Modelling and simulating the dynamics of in-stent restenosis in porcine coronary arteries

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2. Dynamics of in-stent restenosis: impact of stent deployment and design

Coronary heart disease (CHD) is the most common cause of death in Europe, being responsible for approximately 1.9 million deaths per year [1]. Atherosclerosis, the most common form of CHD, is associated with the build-up of atheromatous plaque within the wall of the coronary arteries. Presence of such plaque may lead to a significant decrease in luminal cross-sectional area (a stenosis) and reduce the blood supply to the heart. The current intervention of choice for many symptomatic patients is stent-assisted balloon angioplasty. In this procedure, a tubular mesh-like structure (stent) is deployed within the vessel at the site of the stenosis. Once deployed, the stent behaves as a mechanical scaffold, compressing the plaque and eliminating the possibility of vessel collapse. Whilst this generally results in a successful outcome, in-stent restenosis (ISR), an excessive re-growth of tissue within the stented segment of the artery associated with the injury caused by the stent deployment, is still a negative aspect of the procedure [9,10,11].

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The process of restenosis involves a cascade of complex biological and physical interactions that are activated in response to stent-induced arterial wall damage. Denudation of endothelial cells and the rupture and/or stretch of the elastic laminae, media and adventitia after stent deployment act as stimuli to initiate the development of ISR [60]. The process can be categorised as a sequence of events starting with an initial instantaneous thrombosis formation in response to arterial injury and progressing to an inflammatory stage, granulation tissue development, smooth muscle cell (SMC) proliferation, extracellular matrix deposition and ultimately remodelling of the neointima and vascular wall [27,61]. A detailed description of the underlying biological processes has been presented by Evans et al. [39].

It is widely accepted that the development of restenosis is regulated by a number of key factors including; the severity of injury, inter-strut spacing, strut thickness, stent length, stented segment compliance [27,62,63]. All of these factors have profound effects on flow hemodynamics and wall shear stress (WSS) distribution across the stented segment [64,65]. A decreased restenosis rate has been observed with the use of drug eluting stents (DES), which are coated with antimitogenic or anti-proliferative drugs to inhibit SMC proliferation and neointimal growth [66]. Despite a growing trend for use of DES as the stent of choice in comparison to bare metal stents (BMS), late stent thrombosis (LST) and lack of re-endothelialisation are the major drawbacks associated with the use of DES [62,66,67].

The degree of injury caused by the implantation of a stent has been observed as an independent determinant to estimate the amount of restenosis in an experimental study carried out in pigs [68]. The extent of vascular injury is estimated from the penetration depth of the struts into the tissue; a deeper deployment depth implies a larger injury. In the current study, we have used the Gunn Injury Score system which defines the degree of injury according to the angle of the internal elastic lamina (IEL) at the point of strut impact, the rupture of the IEL or, in the extreme, rupture of external elastic lamina (EEL). Gunn applied this to categorise injury arising from deployment of a stent in the porcine implant model [69].

In previous work, we modelled ISR as a complex multiscale system [39], and realised a coupled multiscale simulation of a simplified version of this model in two dimensions using porcine coronary arteries [40]. In the present work, the overall dynamical response of the porcine coronary artery ISR model is studied in detail, as well as the influence of different stent geometries on the
simulated ISR. The rationale for choosing to simulating ISR in a porcine artery is based on the fact that an extensive literature from clinical and experimental studies performed on pigs is available to us, whereas there is no similar time-series data available for human implants. Moreover, the pig exhibits similarities with the human case in terms of coronary anatomic structure and vessel sizes and, most importantly, in terms of the response to arterial injury caused by the penetration of a stent into the vessel wall [58]. The primary difference between the pig and human is the rate of ISR development. In pigs, the normal time span of restenotic lesion development is about 60 days with a peak SMC proliferation at around 20 days, whereas in the human, ISR development extends to six months with the proliferation rate peaking at around 2 months [58,70]. There is some literature showing the influence of strut thickness on the progression of angiographic and clinical ISR in humans, but such data is relatively sparse [63,71,72,73]. The main conclusion from clinical trials is the impact of strut thickness. This is an independent predictor of ISR, with thin struts resulting in a lower incidence of restenosis as well as lower late luminal loss when compared to the results observed for thick strut [63,65,73].

