

BOSTON UNIVERSITY  
GRADUATE SCHOOL OF ARTS AND SCIENCES

Dissertation

**MUON SPIN PRECESSION FREQUENCY EXTRACTION  
AND DECAY POSITRON TRACK FITTING IN RUN 1  
OF THE FERMILAB MUON  $g - 2$  EXPERIMENT**

by

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Submitted in partial fulfillment of the  
requirements for the degree of  
Doctor of Philosophy

2020

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## **Dedication**

I dedicate this thesis to

## **Acknowledgments**

Here go all your acknowledgments. You know, your advisor, funding agency, lab mates, etc., and of course your family.

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(Order No. )

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ABSTRACT

One of the few indications for new physics is the discrepancy between the theoretical and experimental values for the anomalous magnetic moment of the muon. There is a discrepancy of 3 to 4 standard deviations between theory and the last experimental measurement made at Brookhaven National Laboratory in 2001, which measured the muon magnetic anomaly  $a_\mu$  to 540 parts per billion (ppb). This discrepancy has been consistent for many years with ever improving theoretical calculations. In order to resolve or confirm this discrepancy experiment E989, Muon  $g - 2$ , is underway to measure  $a_\mu$  to 4 times higher precision at 140 ppb. In Run 1 E989 gathered its first production data, consisting of approximately  $8 \times 10^9$  decay positrons above an energy threshold of 1.7 GeV.

This dissertation describes the experimental measurement, the detectors, the precession frequency extraction, and the track fitting of decay positrons in Run 1. The track fitting is done using a  $\chi^2$  minimization algorithm to fit tracks propagated within

a Geant4 reconstruction simulation including error propagation. The precession frequency is extracted using an analysis technique called the Ratio Method. The Ratio Method takes the ratio of time-shifted decay positron spectra in order to remove the decay exponential along with slowly varying effects in the data. Precession frequency extraction analyses for four near-final Run 1 datasets are presented with full systematic error evaluations. The total Run 1 precession frequency error determined in this analysis is 491.2 ppb, where the error is statistics dominated. Combined with the expected error in the magnetic field measurement of 140 ppb, the expected final error on  $a_\mu$  for Run 1 of E989 is  $\mathcal{O}(510 \text{ ppb})$ , comparable to the previous measurement.

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## List of Abbreviations

ADT	.....	artificial dead time
ASDQ	.....	Amplifier Shaper Discriminator with charge (Q)
ASIC	.....	application specific integrated circuit
BNL	.....	Brookhaven National Laboratory
BSM	.....	beyond the standard model
CBO	.....	coherent betatron oscillation
E821	.....	Brookhaven Muon $g - 2$ Experiment
E989	.....	Fermilab Muon $g - 2$ Experiment
EW	.....	electroweak
FID	.....	free-induction decay
FNAL	.....	Fermi National Accelerator Laboratory
Geane	.....	Geometry and Error Propagation
Geant	.....	Geometry and Tracking
HlL	.....	hadronic light-by-light
HVP	.....	hadronic vacuum polarization
IBMS	.....	inflector beam monitoring system
MIP	.....	minimum-ionizing particle
NMR	.....	nuclear magnetic resonance
PMT	.....	photo-multiplier tube
ppb	.....	parts per billion
ppm	.....	parts per million
ppt	.....	parts per trillion
QCD	.....	quantum chromodynamics
QED	.....	quantum electrodynamics
RMS	.....	root mean square
SAM	.....	Sequential Access via Metadata
SDT	.....	shadow dead time
SGT	.....	shadow gap time
SiPM	.....	silicon photo-multiplier
SM	.....	Standard Model
STDP	.....	short-term double pulse
VW	.....	vertical waist

# Chapter 1

## Introduction

The prevailing theory of particle physics, the Standard Model (SM), has had tremendous success in describing our universe. It has been used to predict and explain a wide variety of phenomena, particle properties, and interactions to great precision. However, in spite of its success in explaining nearly all experimental results, there remain unanswered questions about our universe. Some of these include the matter-antimatter asymmetry, the source of mass for the neutrinos, the existence of dark matter, and reconciling general relativity and quantum mechanics. Many particle physics experiments around the world are being devised and conducted in order to shed light on these questions and improve our understanding of reality. One such experiment is the Fermilab Muon  $g - 2$  Experiment (E989), underway at the Fermi National Accelerator Laboratory (FNAL) located in Batavia, Illinois.

This dissertation will describe the E989 experiment and the author's contributions in detail. Chapter 1 will provide experimental and theoretical background to the experiment, as well its motivation. Chapter 2 will describe the experimental principle and specifics of muon production and storage. Chapter 3 will describe the various detector systems. Chapter 4 will describe the straw tracking reconstruction including the track fitting algorithm, as well as some analysis results. Chapter 5 will describe the precession frequency measurement portion of the experiment, and detail analysis results from data taken in the experiment's first run in 2018. Chapter 6 will conclude the dissertation and the results contained within.

