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## Machine Learning Insights into Probing TRSB via Asymmetric SQUIDs

THESIS

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## Machine Learning Insights into Probing TRSB via Asymmetric SQUIDs

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#### Abstract

The BCS theory of superconductivity is associated with time reversal symmetry. Some unconventional superconductors have shown time reversal symmetry breaking (TRSB). Scanning SQUIDs may be used to measure features of TRSB. However, probing TRSB using SQUIDs requires sensitivity at zero field, which only asymmetric SQUIDs achieve. Measurements of three SQUIDs asymmetric in arm length, one SQUID asymmetric in arm width and one symmetric SQUID are presented. Furthermore, the parameter values of the SQUIDs are fitted using a neural network that simulates SQUID voltage response.

| Chapter

## Introduction

The Bardeen-Cooper-Schrieffer (BCS) theory is the first theory that attempts to explain the superconductivity discovered by Kamerlingh Onnes's in 1911. It explains superconductivity phenomena as a microscopic effect due to condensation of so-called Cooper pairs. Bardeen, Cooper and Schrieffer received the Noble prize in Physics for this theory in 1972. One feature of the BCS theory is that the two electrons in a Cooper pair have opposite spin and momentum. This entails that the state of the whole Cooper pair is symmetric under reversal of time (TRS). Superconductors that are well described by BCS theory are named conventional.

Nonetheless, experimental techniques like muon spin rotation ( $\mu$ SR) and the polar Kerr effect have been used to detect time reversal symmetry breaking (TRSB) [1–3]. For example, Mayoh, D. A. et al provide evidence for TRSB in La7Pd3 using ÎŒSR [4]. Figure 1.1 shows increased relaxation below  $T_c$ , indicative of internal magnetic fields which are associated with TRSB.

Some superconductors with TRSB also shown 'domain walls' between areas with different orientations of the superconducting order parameter [5]. Consequently, these domain walls exert noticable patterns in the magnetic field of a superconductor. Garcia Campos, P. uses a scanning superconducting quantum interference device (SQUID) microscope to visualise the presence of TRSB in UPt3 [6]. Figure 1.2 shows magnetic field measurements using a SQUID microscope on UPt3, indicating TRSB via the presence of a domain wall (which appears roughly as an s-shape).

Measuring TRSB with SQUIDs requires sensitivity around zero magnetic field. Asymmetries in the inductances of the two SQUID arms (denoted by the asymmetry parameter  $\alpha_L$ ) and the critical currents of the two junctions (denoted by the asymmetry parameter  $\alpha_I$ ) can create sensitivity



**Figure 1.1:** (a) ZF and LF- $\mu$  SR spectra collected at 0.1 (green) and 2.75 K (red), the data is fit using the Gaussian Kubo-Toyabe model (dashed lines). (b) Temperature dependence of the electronic relaxation rate L can be seen to increase below 1.2 K just below Tc. (c) Temperature dependence of the nuclear relaxation rate  $\Lambda$  shows no change at Tc. Figure and caption from ref. [5].

at zero field. (The SQUID used for measurements in Figure 1.2 has asymmetric inductance.) In this thesis, I present measurements and analysis of geometrically asymmetric SQUIDs, with the goal of finding out how to create SQUIDs suitable for domain wall measurements of superconductors exhibiting TRSB.

The sensitivity of a SQUID around zero field can be analysed using raw measurement data. Another method of analysis is to fit measured data to simulated data to obtain values for  $\alpha_L$  and  $\alpha_I$ . Simulating the voltage response of a dc SQUID involves solving a stochastic differential equation. Fitting can therefore be a costly process. Neural networks have shown potential to drastically speed up physics simulations [7]. In that line, I will analyse my data of geometrically asymmetric SQUIDs using neural network simulations.



**Figure 1.2:** Scans with -400 T applied field before cooling down at a scanning height of 350 nm. The scans were taken decreasing the temperature (from the A-phase to the B phase). Between 450 mK and 440 mK, a sinusoidal domain wall appears, which creates an unusual arrangement of the magnetic flux. At the coldest temperature (250 mK), the mean f ield of the whole scan matches the applied field, and the mean field of one of the two chiral phases (white ROI). Figure and caption from ref. [6].