In the current study, we report results produced by our two dimensional multiscale ISR model. The effect of strut thickness on the development of restenosis was investigated using two bare-metal stents of different strut thickness. The deployment depth was varied, and the number of neointimal cells as a function of time was measured. The resulting growth curves were fitted to a logistic function, and the simulated ISR characterised by just two parameters. The percentage of proliferating vascular SMCs was also measured as function of time. Where possible, our results were compared against the available porcine model in vivo dataset. The paper concludes with a discussion on the validity of the current model, and suggestions for improvements. The implications of this work for further understanding of ISR are also discussed.

2.1 Computational Model
ISR is a multi-science multi-scale phenomenon linking a series of processes which can be described in terms of biology, physics and chemistry. Evans et al. [39] formulated an extended multiscale model for ISR, based on the Complex Automata paradigm (CxA) [40,41, Hoek43], where a multi-scale system is modelled as a collection of single scale models coupled through the scales. Caiazzo et al [47] simplified the original model for ISR to one where single scale models for blood flow, vascular smooth muscle cells and drug diffusion communicate with each other via smart
conduits. These single scale models operate on different temporal and spatial scales. A MUltiScale Coupling Library and Environment (MUSCLE) was used to setup and launch the simulation as a CxA [44]. Here we consider the two-dimensional (2D) version of this model.

We will now briefly describe the model for ISR. For all details, we refer to chapter 1 and Caiazzo et al [47]. During the stent deployment process, stretch was induced in the vessel wall by deploying a stent, modelled as a single 2D strut, to a specific depth into the vessel wall. Next, the IEL was ruptured based on the hoop strain and longitudinal strain threshold criteria. As a result of IEL rupture, vascular SMCs were exposed to the blood flow and, according to contact inhibition criteria in the SMC biological model, only exposed cells are able to proliferate. Blood flow was modelled as a homogeneous, incompressible, Newtonian fluid using a Lattice Boltzmann Method. SMC hyperplasia in a non-axisymmetric geometry was simulated via an agent-based model where each SMC is modelled as an individual agent. A biological solver controlled the progression of SMC proliferation, which was regulated through a cell cycle based on a set of biological rules. The biological rules were based on contact inhibition (presence of neighbouring cells), WSS thresholds and, in case of a drug eluting stent, drug concentration thresholds. A physical solver, based on agent-agent interaction potentials, was used to simulate the structural dynamics of the SMCs. The process of drug elution and subsequent diffusion of the drug into the vessel wall was modelled by solving an anisotropic diffusion equation using a finite difference scheme. A detailed description of all single scale models as well as their mutual couplings has been published by Caiazzo et al [47].

Figure 2.1: SMC Domain. Left: Domain before deployment of thick stent struts (D = 180 µm). Right: Domain after strut deployment to a depth of 90 µm into the tissue (SMCs coloured according to the structural stress).
In the present study, we considered a vessel of segment length 1.5 mm and width 1.24 mm with a wall thickness of 120 µm as a benchmark geometry (Figure 2.1). Wall thickness was obtained from an existing archive of in vivo porcine data [69]. Models were prepared by deploying two, square cross-section, bare metal struts, into the vessel wall (Figure 2.1). We examined the influence of strut thickness and cross-sectional profile on neointimal proliferation. Where strut thickness is denoted as $D$, a thin strut has a strut thickness of $D = 90$ µm whereas the thicker strut has a thickness of $D = 180$ µm. Both struts were deployed into the vessel at depths over the range of 70 µm to 130 µm. We also considered a square drug eluting stent strut deployed to a depth of 90 µm. Additionally, we changed the shape of the bare metal strut to a rounded strut with a base length of 180 µm and height of 90 µm.

WSS is known to be a key parameter in the process of re-endothelialisation and neointimal formation, resulting in a higher neointimal growth in the regions of low WSS [74,75]. Steady state blood flow simulations, with a constant flow rate, were carried out using a fluid of constant viscosity $\mu = 4$ mPa.s, density $\rho = 1000$ Kg/m$^3$ and $Re = 120$. The rationale for maintaining a constant flow rate is based on the normal physiological reaction of the cardiovascular system to a moderate stenosis. This assumption served to increase the WSS as the lumen narrowed and fits well with the published theory that a blood vessel remodels in response to changes in fluid forces until these forces are normalised [76,77].

