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Online Supplement "Cardiac sarcomere length in situ and in vitro", Bub et. al

The software package and source code developed for semi-automated SL measurement is
 freely available. Please contact <u>gil.bub@dpag.ox.ac.uk</u> for details.

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Supplement Section 1: Data analysis:

8 Off-line compensation for in-plane error: To correct for in-plane errors in path alignment 9 perpendicular to striation patterns, and to account for measurement errors associated with 10 artefacts associated with path length (resolution bias error), the program optionally iterates over a sequence of path angles (± 5 degrees from the original path, 1 degree steps) and lengths 11 $(\pm 10 \text{ pixels})$. For each angle the program averages the periods found for the range of ruler 12 13 lengths, and determines the magnitude of the DFT peak. Angles where the DFT peak magnitude are less than that of the original are discarded, and the user path angle is reset so 14 15 that the fluorescence intensity period is minimized. This method was found to converge to the 16 correct periodicity within 0.2 % for a wide range computer generated striation patterns. We opt to report uncorrected data in the present publication since the algorithm may 17 18 introduce a bias in the measurements. The computer generated striation patterns used to test 19 the algorithm assume a perfectly periodic underlying SL pattern. The corrected trajectories may underestimate SL if there are significant sarcomere-to-sarcomere length variations 20 21 within each cell.

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Additional filtering criteria for SL measurements: In order to remove measurements biased
by tissue motion or poor image quality, all measurements were filtered based on the relative
magnitude of the peaks of the DFT obtained from the trajectories intensity profiles.
Measurements were discarded if the magnitude of the highest DFT peak was less than 3 fold
greater than any neighbouring peak corresponding to a SL period between 1.4 to 3.0 µm.
Some data sets were also filtered based on experiment time (parsed from Leica meta-data
files on an image by image basis) and ruler angle in the image plane as described in Results.

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31 *Operator bias:* Each operator has to visually identify cells within each image that have clear 32 striation patterns prior to drawing a path through the cell. Since the initial choice of cells is 33 subjective, operators measure SL from different sub-populations of cells within the data sets. 34 In addition, the location of the path within each cell is operator-dependent; parallel paths within a cell that give acceptable intensity profiles and sharp frequency peaks may give 35 36 slightly different values for SL. In order to assess the effect of operator bias on the final 37 measurements, four data sets were independently analysed by three different operators. Initial 38 assessment of the three data sets revealed that one of the operators had drawn several 39 trajectories that crossed cell boundaries and were outside the 5 degree range that could be 40 compensated for by the off-line angle correction algorithm. The results from this operator had a significantly increased median SL value. The trajectories were re-drawn by the same 41 42 operator using stricter guidelines. The resulting corrected mean SL and distribution mode fell 43 within a 1% range of aggregate values.

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Supplement Section 2: Monte Carlo estimation of error bounds on measurement:

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48 Monte Carlo Estimates of SL distribution: In order to guide theoretical estimates of the lower

49 bound of SL_R , we simulate the effect of cell angle on SL_M as a function of length of the cell

50 in the imaging plane (L), prescribing a normal distribution of SL_R (mean 2.0 μ m, variance

51 0.01), cell dimensions of $l=142 \,\mu\text{m}$, $w=32 \,\mu\text{m}$, $h=13 \,\mu\text{m}$ (obtained from a widely referenced 52 study by Satoh et al. (26)), and randomly chosen orientations relative to the image plane (see 53 Methods). Figure S1 shows the probability distributions of SL_M for three different minima of 54 user-defined path lengths (L_{min}) for SL_R with the above stated normal distribution (black 55 continuous line). SL is well constrained for $L_{min} = 90 \ \mu m$ and $L_{min} = 70 \ \mu m$, but is poorly 56 constrained for $L_{min} = 50 \ \mu m$. However, it is evident from Figure S1a that the *mode* of the 57 distribution while biased towards longer SL with reduced L_{min}, is relatively unaffected. 58 59 Figure S1b plots mean measured SL and the *mode* of the distributions of SL for L_{min} between 60 40 and 120 μ m. Both the mean and *mode* overestimate the true mean of the underlying SL 61 distribution, but the magnitude of the overestimate for the mean is far greater than that of the 62 *mode*. This is particularly evident at shorter L_{min} . Although the *mode* of the distribution is 63 closer to the underlying SL than the measured mean, the accuracy of the mode estimate 64 depends on the sample size. Figure S1c shows an error estimate for the *mode* calculation as a function of n, determined from the standard deviation of the mode estimates for multiple 65 distributions of size *n*. The error associated with the *mode* estimate is high for small numbers 66 67 of samples, but reaches less than 0.02 μ m (1% of the base SL measurement) for n>15068 samples. 69 70 The Monte-Carlo simulations do not allow one to directly calculate mean SL, because cell 71 alignment relative to the imaging plane is not uniformly distributed in the real data (see 72 Discussion). For this reason, we conducted simulations to estimate an upper and lower bound 73 for the underlying mean of the distribution. A function of the magnitude of the estimates bounds can be obtained by first fitting an exponential to the results shown in figure S1b: 74 75 Eq. 1 $F(L_{min}) = 0.0599 \ e^{-0.0177 L_{min}} + 1.0$ 76 77 78 Similarly, the error in estimating the *mode* of the distribution is given by a fit to the 79 simulation results shown in figure S1c: Eq. 2 $G(n) = 0.0479 e^{-0.0198n} + 0.016$ 80 81 82 where *n* is the number of samples in the data set with $L \ge L_{min}$. The 'true' mean SL is expected 83 to lie between an upper bound (mode + G(N)) and a lower bound (mode-F(L_{min}) -G(n)). The 84 magnitude of the bounds is therefore given by: 85 Eq. 3 $H(L_{min}, n) = F(L_{min}) + 2 G(n)$ 86 87 88