## Theory

#### 2.1 London equations & penetration depth

A fundamental feature of superconductivity is the Meissner effect, where a superconductor expels all magnetic fields inside the superconductor. The London equations provide an explanation of the decay of external magnetic fields inside of superconductors [8].

The first London equation is

$$\frac{\partial j_s}{\partial t} = \frac{n_s e}{m_e} E \tag{2.1}$$

Followed by the second London equation:

$$\boldsymbol{\nabla} \times \boldsymbol{j}_s + \frac{n_s e^2}{m} \boldsymbol{B} = 0 \tag{2.2}$$

This ultimately leads to

$$\nabla^2 B = \frac{1}{\lambda_L^2} B \tag{2.3}$$

with the London penetration depth  $\lambda_L = \sqrt{\frac{m_e}{\mu_0 n_s e^2}}$ . This means that external magnetic fields in superconductors decay exponentially.

## 2.2 The Josephson effect

A Josephson junction consists of two superconducting electrodes with a weak link connecting them [9]. The type of weak link determines the type

of junction. In the cases of SNS junctions or SIS junctions, the weak link is provided by a normal metal layer or insulating layer respectively. The first Josephson relation provides a one-to-one relation between the supercurrent across two weakly linked superconducting electrodes and the phase different between them for an SIS junction

$$I_s = I_0 sin\delta \tag{2.4}$$

Here  $I_s$  denotes the supercurrent across a SIS junction,  $I_0$  the critical current at which voltage becomes non-zero and  $\delta$  the gauge invariant phase difference. The second Josephson relation relates the voltage U that is developed across the junction to the time derivative of the gauge invariant phase difference  $\dot{\delta}$ 

$$\dot{\delta} = \frac{2\pi}{\Phi_0} U \tag{2.5}$$

Where  $\Phi = h/2e$  is the magnetic flux quantum.

The resistively- and capacitively-shunted junction (RCSJ) model of a Josephson junction provides a differential equation relating the phase difference, current and potential difference across the junction. In this model, we apply Kirchhoff's law to the circuit of Figure 2.1 and get:

$$C\dot{U} + \frac{U}{R} + I_0 sin\delta = I + I_N(t)$$
(2.6)

where *C* is the capacitance, *R* is the resistance, *I* is the bias current and  $I_N(t)$  is the Nyquist noise current. Using the second Josephson relation, we get:

$$\frac{\Phi_0}{2\pi}C\ddot{\delta} + \frac{\Phi_0}{2\pi R}\dot{\delta} = I - I_0 sin\delta$$
(2.7)

The current-voltage characteristics of Josephson junctions can show hysteresis: when you go from zero to a current above  $I_0$ , I(V) is different than when you go from a current above  $I_0$  to zero. Extermination of hysteresis is usually achieved by using an external shunt resistance in the system with the Josephson junction. This is why the current-voltage characteristics of systems without hysteresis are well explained by the resistivelyand capacitively-shunted junction (RCSJ) model.

#### 2.2.1 DC SQUIDs

Direct current superconducting quantum interference devices (dc SQUIDs) are superconducting loops with two arms, where each arm contains a junc-



**Figure 2.1:** Electrical circuit of the RCSJ model. Starting from the left, the capacitance is located at *C*, the resistance at *E*, followed by the Josephson junction with critical current  $I_0$ , a Nyquist noise current  $I_N$  and lastly a potential difference *U*.

tion as shown in Figure 2.2.

The current through each junction is the sum of a contribution from the bias current and a contribution from the circulating current. That is,  $I_1 = \frac{I}{2} + J$  and  $I_2 = \frac{I}{2} - J$ . Combining this with Eq. 2.7 gives us

$$\frac{I}{2} \pm J = I_{0,k} sin\delta_k + \frac{\Phi_0}{2\pi R_k} \dot{\delta_k} + \frac{\Phi_0}{2\pi C_k} \ddot{\delta_k} + I_{N,k}$$
(2.8)

with k = 1, 2 for the separate junctions.

