4 layers \times normal/EOS), and
 258 they are combined in their proper proportion to simulate the entire barrel system.

259 In the endcap subsystem, the total ionizing dose and dose rates vary significantly depending on the
 260 position of the module; furthermore, the design of each module on a petal differs significantly. Therefore, all
 261 36 module types (6 rings \times 6 disks) are simulated independently, and combined to represent the full endcap.

262 We have implemented this algorithm in Mathematica (barrel) and Python (endcaps). In both cases, the
 263 calculation for a set of operating conditions over the full lifetime of the LHC takes between 5 and 10 minutes
 264 on a standard PC, thus enabling a quick turn-around for systematic studies of the parameter space.

265 7. Outputs of the thermo-electrical model

266 The thermo-electrical model provides a wide range of predictions for the operation of the strip system. A
 267 detailed discussion of all results is beyond the scope of this article; instead, we present here a subset of the

³The correct module type, short-strip in the inner two layers and long-strip for the outer two layers, is used for each layer.

268 results to demonstrate the capabilities and use of the thermo-electrical model for the design of the detector
 269 system.

270 7.1. Operational scenarios

271 To study the different aspects of our predictions for the operation of the ITk strip system throughout
 272 its lifetime, we performed the calculation of the system parameters over the expected 14 years of operation
 273 in monthly steps as outlined in Section 6. Time-dependent operational inputs to the calculation were taken
 274 from the expected performance of the HL-LHC (Fig. 9a). For the cooling temperature, which can be adjusted
 275 during data taking using detector control systems, we studied flat cooling profiles with temperatures as low
 276 as -35°C , the lowest evaporation temperature achievable with the ITk evaporative CO_2 cooling system, as
 277 well as a ‘ramp’ scenario in which the cooling temperature starts at 0°C and is gradually lowered down to
 278 -35°C over the course of 10 years (Fig. 9b).

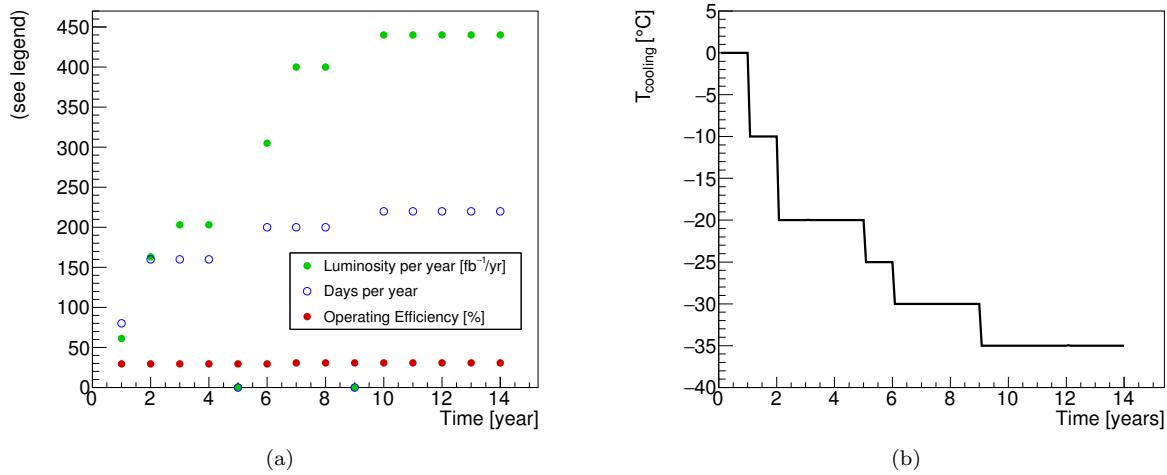


Figure 9: (a) Expected HL-LHC performance and (b) ‘cooling ramp’ scenario for the coolant temperature. Year-long shutdowns of the LHC are anticipated in years 5 and 9.