## 1.1 Magnetic moments of particles

In order to understand the purpose of the Fermilab Muon  $g - 2$  Experiment, first we need to understand the meaning of the  $g$  in  $g - 2$ , which is what the experiment is measuring. All particles have intrinsic properties. One property of charged particles is the magnetic dipole moment.<sup>1</sup> This property of a particle is related to its spin through the equation

$$\vec{\mu} = g \frac{q}{2m} \vec{s}, \quad (1.1)$$

where  $\vec{\mu}$  is the magnetic dipole moment of a particle,  $\vec{s}$  is its spin vector,  $m$  is its mass,  $q = \pm e$  where  $e$  is the elementary charge, and  $g$  is the so called “Landé g-factor”.  $g$  is a measurable and predictable constant. Since the torque on a particle in a magnetic field is

$$\vec{N} = \vec{\mu} \times \vec{B}, \quad (1.2)$$

the rate at which a particle’s spin precesses in a magnetic field will depend on  $g$ . This is one of the key physics principles in the E989 experiment as will be discussed later.

In a Dirac theory,  $g$  is equal to 2 for spin- $\frac{1}{2}$  particles with no internal structure [1]. A simple derivation of this result is given in Reference [2]. However, even for these types of particles,  $g$  is not quite equal to 2. Motivated by early experimental measurement discrepancies such as the measurements of the hyperfine structure in hydrogen [3], in 1948 Schwinger calculated the first “radiative correction” to the electron magnetic moment [4]. In a quantum field theory, interactions of the particle with virtual particles in loops will contribute to the value of  $g$ . In this context, for

---

<sup>1</sup>*Magnetic dipole moment* and *magnetic moment* are equivalent when talking about particles.

charged leptons, it is convenient to recast the magnetic dipole moment formula as

$$\vec{\mu} = 2(1+a) \frac{q}{2m} \vec{s},$$

$$a = \frac{g-2}{2}, \quad (1.3)$$

where  $a$  is called the “magnetic anomaly,” and contains all higher order corrections to  $g$ . The first correction to  $a$ , calculated by Schwinger, was  $a = \alpha/2\pi \approx 0.00116$ , where  $\alpha$  is the fine structure constant. By measuring  $a$ , the SM theory can be tested and extensions to it constrained. A precise measurement of the magnetic anomaly, or the anomalous magnetic moment of the muon, is the main goal of the Fermilab Muon  $g - 2$  Experiment.

## 1.2 Standard Model contributions to $a_\mu$

Any complete theory must agree with experimental observations. The latest theoretical predictions for the muon magnetic moment will be presented here. The contributions to  $a_\mu$  can be divided into three SM classes. These include the quantum-electrodynamics (QED) corrections purely from other leptons and photons, the electroweak (EW) corrections from interactions involving the weak force bosons  $W^\pm$  and  $Z^0$ , and the hadronic corrections from interactions with strongly-interacting hadrons:

$$a_\mu^{\text{SM}} = a_\mu^{\text{QED}} + a_\mu^{\text{EW}} + a_\mu^{\text{Had}} \quad (1.4)$$

### 1.2.1 QED

The QED contributions to  $a_\mu$  stem solely from loops with virtual leptons and photons. They are very well understood and have been calculated to very high order, having been calculated up to five-loop level from over 12,000 Feynman diagrams. This has been done with both analytical and numerical methods. The first couple of diagrams,



Figure 1.1: The first of many QED diagrams contributing to  $a$ . B is an external magnetic field. Feynman diagrams made with References [5, 6].

including the Dirac  $g = 2$  and Schwinger diagrams, are shown in Figure 1.1. The value is [7, 8]

$$a_\mu^{\text{QED}} = \sum_{n=1}^{\infty} C_n \left(\frac{\alpha}{\pi}\right)^n, \quad (1.5)$$

$$= (11658471.8971 \pm 0.0007) \times 10^{-10},$$

where in the first line  $a_\mu^{\text{QED}}$  is expressed as a perturbative expansion in the fine structure constant, and the value comes from the expansion up to  $n = 5$ .  $C_1 = 1/2$  is the Schwinger result mentioned previously, stemming from the diagram shown in Figure 1.1b. While over 99% of the value of  $a_\mu$  comes from the QED sector, the error is much smaller than the EW and hadronic contributions.

### 1.2.2 Electroweak

The electroweak contributions to  $a_\mu$  are known to two-loop level, with some three-loop parts estimated. The different one loop diagrams are shown in Figure 1.2. In even the simplest EW diagram propagators from both the leptons and EW gauge bosons will be included. Per usual Feynman rules, the propagators will contain the masses of the interacting leptons and bosons. The heavy masses of the gauge bosons



Figure 1.2: First order weak diagrams contributing to  $a_\mu$ . B is an external magnetic field. Feynman diagrams made with References [5, 6].

will produce contributions with characteristic scales of  $\sim (m_l/m_{Z^0,H,W^\pm})^2$ . Because the mass of the gauge bosons are so much larger than the muon, these processes are necessarily suppressed and the electroweak contributions to  $a_\mu$  are small. For this reason knowing these contributions only up to two loop level is sufficient, since higher order diagrams will be suppressed even further. To this degree of accuracy, the value of the electroweak contributions is [9, 10]

$$a_\mu^{\text{EW}} = (15.12 \pm 0.01) \times 10^{-10}, \quad (1.6)$$

with improvements having been made recently. Again the error on these contributions is small compared to the that on the hadronic contributions discussed next.