For any point in the loop visible in Figure 2.2, the wave function must be single-valued. This means that when a path is traversed around the loop from one point to itself, the sum of the phase changes along that path must equal an integer amount of  $2\pi$ . In the limit of bulk electrodes there are two notable sources of phase change in the dc SQUID: traversing a Josephson junction and traversing a magnetic vector potential. The latter of these effects is commonly accredited to flux, because a closed line integral over the magnetic vector potential is equal to a closed surface integral (Stokes's theorem) over the magnetic field. This in turn can be simplified to a flux quantity  $\Phi$ .

Following a path around a loop in the dc SQUID, we find that the direction with which the first junction is encountered differs from the second junction. That is, as a path is traversed around the loop, the first junction



**Figure 2.2:** A schematic of the dc SQUID. I represents the external bias current.  $I_1$  and  $I_2$  represent the portions of the bias current going through each arm (and junction). The circulating current induced by the junctions is J. The Josephson juntions are visualised by two black strokes.

contributes to a phase change of  $\delta_1$  while the second junction contributes to a phase change of  $-\delta_2$ . The phase change due to flux has contributions from both the external magnetic field and the current density *J*. Then, the single-valuedness of the wave function at any point in the loop requires that

$$\delta_1 - \delta_2 + \frac{2\pi}{\Phi_0} \left( \Phi_a + LJ \right) = 0 \tag{2.9}$$

with  $\Phi_a$  being the applied flux, *L* being the total inductance. This may be rewritten as

$$\delta_2 - \delta_1 = \frac{2\pi}{\Phi_0} \Phi_T \tag{2.10}$$

with  $\Phi_T = \Phi_a + LJ$ .

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Eq. 2.8 may be normalized using  $I_0 = (I_{0,1} + I_{0,2})/2$  for the average critical current,  $R = 2R_1R_2/(R_1 + R_2)$  for twice the parallel resistance and  $C = (C_1 + C_2)/2$  for the capacitance. Defining the Stewart-McCumber parameter  $\beta_c = \frac{2\pi}{\Phi_0}I_0R^2C$  and the screening parameter (also called nor-

malised inductance)  $\beta_L = \frac{2LI_0}{\Phi_0}$ , we may write Eq. 2.8 in normalised form:

$$\frac{i}{2} \pm j = \sin\delta_k + \dot{\delta}_k + \beta_c \ddot{\delta}_k + i_{N,k}$$
(2.11)

with k = 1, 2 for two junctions and

$$\delta_1 - \delta_2 + 2\pi \left( \phi_a + \frac{1}{2} \beta_L j \right) = 0 \tag{2.12}$$

where  $\phi_a = \frac{\Phi_a}{\Phi_0}$ .

One last note about the dynamics of (dc) SQUIDs is that the effect of the Nyquist noise term on IV's is characterised by the noise parameter  $\Gamma$ 

$$\Gamma = \frac{k_b T}{E_I} = \frac{2\pi k_b T}{\Phi_0 I_0} \tag{2.13}$$

In practice,  $\Gamma$  leads to a rounding of the superconducting transition at the critical current. Figure 2.3 shows how rounding occurs in the voltagecurrent characteristics of a dc SQUID, which is particularly pronounced at the critical current.