279 7.2. Safety factors

280 To ensure the robustness of the system design against uncertainties in the assumptions used in the model,
 281 we also evaluate the model using a set of input parameters with some key inputs degraded. The set of safety
 282 factors used is given in Table 1. Each safety factor has been estimated individually based on experience, the
 283 complexity of the system aspect described by the parameter, and from available data or the absence of such
 284 data. Note that the model can be evaluated with all the safety factors listed in Table 1 used together, a
 285 situation that is unlikely to occur in the real system, to provide a worst-case estimate for the performance of
 286 the ITk strip system. The individual effects of the different safety factors are demonstrated in Fig. 10.

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

292 7.3. Module properties

293 Several module properties predicted by the thermo-electrical model are shown in Figs. 11 and 12 for
 294 the barrel system. The different radiation-dependent effects occur on different timescales. The maximum
 295 in the digital chip power due to the TID effect occurs relatively early (in year 1 to 4), although the bump
 296 has a long tail, particularly in the outer layers of the barrel. The sensor leakage power, on the other hand,
 297 grows towards the end of the lifetime of the ITk. If the leakage current continued to increase in the case of

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

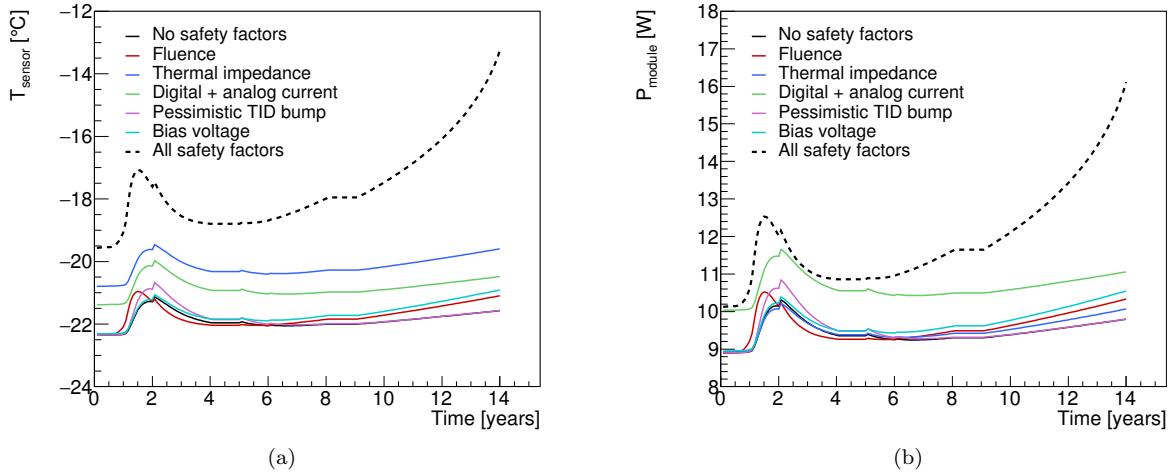


Figure 10: Comparing the impact of different safety factors on (a) the sensor temperature and (b) the module power for the endcap R3-type module, using a flat cooling scenario (-30°C). The dotted line depicts the effect of all safety factors applied at once.

298 further irradiation, or if the cooling temperature were raised, this growth would ultimately lead to thermal
 299 runaways. Due to the radial dependence of the radiation environment, the radiation-induced effects are most
 300 pronounced in the innermost barrel layers.