90 FIGURE S1. Monte-Carlo simulations of the effect of cell angle on SL_M. A) Simulations, 91 assuming a completely random cell orientation relative to the image plane, shift and skew a 92 normally distributed SL distribution (dark line) as a function of minimum length of the 93 analysis path(L_{min}) in the image plane (diamonds: L_{min} = 50 µm, circles: L_{min} = 70 µm, stars: 94 $L_{min} = 90 \ \mu m$). B) SL_M overestimation error as a function L_{min} . Open diamonds show percent 95 increase in measured vs. actual mean; closed diamonds illustrate percent increase in measured 96 vs actual *mode*. The distributions *mode* is determined by kernel density estimation as 97 described in the text C) Standard deviation (SD) of the mode estimate as a function of the 98 number of samples in the distribution. Dashed lines in panels B and C are exponential

- 99 functions fitted to the simulation results, using a non-linear least-squares (NLLS) Marguardt-
- 100 Levenberg algorithm (Gnuplot, www.gnuplot.org).
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103 Section 3: Compensation for tissue motion:

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FIGURE S2. A: Distribution of sarcomere lengths measured from the intact Langendorff-106 perfused rat heart. Open bars: SL for all cells; Hatched bars: cells aligned within 30 degrees 107 108 of the fast scan axis of the two photon fluorescence microscope. B: SL as a function of cell 109 orientation in the image plane (cells at 0 degrees are aligned in parallel with the fast scan axis 110 of the two photon fluorescence microscope).

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112 Mean SL from the first 5 minutes of recording time (i.e. about 30 minutes after sacrifice) was 113 $2.05\pm0.18 \ \mu m \ (n = 300; N = 11 \ experiments)$. Means in individual experiments varied from 114

1.89±0.06µm to 2.36±0.10 µm. A histogram (Fig S2a, open bars) of SL_M shows a dominant

115 peak in SL frequency between 1.9 and 2.0 μ m, and a second lower peak between 2.3 and 2.4 116 μ m. The second peak is predominantly comprised of SL measurements from 2 experiments,

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although infrequently cells with long SL were observed in most experimental runs. Although 118 we do not rule out that a population of long SL cells exists in our preparation, further analysis

119 suggests that the second peak is caused by an experimental artefact, as described next.

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121 Figure S2b shows the dependence of measured SL and angle between the cell's long axis and

- 122 the imaging plane x-axis (corresponding to the fast scan axis of the microscope). Cells
- 123 aligned with the x-axis display shorter SL_M and less variance than cells at other angles. We
- 124 hypothesize that the in-plane angle dependence on SL is caused by the interplay of tissue
- 125 motion (unavoidable in Langendorff perfused living tissue), local cell length changes, the
- large difference in acquisition times for a single scan line (\sim 3ms) and a whole frame (1,625
- 127 ms), and the scanning direction of the two photon microscope excitation light. If we filter the
- data set by excluding SL measures that fall outside of $\pm 30^{\circ}$ of the fast scan axis of the
- microscope, we obtain a mean SL of $1.99\pm0.12 \,\mu\text{m}$. The histogram of filtered SL has a unimodel distribution (Figure 7a hetched here) with a neal hetware 1.0 and 2.0 μm .
- 130 unimodal distribution (Figure 7a, hatched bars) with a peak between 1.9 and 2.0 μ m.
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The presence of a correlation between in-plane imaging angle and SL_M supports the
 hypothesis that non-structural aspects, such as tissue motion, modulate SL_M in the intact

- 134 isolated heart. Two-dimensional images are built up from a series of sequentially acquired
- data points, resulting in data being rapidly collected for cells aligned along the x axis of an
- 136 image (3 ms), and two orders of magnitude more slowly (1.625 s) for cells aligned along the
- 137 y-axis of the image. Tissue drift can both increase and decrease SL_M , however we observe an 138 increase in SL_M with cell angle. We believe that this bias may caused by two mechanisms.
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First, compression due to motion distortion may reduce SL contrast. Since operators tend to
 favour images of cells with clear, well separated striation patterns to define long analysis

142 paths, cells with short SL may be under-sampled in images with both motion-induced

- 143 reduction and enhancement of apparent SL.
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Alternatively, motion may be caused by transient sporadic cell contractions, as opposed to

bulk movement of the entire tissue. Although the arrested tissue is largely quiescent,

147 intermittent local contractions are nonetheless expected to occur. These regional transient

148 contraction events would stretch cells in the surrounding tissue volume. Assuming that the

area in which stretched tissue may be found is greater than that occupied by the contracting

cell population, and taking into account the small imaging field of view, SL measurements

- are more likely to cell stretch rather than contraction.
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Both the tissue drift and local tissue contraction mechanisms suggested above are speculative,

and at present we do not have a method for confirming or refuting either of them. It is clear,

- however, that experimental procedures may be optimized by using the fast scan direction of
- the microscope as a guide for cell alignment before acquisition of images.
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