#### 2.2.2 Asymmetry parameters

The two junctions in a dc SQUID may have different critical currents, resistances and capacitances. Furthermore, the different arms of the SQUID to which the junctions are attached may have different inductances. SQUIDs with such differences may be called asymmetric. Asymmetry is modelled by the parameter  $\alpha$ . It is defined such that

$$I_{0,1} = I_0(1 - \alpha_I), \quad R_1 = \frac{R}{1 - \alpha_R}, \quad C_1 = C(1 - \alpha_C)$$
 (2.14)

$$I_{0,2} = I_0(1 + \alpha_I), \quad R_2 = \frac{R}{1 + \alpha_R}, \quad C_2 = C(1 + \alpha_C)$$
 (2.15)

for the critical currents, resistances and capacitances of the two junctions. Adding these parameters to Eq. 2.11, we get

$$\frac{i}{2} \pm j = (1 \mp \alpha_I) \sin \delta_k + (1 \mp \alpha_R) \dot{\delta}_k + \beta_c (1 \mp \alpha_C) \ddot{\delta}_k + i_{N,k}$$
(2.16)

Furthermore, for asymmetric arm inductance, Eq. 2.12 changes into:



**Figure 2.3:** Simulated normalised voltage-current characteristics of a dc SQUID with two different levels of voltage noise. The red scatter plot has  $\Gamma = 0.05$ , whereas the black scatter has  $\Gamma = 0.00$ .

$$\delta_2 - \delta_1 = 2\pi \left( \phi_a + \frac{1}{2}\beta_L j - \frac{1}{4}\alpha_L \beta_L i \right) = 2\pi \phi_a + \pi \beta_L \left( j - \frac{\alpha_L}{2} i \right), \quad (2.17)$$

These equations should suffice for modelling the dynamics of a dc SQUID with asymmetric critical current, resistance, capacitance and inductance according to the RCSJ model.

Correspondingly, the shift between the maxima of the negative and positive critical current oscillations due to asymmetry is given by [10]:

$$\frac{\Delta\Phi}{\Phi_0} = \beta_L(\alpha_L + \alpha_I) \tag{2.18}$$

Figure 2.4 shows how asymmetry leads to a shift in measured voltages as a function of current and flux.

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**Figure 2.4:** Simulated SQUID interference pattern with  $\alpha_L = 0.25$  and  $\beta_L = 1$  (with symmetrical critical currents and resistances) operating in the over damped limit ( $\beta_c \rightarrow 0$ ). The color transition from black to red points to the critical current. The combination of parameters induces a shift between corresponding peaks in negative and positive critical current oscillations. The shift is  $\frac{1}{4}$  normalised flux units.

Chapter 3

## Design and methods

We are interested in dc SQUIDs that are shifted in this manner as to create a field-sensitive dc SQUID around a magnetic field/flux of zero. I provide a two-component definition of 'field-sensitive' at a given field  $B_g$ :

- 1. There must be some current for which there is a non-zero transfer function  $\delta V / \delta B$  at  $B_g$ .
- 2. For such a current the transfer function  $\delta V / \delta B$  must behave linearly within a range  $\rho$  around  $B_g$ , such that  $\rho$  is approximately equal to the desired range of field measurements.

If a dc SQUID obeys (1) but not (2), field measurements (through voltage measurements) will fail to measure at the complete desired range of field measurements. Thus, a positive transfer function ( $\delta V / \delta B > 0$ ) at  $B_g$  is necessary but insufficient for field-sensitivity at  $B_g$ .

Figure 3.1 illustrates how asymmetry induces a shift towards a positive transfer function at zero field (for some dc currents).

## 3.1 Geometry design and kinetic inductance

For a straight wire of length *l* with cross section *A*, the kinetic inductance is given by [11]:

$$L_k = \frac{ml}{nq^2A} = \frac{\mu_0 \lambda_L^2 l}{A} \tag{3.1}$$

With  $\lambda_L$  the London penetration depth from Eq. 2.3.

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**Figure 3.1:** Transfer function  $\frac{\delta v}{\delta \phi}$  of the normalised voltage  $v = V/I_0R_n$  as a function of the normalised current *i* and the normalised flux  $\phi$  with  $\Gamma = 0.0$  and  $\beta_L = 1.0$ . (a) shows the simulated interference pattern for a symmetric SQUID. In (b),  $\alpha_L = 0.25$ . The white dashed lines are positioned at zero field/flux. For (b), but not for (a), there are currents for which there is a positive transfer function at zero field.