301 7.4. System properties

302 One of the key concerns for the design of the strip system is thermal stability of the system. If the
 303 cooling temperature is too high to limit the leakage power from the radiation-damaged sensors to a level
 304 where the heat can still be removed, the system is unstable (it goes into ‘thermal runaway’). To find the
 305 cooling temperature T_C at which this condition is reached, we make repeated simulations of the ITk strip
 306 system using the thermo-electrical model, with each simulation representing the full 14-year operation of
 307 the ITk at a fixed T_C . Between simulations, T_C is increased in steps of 5°C until the model finds thermal
 308 runaway. In the numeric evaluation of the thermo-electrical model this manifests itself in the absence of a
 309 solution to the system of equations. In the endcap strip system, this occurs at a cooling temperature of
 310 -15°C under nominal conditions (i.e. with no safety factors applied); in this scenario, thermal runaway
 311 would be reached in the 12th year of operation. With all safety factors applied, thermal runaway would occur
 312 at a cooling temperature of -25°C (in year 10). In the barrel system, where the radiation environment is
 313 slightly less intense, the conditions for thermal runaway occur at the same cooling temperatures, but a few
 314 years later than in the endcaps: in the final year of operation and a cooling temperature of -15°C under
 315 nominal conditions, and at -25°C (in year 13) with safety factors applied. As the design cooling temperature

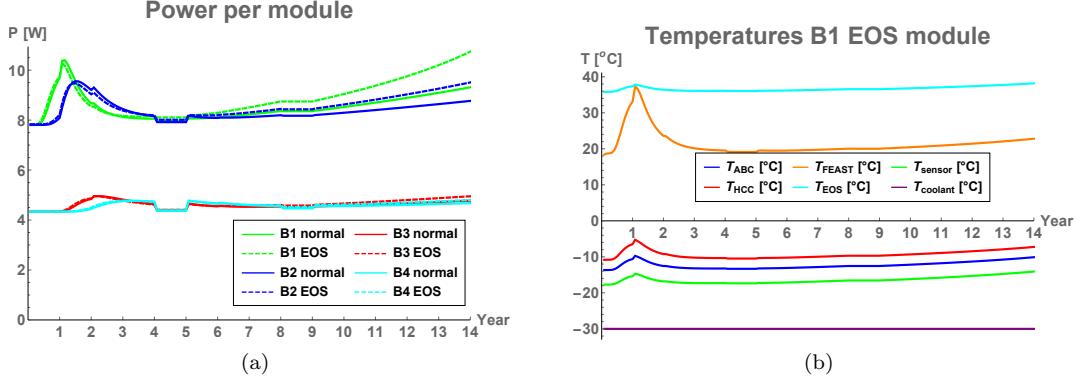


Figure 11: Examples of barrel module performance predictions for a flat cooling scenario (-30°C) including safety factors. (a) Power per module. (b) Temperatures for different nodes of an end-of-stave barrel module in the innermost barrel. The discontinuities in year 5 and 9 are due to anticipated year-long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

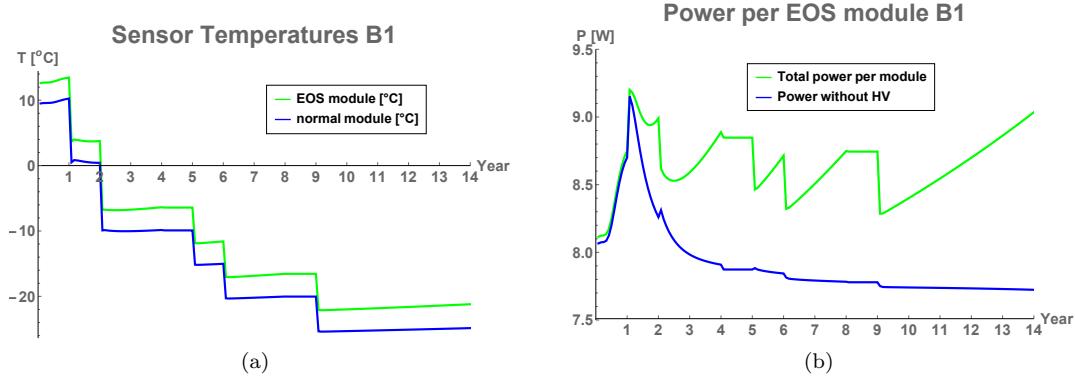


Figure 12: Examples of barrel module performance predictions for the ramp cooling scenario including safety factors. (a) Sensor temperature in the innermost barrel modules. (b) Power in an end-of-stave barrel module in the innermost layer.