### 2.2 Results

A previous publication presented preliminary results of the 2D ISR model [49] obtained at a very early stage of model development. Further tuning of the model parameters such as WSS, drug concentration and contact inhibition was carried out within the COAST (www.complex-automata.org) consortium (data not shown). This parameter tuning allowed us to investigate the relationship between the injury index and the restenotic lesion growth by exploring the predictive power of our model.

In the present study, we performed simulations for two strut sizes ($D = 90$ µm and $D = 180$ µm) deployed at four different depths (70 µm, 90 µm, 110 µm and 130 µm). Each simulation was repeated five times to obtain the mean and standard deviation of the neointimal growth measurement.
**Figure 2.2:** Simulation results for the thin strut \((D = 90\, \mu m)\) and the thick strut \((D = 180\, \mu m)\), both deployed to a depth of 90 \(\mu m\), at 7, 14 and 28 days after stent deployment. The SMCs are shown as red circles and the square struts as blue squares surrounded by red obstacle agents. In the lumen, blood flow is visualized by streamlines, and the colours represent the shear stress, ranging from 0 Pa (dark blue) to 60 Pa (dark red).

Figure 2.2 shows a detailed visualisation of the simulated ISR response as a function of time post-deployment, for small and large struts, deployed to a depth of 90 \(\mu m\). Whilst there is a considerable SMC response in both cases, the thicker strut \((D = 180\, \mu m)\) produced a larger amount of neointima when compared with the thin strut \((D = 90\, \mu m)\). An asymmetric proliferation between the upper and lower halves of the thin strut geometry was observed (Figure 2.2). This asymmetry occurs mainly due to the random selection of the biological state of each cell after the stent deployment.

For each simulation, we measured the number of neointimal SMCs \((N)\) as a function of time \((t)\) after stent deployment. Figure 2.3a shows \(N(t)\) for a thick strut \((D = 180\, \mu m)\) deployed at a depth of 90 \(\mu m\). The dynamics of the simulated ISR response bear a close resemblance to a logistic growth curve, i.e. an initial exponential growth phase, followed by a slowing down of the growth and final settling to a maximum number of neointimal cells \(N_m\). In all cases studied, the number of neointimal cells as function of time exhibited this type of dynamics (data not shown).
In logistic growth, a population is modelled as

$$\frac{dN}{dt} = rN(1 - N/N_m) \quad \text{Eq}(2.1)$$

where \( r \) is the growth rate. If \( N_0 \) is the initial number of cells, Eq. 2.1 can be solved analytically, yielding

$$N(t) = \frac{N_mN_0e^{rt}}{N_m + N_0e^{rt}} \quad \text{Eq}(2.2)$$

We performed non-linear fits of the logistic growth dynamics (Eq. 2.2) to the measured neointimal growth, resulting in fitted values for the model parameters \( N_0, N_m \) and \( r \). The curve fitting was carried out using the Levenberg-Marquardt algorithm (Marquardt et al. 1963) to optimize parameter values in an iterative process. We kept \( N_0 = 10 \) based on our assumption that the SMCs, which are exposed to blood after the removal of endothelial cells, act as proliferative neointimal cells. In the current simulations, we observed that for the majority of cases, the number of exposed
SMCs was approximately $10 \pm 2$. We therefore chose $N_0 = 10$ as an average initial cell number to perform curve fitting. For the case reported in Figure 2.3a, this fitting procedure resulted in $N_m = 1392$ and $r = 0.33$. The best-fit curve has a very close agreement to the measured data and is also shown in Figure 2.3a. The quality of the fitted curve was calculated by means of an R-Squared ($R^2$) test, this gave a value of $R^2 = 0.997$. Similar good fits were obtained for all other experiments (data not shown). Using this procedure, we were able to better study and compare the simulated ISR as a function of parameters such as deployment depth, and strut size or shape. The results are summarised in Figures 2.3b and 2.3c that show $N_m$ and $r$ as a function of deployment depth for both strut sizes.