According to Eq. 2.18, asymmetry in the inductance of the two arms of a dc SQUID induces a shift. For thin films, it may be hypothesized that the kinetic inductance is the main term in the total inductance, and that magnetic inductance is negligible. As Eq. 3.1 shows, the inductance of a wire with an approximately uniform cross section has two geometric components: the length *l* and the cross section *A*. In the case of a thin film, the cross section can further be divided into a thickness *t* and a width *w*, as to approximate a rectangular cross section. Whereas  $L_k$  is proportional to *l*, it is inversely proportional to *t* and *w*. Nonetheless, inductance asymmetry in any of these geometric parameters should theoretically translate to an asymmetry in inductance according to Eq. 3.1.

I therefore attempt to design inductively asymmetric SQUIDs. In thinfilm nanofabrication process, the thickness of the film is generally kept constant. Therefore, I design SQUIDs that are asymmetric in the lateral dimensions *l* and *w* of each arm of the SQUID. Figure 3.2 shows two sample geometrically asymmetric SQUID designs that are intended to induce a non-zero value for the inductance asymmetry parameter  $\alpha_L$ .

#### 3.2 Methods

To test the designs, two SQUIDs devices are made on top of Si/SiO substrates. Before deposition, lithography is used to prepare device patterns



**Figure 3.2:** geometrically asymmetric SQUID designs. In (a), the lengths of the two arms are symmetric ( $l_1 > l_2$ ). In (b), the arms lengths are symmetric, but the arm widths are asymmetric ( $w_1 < w_2$ ). If we approximate the loop wire as a straight wire with uniform cross sections, the kinetic inductance will be asymmetric according to Eq. 24.

that allow wire-bonding of the devices. The used method of deposition is ultra-high vacuum (UHV) sputtering. The first layer deposited is Cu, followed by Nb and lastly Pt. The Cu layer will function as the normal metal layer separating Nb superconducting electrodes. Lastly, Pt has been used as a capping layer, to provide protection for the Nb layer.

After deposition, SQUID patterns are milled out of the metal layers using focused ion beam (FIB) milling. We use a ThermoFisher Aquilos dual beam-FIB, provided by NeCEN [12].

Lastly, the contact pads of the devices containing the SQUIDs are wirebonded onto the contact pads of the insert puck.

We perform our measurements inside a TeslatronPT cryostat by Oxford Instruments, with a base temperature of 1.5 K. All transport measurements are carried out using standard lock-in techniques with a SynkTek MCL1-540.



## Results

Five SQUIDs have been measured in total. The next sections will show the devices and the measurements.

## 4.1 Febricated devices

Two chips are fabricated. Chip  $\alpha$  has a 20 nm Cu layer on the bottom, stacked by 60 nm Nb and 4 nm Pt. Chip  $\alpha$  has a 20 nm Cu layer on the bottom, stacked by 80 nm Nb and 4 nm Pt. Figure 4.1 shows scanning electron microscope (SEM) images of the FIB fabricated SQUIDs for chip  $\alpha \& \beta$ . Figure 4.1 shows a total of five SQUIDs: one of which is symmetrically designed (7a), one is designed with asymmetric arm thickness (7c), and three are designed with asymmetric arm lengths (7b, 7d and 7e).

#### 4.2 **Resistance versus temperature**

Figure 4.2 shows the measured resistances as a function of temperature (RT) for all three SQUIDs from chip  $\alpha$  (left plot), and SQUID A from chip  $\beta$  (right plot).

In the left plot, the symmetric SQUID A has a significantly lower resistance above  $T_c$ . The critical temperature can be roughly read off by looking at the point where the resistance goes below  $10^{-2}$ . For SQUID  $\alpha$ A this is at 3 K, for SQUID  $\alpha$ B this is at 2.5 K and for SQUID  $\alpha$ C this is at 2.7 K. SQUID  $\beta$ A has a notably higher  $T_c$  than all of these at 5 K. This can be attributed to wider arms than those on chip  $\alpha$  (see Figure 4.1) and greater thickness.