316 of the ITk cooling system is -35°C , we have confidence that the ITk strip system has a sufficient margin for
317 thermal stability.

318 Beyond the issue of stability, the thermo-electrical model delivers predictions for the development of
319 current and power requirements for the overall system. Some of the predictions are shown in Fig. 13. Again,
320 the different timescales of the various radiation-induced effects are visible; ignoring this time dependence
321 could lead to over-specification of some system aspects like the total cooling power.

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

326 Because of the different timescales for the peak power due to the TID effect and the radiation-induced
327 sensor leakage, there is room to optimize the cooling temperature profile to minimize the total power in the
328 strip system while avoiding thermal runaway. The thermo-electrical model is a powerful tool to plan such an
329 optimized cooling profile. In fact, the cooling ‘ramp’ scenario introduced in Section 7.1 is the result of such
330 an optimization. In this scenario, depicted in Fig. 14, the cooling temperature begins at a relatively high
331 value (0° C) to minimize the impact of the TID bump in the first two years of operation, thus avoiding a
332 peak in the module power (see Fig. 14a). In subsequent years, T_C is steadily decreased to maintain a sensor
333 current at or below about 1 mA, as illustrated in Fig. 14b, in the interest of both minimizing the module

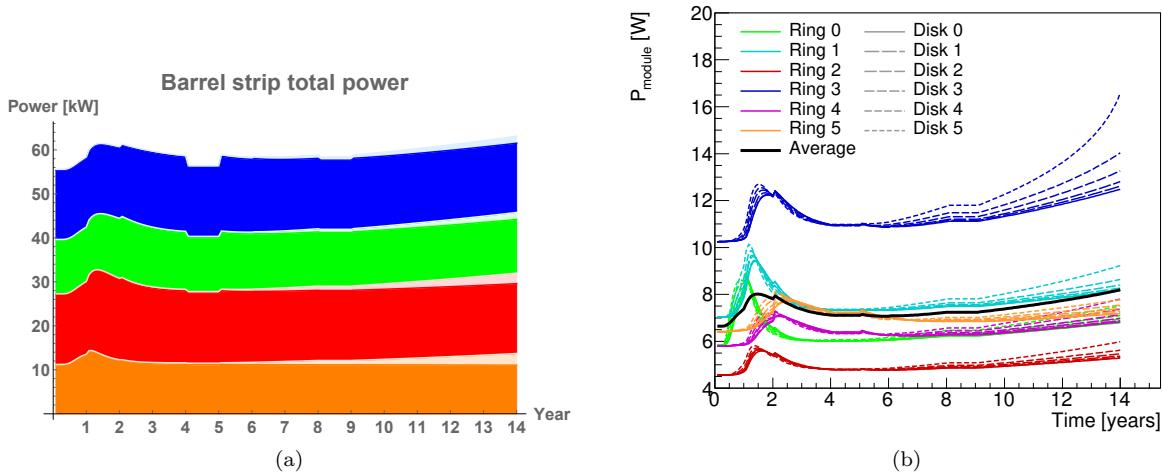


Figure 13: Examples of system performance predictions. (a) Barrel total power requirements. The plot shows the stacked power requirements for the four barrel layers (orange: innermost barrel, blue: outermost barrel). Full colour indicates power from the front-end electronics, greyed parts are contributions 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 each of the 36 simulated endcap modules, labeled according to their ring type and disk position. The solid black line indicates the average module power. Both predictions use a scenario with flat -30°C cooling and including all safety factors.

334 power and avoiding thermal runaway.