Figure 2.3b demonstrates that the final endpoint of restenosis is approximately constant ($\pm 10\%$) for both strut sizes. We obtained a standard deviation by averaging the results from five runs per parameter set (deployment depth, strut thickness). This was used as an input to the fitting routine. The combined error (from standard deviation and fitting error) is plotted as an error bar for each simulation set. The major difference that can be observed from Figure 2.3b is the effect of the thin strut, which results in a higher number of neointimal cells in comparison to the thick strut. The fitted growth rates including error bars from thin strut simulations were plotted and compared with the growth rate curve for neointimal formation in thick strut simulations (Figure 2.3c). The assessment of the simulation results demonstrates that a thinner strut results in slightly higher $N_{max}$ (approximately 100-150 SMCs) but lesser growth rate compared to thicker strut. This difference in the $N_{max}$ is due to the fact that the thin strut occupies less space and provides more room for SMCs to proliferate. Moreover, the data also suggests that deployment depth correlates with the growth rate; when the strut is deployed further into the vessel wall, a larger neointimal growth rate is observed. The only exception to this was for a thin strut deployed to a depth of 130 µm.

Another useful metric for use in qualitative comparison, is the normalised peak absolute growth fraction (NPAGF) as defined by Schwartz [70]. NPAGF is the product of growth fraction (Cells in Mitosis Phase / total cell number), divided by the maximum value of growth fraction across the series. This value indicates the time point at which the greatest numbers of cells are proliferating. Figure 2.4a, shows the NPAGF curve for a thick strut, deployed to a depth of 90 µm. It is clear from this curve that peak proliferation occurs at around 22 days.
The fluctuations seen in the NPAGF curve are due to the presence of discrete events (mitosis of cells). To better estimate the time where maximum proliferation is occurring, we fitted a Gaussian function (Fig. 2.4a) to obtain peak proliferation time ($T_{peak}$). Figure 2.4b shows the resulting peak proliferation time as a function of deployment depth. A downward trend of the peak proliferation time was observed with an increase in the deployment depth for both strut thicknesses, except the thin strut that was deployed at 130 µm. Analysis of Figure 2.4b also suggests that the thick strut is associated with a lower peak proliferation time at all deployment depths.

Figure 2.4: (a) Normalised peak proliferation curve from the thick strut ($D = 180 \mu m$) deployed at 90 µm. The fitted Gaussian curve highlights the time at which maximum proliferation occurred. (b) Peak proliferation time plotted against deployment depth for both strut sizes.

We also examined the effect of a drug eluted from the thick strut ($D = 180 \mu m$) deployed at a depth of 90 µm. The results were obtained by applying the curve fitting techniques explained above (see Table 2.1). A nominal drug concentration threshold of 0.4 was used. For more details about drug concentration, we refer to Caiazzo et al [47]. Data from a DES simulation indicates a delay in the overall neointimal growth process when compared to the results obtained for a similar bare metal strut. Table 2.1 also includes the results for a bare metal strut with rounded cross-section deployed to a depth of 90 µm. A reduction in the growth rate, resulting in a late peak proliferation time was observed as compared to bare metal square strut. The morphological images from rounded strut simulation are also shown in Fig 2.5 in order to allow the reader to understand the strut shape and its effect on the neointimal growth.
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**Figure 2.5:** Simulation results of a rounded strut deployed at 90 µm, showing neointimal growth after 7, 14 and 28 days after the stent deployment.

**Table 2.1:** Results from different strut type (bare metal and Drug eluting) and strut shape (square and rounded).

<table>
<thead>
<tr>
<th>Strut Type</th>
<th>( N_{\text{max}} )</th>
<th>Growth Rate (r)</th>
<th>( T_{\text{peak}} ) (Days)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Square Bare Metal</td>
<td>(1.4 ± 0.11) x 10^3</td>
<td>0.29 ± 0.024</td>
<td>20 ± 0.03</td>
</tr>
<tr>
<td>Square Drug Eluting</td>
<td>(1.3 ± 0.10) x 10^3</td>
<td>0.22 ± 0.016</td>
<td>26 ± 0.36</td>
</tr>
<tr>
<td>Rounded Bare Metal</td>
<td>(1.4 ± 0.13) x 10^3</td>
<td>0.23 ± 0.022</td>
<td>25 ± 0.33</td>
</tr>
</tbody>
</table>