**Figure 4.1:** SEM images of FIB milled SQUIDs on chip  $\alpha \& \beta$ , with SQUIDs named following the Latin alphabet for each chip. (a)-(c) show SQUIDs from chip  $\alpha$ , whereas (d) and (e) show SQUIDs from chip  $\beta$ . (b)-(e) are asymmetric: (b), (d) and (e) have asymmetric arm lengths, whereas (c) has asymmetric arm thickness. (a) is symmetric.

## 4.3 SQUID interference patterns

SQUID interference (SQI) patterns are measured by taking IV's over a linear space of magnetic fields. Figure 4.3a-e show SQI's for all five SQUIDs presented in this thesis. Figure 4.3a and 4.3b show the same range of magnetic fields, and their  $I_c$  oscillations show a similar period. If one looks at



**Figure 4.2:** Resistance as a function of temperature for (left) all SQUIDs on chip  $\alpha$ , and (right) for SQUID A of chip  $\beta$ .

Figure 4.3c-e, one notes that there is a red border around the  $I_c$  oscillations that has roughly uniform thickness throughout the oscillation. This entails that the voltage increases at roughly the same rate starting at  $I_c$  for all magnetic fields. However, for Figure 4.3a-b (showing SQUIDs  $\alpha$ A and  $\alpha$ B), this is not the case. Simulations show that the uniform 'border thickness' (i.e. rate of voltage increase around  $I_c$ ) is typical in the RCSJ model (see for example Figure 2.4). As Figure 4.1 shows, the arms of SQUIDs  $\alpha$ A and  $\alpha$ B have relatively thin arms. This may lead to a superconducting transition in the arms at a current close to  $I_0$ , the maximum critical current of the junctions. This would lead SQUIDs  $\alpha$ A and  $\alpha$ B to less closely resemble the RCSJ model.



**Figure 4.3:** SQUID interference patterns for SQUIDs on chip  $\alpha \& \beta$ .

## Chapter 5

# Fitting neural network simulations to data

The differential equations provided by the RCSJ model can be simulated to obtain voltages given a set of parameters, currents and fluxes. These simulations can however be temporally costly, requiring over one second to simulate a single voltage. Thus, if one wants to fit measured SQUID data, there is a strenuous contraint on the amount of simulations that one can compare with one's measurements.

Neural networks can speed up this process, because they consist of layers of functions that connect their inputs and outputs, and thereby remove the need to directly solve differential equations. In the next sections, I shall share the training process and architecture of neural networks that I have trained to simulate the RCSJ model, and use it to fit my measured SQUID data.

## 5.1 Neural network training and architecture

I have trained two neural networks. The first, which I shall name SQUIDnet 1.0, is trained on a dataset of simulated voltages, where the parameters  $\alpha_L$ ,  $\alpha_I$ ,  $\beta_L$ ,  $\beta_c$  and  $\Gamma$  are varied. The second model, which I shall call SQUIDnet 1.1, is trained on a dataset where  $\alpha_L$ ,  $\alpha_I$ ,  $\beta_L$  and  $\Gamma$  are varied, as well as the amplitudes of the second to fourth harmonics of the currentphase relations of each junction (separately). The amplitude of the first harmonic is constant at unity, so the total amount of varying higher order harmonic terms is six (three for each of two junctions).  $\beta_c$  is kept constant in the training data of SQUIDnet 1.1, as it came to my attention that simulated voltages were invariant to changes in  $\beta_c$  in the overdamped limit. Lastly, because of computational constraints, the training data for SQUIDnet 1.1 has been created on the ALICE supercomputer.

The SQUIDnet 1.0 model takes  $\alpha_L$ ,  $\alpha_I$ ,  $\beta_L$ ,  $\beta_c$ ,  $\Gamma$ , *i* and  $\phi$  as inputs. It then consists of five layers of nodes, followed by a predicted voltage output layer of a single node. The activation functions for the five hidden layers are (respectively) ReLu, Sigmoid, ReLu, Sigmoid and lastly ReLu. Each of these layer contains 64 nodes. The mean squared error (MSE) of SQUIDnet 1.0 on a validation dataset (randomised sample of dataset that has not been used for training) was in the order of  $10^{-5} v = \frac{V}{I_0 R_n}$  units.