335 8. Model performance verification

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

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

347 Because the variable model inputs are kept constant for this study, we can reduce the complex thermal
 348 network to its Thévenin equivalent, which is identical to the network studied in Ref. [5], and use the analytical
 349 expressions given there. The reduced network is described by the base temperature T_0 , defined as the sum
 350 of the coolant temperature and the temperature rise due to the front-end electronics alone, and the total
 351 thermal impedance R_t from the sensor to the coolant. Using the nominal resistances and representative power
 352 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°C
 353 and 1.147 K/W obtained directly from the FEA. The comparison of the predicted sensor temperatures for
 354 both cases is shown in Fig. 15. Despite a large temperature variation of about 10°C across the sensor,
 355 the network model runaway prediction agrees well with the FEA⁴. This gives us confidence that the use

⁴The critical temperature here is -12.4°C , which is higher than the numbers given in Section 7.4, because the study here ignores temperature effects such as the FEAST efficiency, which can only be modelled in the network model.

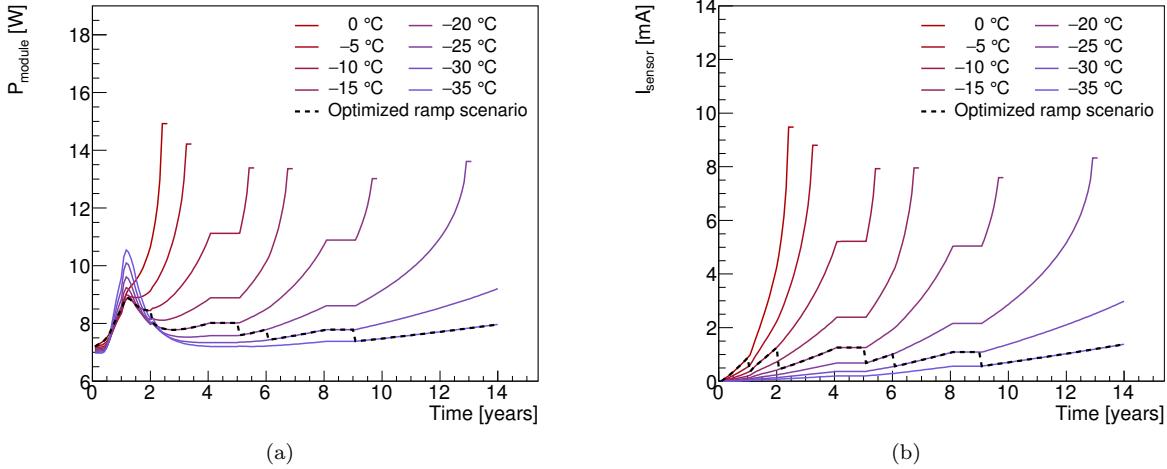


Figure 14: (a) Total power and (b) sensor leakage current of the endcap R1-type module for eight different flat cooling profiles, ranging from 0°C to -35°C , as well as the cooling ramp scenario specified in Fig. 9b (dashed curve). The curves that are discontinued before year 14 correspond to scenarios 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.

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

358 9. Conclusions

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

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

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

379 10. Acknowledgements

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

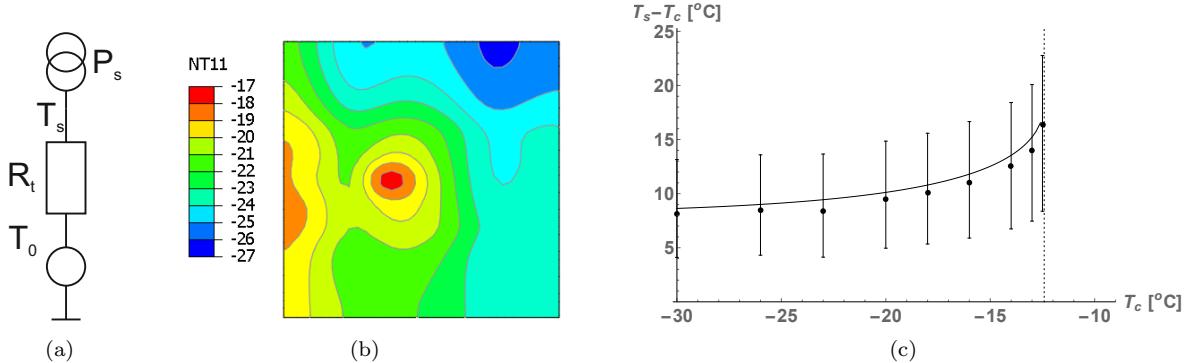


Figure 15: (a) Thévenin equivalent of the thermal network. (b) Result of sensor surface temperature calculations using FEA, assuming zero sensor power. The EOS card is to the left of the module, and the cooling pipes run from top to bottom about a quarter of the module width from each edge. (c) Difference of average sensor and coolant temperature, comparing FEA (dots) and the network model prediction (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).