**2.3 In Vivo Dataset**

We have access to a unique archive of histological sections from stented porcine coronary arteries collated from a number of experimental studies over the last decade [78,79,80,81,82,83]. The archive contains transverse sections from over 500 porcine arteries in which stents of several different designs have been deployed. For the present study, we selected transverse sections derived from more than 50 arteries harvested at 6 hours, 14, 21 and 28 days after the deployment of a bare metal BiodivYsio™ stent. An average of 10 struts were present in each section. Each of these struts was individually scored using the Gunn Injury Score [69] according to the degree of injury they imposed on the artery. These data were used to generate a database relating the degree of injury caused by stent strut deployment, to the neointimal thickness observed at the various time points after stenting. Figure 2.6 summarises these results, showing averages and standard deviations of the full set of available histological measurements. The time points in Figure 2.6 were determined by the availability of samples. The neointimal thickness measured at 6 hours does not specifically correspond to a SMC rich neointima, but represents a measurement of thrombus which forms immediately after stent deployment. Analysis of the in vivo
dataset shows that an increase in injury score correlates with increased neointimal thickness (Figure 2.6). The decrease at 28 days could be related to vessel remodelling, a process which is not considered in our current computational model.

2.4 Discussion

We report a series of simulations investigating the relationship between an injury index (in our case, deployment depth) and ISR growth. Despite the limited amount of quantitative literature available to model such complex phenomena, our results show a good qualitative correlation with the available experimental, clinical and in vivo data, as will be discussed below. We showed results from simulations deploying two different strut sizes at four deployment depths into the tissue (Figure 2.3) and interpreted the simulation results with two parameters using logistic curve fitting, summarising the neointimal growth curves in terms of a growth rate $r$ and a maximum of smooth muscle cells $N_m$.

Since there are no experimental or clinical investigations available measuring detailed time-response curves as illustrated in Figure 2.3, and showing the impact of injury on the growth rate of neointima in the initial few weeks, it is far from trivial to compare our model results against in vivo data. The effect of strut thickness on ISR has been assessed in human clinical trials [71,72,73,84]. The observed angiographic and clinical restenosis growth measured at 6 months follow-up suggest that a thicker strut is associated with greater neointimal growth. However, this effect was not observed in the output of simulations shown in Figure 2.3b where the number of smooth muscle cells ($N_m$) is almost constant, irrespective of strut size. In our model, SMCs proliferation is inhibited when the WSS exceeds a preset threshold. As the blood flux through the vessel is kept constant, a decreasing lumen diameter due to ISR results in a gradual increasing WSS, and the threshold value is reached at a more or less constant diameter (see also Figure 2.2), independent of strut size or deployment depth. As it is clear from both clinical data and the porcine data reported in Figure 2.6, such constant endpoint is not realistic and is a flaw of our current model. Despite this obvious problem in the long term dynamics of our model, it does provide more insight to the initial short time dynamics and provides the capability for investigating different hypotheses. According to our simulation results, a thicker strut results in a faster neointimal growth (Figure 2.3c) in the initial few weeks, which might explain the greater neointimal growth reported in clinical trials at 6 months post-implantation.
Figure 2.6: Neointimal thickness as function of time after stent deployment. The mean and standard deviation of the neointimal measurements are plotted at 6 hours, 14, 21 and 28 days post stent deployment. The data shows a positive correlation between neointimal thickness and the Gunn injury scores.

Schwartz et al [70] published NPAGF data calculated for several studies carried out in rats, pigs and humans. There is an excellent agreement when we compare the values of NPAGF from the computational output (Fig. 2.4a) to those published by Schwartz et al. As it is not possible to determine the strut deployment depth or injury level achieved in the porcine model used by Schwartz to generate the data, it is difficult to make a direct comparison. However, the values of NPAGF calculated from our simulation data are comparable to those produced by Schwartz, showing a peak in proliferation at around 20 days (Figure 2.4a). The lack of detail provided by Schwartz regarding the precise experimental conditions under which the data were generated highlights the need for better quantitative experimental data regarding IEL rupture and the need for improved documentation of experimental conditions. The most apparent qualitative difference in comparison to Schwartz’ data is the asymmetry of the curve where a long tail has been observed by Schwartz,
showing that the proliferation process continues for a longer period. This is not the case in our model where we commonly observed a symmetrical curve. This difference could be in part due to our stopping conditions where an increase in the WSS threshold stops the whole process of neointimal thickening. The simulation results also predict that the peak proliferation occurs earlier in case of deep injury / stretch (Figure 2.4b).