The SQUIDnet 1.1 model takes  $\alpha_L$ ,  $\alpha_I$ ,  $\beta_L$ ,  $\Gamma$ ,  $a_{12}$ ,  $a_{13}$ ,  $a_{14}$ ,  $a_{22}$ ,  $a_{23}$ ,  $a_{24}$ , i and  $\phi$  as inputs, where  $a_{ij}$  represents the *j*th sine harmonic of the *i*th junction. It has five hidden layers with the same activation functions as SQUIDnet 1.0, and also a single voltage prediction output node. Next to difference in input parameters, SQUIDnet 1.1 differs from its predecessor in that each hidden layer contains 256 rather than 64 nodes. This also makes SQUIDnet 1.1 significantly slower at prediction than 1.0. The MSE of SQUIDnet 1.1 on a validation dataset was in the order of  $10^{-4} v = \frac{V}{I_0 R_n}$  units.

## 5.2 Fits of SQUID data

A gradient descent algorithm has been applied to minimize the mean squared error (MSE) between measured and simulated data. Figures 5.2-5.6 illustrate fits for SQUIDs  $\alpha$ C,  $\beta$ A and  $\beta$ B. Estimation of error on parameters has shown tedious if not unachievable. Instead, Figures 5.2-5.6c show histograms of the absolute errors with regards to measured data for each respective fit.

Table 5.1 shows fitted values for  $\alpha_L$ ,  $\alpha_I$ ,  $\beta_L$ ,  $\Gamma$ , the period,  $I_0$ ,  $R_n$  and the slope of the linear fits in figures 5.2-5.6d (which should approximate  $R_n$  when  $I_{dc}$  is well over  $I_0$ . These parameter values need to be interpreted with the error of the measured data, the neural network and the computational constraints of the gradient descent in mind.

It may be noted that although four out of five SQUIDs are geometrically asymmetric, the biggest absolute value of  $\alpha_L$  is 0.1 (for SQUID  $\beta A$ , see table 5.1). Furthermore, the SQUID with asymmetric arm widths has a large  $\alpha_I$  of 0.696. One way to explain the unexpectedly low  $\alpha_L$  is that the contribution to the total inductance from the geometric inductance is greater than expected. It was hypothesized that kinetic inductance dominates, which is proportional to length. However, our fitted  $\alpha_L$  values don't

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match the asymmetry in arm length, which should have an  $\alpha_L$  of approximately 0.3. At the same time, SQUIDs  $\alpha B$ ,  $\beta A$  and  $\beta B$ , which are the only SQUIDs with asymmetric arm lengths, also have the highest fitted absolute values for  $\alpha_L$ .

Parameter	αA	<i>α</i> B (As. l)	$\alpha C$ (As. w)	$\beta A$ (As. 1)	$\beta$ B (As. 1)
$\alpha_L$	0.008	0.082	0.002	-0.100	-0.030
$\alpha_I$	0.096	0.050	0.696	0.013	0.028
$\beta_L$	0.862	1.195	0.239	1.32	1.34
Γ	0.009	0.006	0.011	0.035	0.109
Period (mT)	28.7	29.0	13.0	15.6	20.9
$I_0 (\mu A)$	86	15	12	52	31
$R_n(\Omega)$	3.12	23.28	24.99	5.39	8.43
а	2.59	17.13	21.44	7.64	4.66

**Table 5.1:** Fitted parameters for all five SQUIDs. The variable *a* refers to the slope of the linear fit in figures 5.2-5.6(d). In the top row, asymmetric SQUIDs have a note with 'As' (followed by 1 or w, denoting asymmetry in either arm length or arm width respectively). Figure 5.1 (below) shows SEM images of the SQUIDs.