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387 References

- 388 [1] A. Chilingarov, Temperature dependence of the current generated in Si bulk, Journal of Instrumentation
 389 8 (10) (2013) P10003–P10003. doi:10.1088/1748-0221/8/10/p10003.
 390 URL <https://doi.org/10.1088%2F1748-0221%2F8%2F10%2Fp10003>
- 391 [2] A. Collaboration, Technical Design Report for the ATLAS Inner Tracker Strip Detector, Tech. Rep.
 392 CERN-LHCC-2017-005, ATLAS-TDR-025, CERN, Geneva (Apr 2017).
 393 URL <https://cds.cern.ch/record/2257755>
- 394 [3] Radiation induced effects in the ATLAS Insertable B-Layer readout chip, Tech. Rep. ATL-INDET-PUB-
 395 2017-001, CERN, Geneva (Nov 2017).
 396 URL <https://cds.cern.ch/record/2291800>
- 397 [4] A. Affolder, B. Allongue, G. Blanchot, F. Faccio, C. Fuentes, A. Greenall, S. Michelis, DC-DC converters
 398 with reduced mass for trackers at the HL-LHC, Journal of Instrumentation 6 (11) (2011) C11035.
 399 URL <http://stacks.iop.org/1748-0221/6/i=11/a=C11035>
- 400 [5] G. Beck, G. Viehhauser, Analytic model of thermal runaway in silicon detectors, Nucl. Instrum. Meth.
 401 A618 (2010) 131–138. doi:10.1016/j.nima.2010.02.264.
- 402 [6] N. Lehmann, Tracking with self-seeded Trigger for High Luminosity LHC, Master's thesis, Section of
 403 Electrical and Electronical Engineering, École Polytechnique Fédérale de Lausanne, Lausanne Switzerland
 404 (2014).
 405 URL https://documents.epfl.ch/users/n.nl/nlehmann/www/SelfSeededTrigger_MasterThesis/SelfSeededTrigger_NiklausLehmann_Thesis.pdf
- 406 [7] Atlas experiment - radiation simulation public results [cited 2018-11-17].
 407 URL https://twiki.cern.ch/twiki/bin/view/AtlasPublic/RadiationSimulationPublicResults#FLUKA_Simulations

- 410 [8] M. Smith, ABAQUS/Standard User's Manual, Version 6.9, Simulia, 2009.
- 411 [9] ANSYS, Inc., Ansys academic research mechanical, release 18.2.
- 412 URL <http://www.ansys.com/>
- 413 [10] F. Faccio, G. Cervelli, Radiation-induced edge effects in deep submicron cmos transistors, IEEE Transactions on Nuclear Science 52 (6) (2005) 2413–2420. doi:[10.1109/TNS.2005.860698](https://doi.org/10.1109/TNS.2005.860698).
- 414
- 415 [11] F. Faccio, H. J. Barnaby, X. J. Chen, D. M. Fleetwood, L. Gonella, M. McLain, R. D. Schrimpf, Total ionizing dose effects in shallow trench isolation oxides, Microelectronics Reliability 48 (7) (2008) 1000 – 1007, 2007 Reliability of Compound Semiconductors (ROCS) Workshop. doi:<https://doi.org/10.1016/j.microrel.2008.04.004>.
- 416
- 417 URL <http://www.sciencedirect.com/science/article/pii/S0026271408000826>
- 418
- 419