Another important finding of the current study is the effect of deep injury caused by the stent on the development of neointimal growth. If we ignore the limitation of the final endpoint, then the computational output follows the same qualitative patterns as the *in vivo* data; the degree of SMC proliferation increases with the degree of injury caused by stent deployment (Figures 2.3c and 2.6). There is one exception for a thin strut ($D=90\,\mu m$) where a reduction in the growth rate was observed at deployment depth of $130\,\mu m$. The *in vivo* data (Figure 2.6) indicates a decrease in the neointimal thickness measured after 28 days. This may be related to vessel remodelling; a feature which is not captured in the present computational model. Moreover, a clear correlation of this decrease with our simulation output is currently restricted due to the endpoint limitation. We aim to reinvestigate this after implementing a remodelling kernel in our CxA simulations.

Endothelial cell injury and dysfunction play a critical role in the pathogenesis of several cardiovascular diseases including atherosclerosis [85]. It is generally accepted that injury caused by stent deployment results in the denudation or dysfunction of endothelial cells [86] and re-endothelialisation has been observed to play an important role in halting the progression of stent induced neointimal growth [75]. In the current scenario, our model is dependent on a premise that the initial injury / stretch instigates the process of neointimal growth [60,69] and that the growth process stops in the presence of high WSS. WSS has been identified as a key player in the process of re-endothelialisation and neointimal formation. Delayed re-endothelialisation and greater neointimal growth seem to be associated with regions of low WSS [74,75]. The effect of WSS on the ISR has also been investigated by De Santis et al [87] where a WSS-based algorithm was used to simulate vessel remodelling. Results show that a high value of WSS was observed throughout the final remodelled geometry. This investigation highlights the importance of WSS with respect to ISR and describes WSS as a valuable predictor of the final lumen geometry. This effect can also be observed in our simulations where a high WSS threshold stops the whole neointimal growth process. It should be noted however that in the current study we have assumed steady flow conditions. A next step will be to study SMC behaviour under pulsatile flow
conditions, taking into account the oscillatory shear by defining an appropriate oscillatory shear index (OSI) threshold.

Drug eluting stents have been shown to reduce restenosis [66]. This effect is quite prominent in our simulations (Table 2.1) showing an inhibition of SMC proliferation, which results in a delayed response of the whole process of neointimal growth. Preliminary results from a round cross-section BMS strut also showed a decrease in the growth rate, resulting in a delayed neointimal growth. A reason suggested to explain this delayed growth is that the rounded shape results in a more stable flow within the artery lumen thus minimising or eliminating the effect of flow recirculation [62].

In summary, our single scale models, especially the SMC model, are still at an early stage of development and require further improvement. In the near future we will also run simulations with the 3-dimensional version of the model (thus taking into account realistic shapes of the stent, and shape and curvature of the vessel), coupling more single scale kernels (e.g. thrombosis, cell signalling, vessel remodelling etc.) to capture, more closely, the behaviour of this complex phenomenon. Such 3D simulations will require high-end distributed multiscale computing capabilities, as currently being developed within the MAPPER project (www.mapper-project.eu). However, despite the need for an improved 3D model, we should not disregard the importance and relevance of the current 2D model. We will continue to use the 2D model for rapid testing of new hypotheses. Further refinement of this will enable us to test different hypotheses and provide greater insight into the time-dependent behaviour of ISR. Refinement of the current 2D model will include implementation of pulsatile flow as well as the integration of an improved tissue model containing additional levels of complexity, such as addition of an endothelium, extra-cellular matrix (ECM) and the external elastic lamina (EEL). It may even be useful to investigate the effect of vessel remodelling in the 2D model as the effect of this is not well known.

Fitting the data to a logistic growth function helped us to characterise the dynamic response in terms of growth rate. Although we do not claim that ISR is a logistic growth process, this characterisation in terms of $r$ and $N_{max}$ may prove useful in the preclinical testing of stent designs, and may eventually provide a predictive tool for use in the clinical setting. We will therefore continue to calculate the restenotic lesion growth rate $r$ as a function of strut thickness, deployment depth and strut shape. Experimental studies providing a measure of ISR as a function of time would allow further validation of our models, and
more importantly, would enable us to test the hypothesis that the complex ISR response can be characterised by just a small set of parameters, as in logistic growth.