**Figure 5.1:** SEM images of FIB milled SQUIDs on chip  $\alpha \& \beta$ . (a)-(c) show SQUIDs from chip  $\alpha$ , whereas (d) and (e) show SQUIDs from chip  $\beta$ . (b)-(e) are asymmetric: (b), (d) and (e) have asymmetric arm lengths, whereas (c) has asymmetric arm thickness.



**Figure 5.2:** Snippet of measured data (a) for SQUID  $\alpha A$  versus fitted data (b) using SQUIDnet 1.0. (c) shows the frequency distribution of the absolute errors between data from (a) and (b) in  $\mu V$ . (d) shows overlap between measurement and fit for a single IV (located in (a) and (b) at vertical lines). (d) also includes a linear fit on the measured IV above  $I_C$ , so that its slope *a* approximates the sheet resistance  $R_n$ .



**Figure 5.3:** Snippet of measured data (a) for SQUID  $\alpha$ B versus fitted data (b) using SQUID net 1.0. (c) shows the frequency distribution of the absolute errors between data from (a) and (b) in  $\mu$ V. (d) shows overlap between measurement and fit for a single IV (located in (a) and (b) at vertical lines). (d) also includes a linear fit on the measured IV above  $I_C$ , so that its slope *a* approximates the sheet resistance  $R_n$ .



**Figure 5.4:** Snippet of measured data (a) for SQUID  $\alpha$ C versus fitted data (b) using SQUIDnet 1.1. (c) shows the frequency distribution of the absolute errors between data from (a) and (b) in  $\mu$ V. (d) shows overlap between measurement and fit for a single IV (located in (a) and (b) at vertical lines). (d) also includes a linear fit on the measured IV above I<sub>C</sub>, so that its slope *a* approximates the sheet resistance  $R_n$ .



**Figure 5.5:** Snippet of measured data (a) for SQUID  $\beta A$  versus fitted data (b) using SQUIDnet 1.0. (c) shows the frequency distribution of the absolute errors between data from (a) and (b) in  $\mu V$ . (d) shows overlap between measurement and fit for a single IV (located in (a) and (b) at vertical lines). (d) also includes a linear fit on the measured IV above  $I_C$ , so that its slope *a* approximates the sheet resistance  $R_n$ .



**Figure 5.6:** Snippet of measured data (a) for SQUID  $\beta B$  versus fitted data (b) using SQUIDnet 1.0. (c) shows the frequency distribution of the absolute errors between data from (a) and (b) in  $\mu V$ . (d) shows overlap between measurement and fit for a single IV (located in (a) and (b) at vertical lines). (d) also includes a linear fit on the measured IV above  $I_C$ , so that its slope *a* approximates the sheet resistance  $R_n$ .



**Figure 5.7:** Asymmetry parameters for fits in figures 5.2-5.6. On the horizontal axis, asymmetric SQUIDs have a note with 'Asym' (followed by l or w, denoting asymmetry in either arm length or arm width respectively).

# Chapter 6

## Conclusion

It can first be concluded that neural networks are apt to speed up simulations of the RCSJ model. This allows for fitting measured SQUID data within an acceptable time span. SQUIDs have been made by varying the lengths and widths of the SQUID arms. Varying the lengths of the two arms does not change the fitted  $\alpha_L$  values as much as one would expect with the hypothesis that kinetic inductance dominates. Still, the SQUIDs with asymmetric arm length did have the highest fitted  $\alpha_L$  values. It may be concluded that asymmetric arm length did affect  $\alpha_L$  to some extent, but that kinetic inductance does not dominate other effects like geometric inductance. In addition, asymmetric arm width induced a large  $\alpha_I$ .

Furthermore, the SQUID with asymmetric arm widths also showed a kink in its critical current oscillation. This is due to the presence of the second harmonic in the current phase relation. It would be interesting to see whether this presence of higher harmonics can be reproduced. If so, this would require a new theoretical explanation for the relation between asymmetric arm or junction width and the presence of higher harmonics. If not, then variation of the arm widths is a highly effective way to make a SQUID suitable for TRSB measurements at zero field, because it creates a large  $\alpha_I$ .

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