### 1.2.3 Hadronic

The hadronic contributions to  $a_\mu$  arise from loop diagrams with virtual hadrons. Because QCD couplings are large at low energies, the amplitudes for these processes cannot be calculated perturbatively and their errors dominate the theoretical uncertainty in the SM calculation. This makes their error estimation extra important when comparing to experiment. Most active work on  $a_\mu$  in the theoretical community is in

this sector. These contributions can be separated into two parts:

$$a_\mu^{\text{Had}} = a_\mu^{\text{HVP}} + a_\mu^{\text{HLbL}} \quad (1.7)$$

### Hadronic Vacuum Polarization

The first term in Equation 1.7 refers to contributions from hadronic vacuum polarization (HVP), the first order diagram of which is shown in Figure 1.3a. There are two main prescriptions for calculating these contributions. The first is to use a dispersive approach to introduce a virtual hadron bubble into the integral calculation for the photon propagator<sup>2</sup>, and then utilize the optical theorem to relate the imaginary part of that propagator to the total cross-section of electron-positron annihilation to hadrons [12]. While this could be solved perturbatively for a lepton bubble in place of the hadron bubble, this is instead a data driven approach when considering non-perturbative QCD. The leading order contribution can be written as

$$a_\mu^{\text{HVP;LO}} = \left( \frac{\alpha m_\mu}{3\pi} \right)^2 \int_{m_\pi^2}^{\infty} \frac{ds}{s^2} K(s) R(s) \quad (1.8)$$

where  $K(s)$  is some calculable kinematic factor, and  $R(s)$  is a ratio of cross-sections

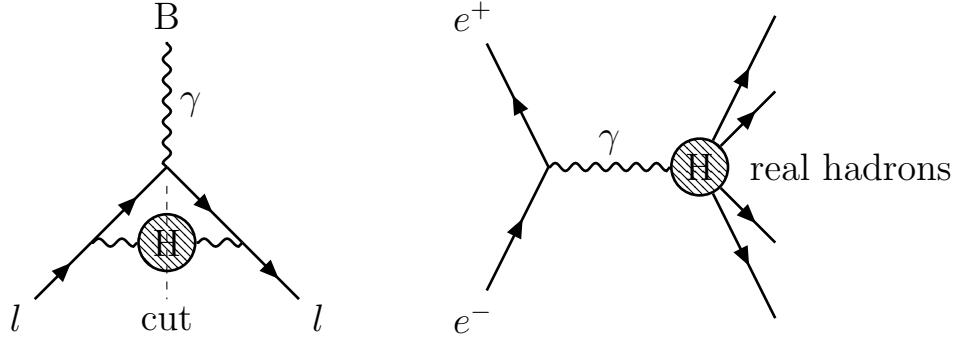
$$R(s) = \frac{\sigma(e^+e^- \rightarrow \text{hadrons})}{\sigma(e^+e^- \rightarrow \mu^+\mu^-)}. \quad (1.9)$$

The cross-section data for this relation has been measured in different energy ranges by various experiments, including KLOE, CLEO, BaBar, and BESIII [13, 14, 15, 16]. The analysis by Keshavarzi et al. [17] gives results as

$$\begin{aligned} a_\mu^{\text{HVP;LO}} &= (693.26 \pm 2.46) \times 10^{-10}, \\ a_\mu^{\text{HVP;NLO}} &= (-9.82 \pm 0.04) \times 10^{-10}, \end{aligned} \quad (1.10)$$

---

<sup>2</sup>The details of dispersion theory will not be described here but a pedagogical introduction is given in Reference [11].



(a) The first order HVP diagram. The bubble H in the middle indicates any combination of hadrons which satisfy the Feynman rules.

(b) The Feynman diagram for electron positron annihilation to hadrons.

Figure 1.3: The first order HVP diagram on the left, which can be related to the diagram on the right by making a ‘cut’ across the virtual hadrons bubble. B is an external magnetic field. Feynman diagrams made with References [5, 6].

where  $a_\mu^{\text{HVP;NLO}}$  is the next-to-leading order calculation. This evaluation is consistent with Davier et al. [18].

The second prescription for estimating the HVP contributions is a first principles approach, using lattice QCD and QED. Lattice QFT is a gauge theory defined on a matrix of points in time and space. In the limit that the matrix is taken infinitely large with the spacing between the points infinitely small, the behavior from a continuous theory is recovered. The lattice-based estimates of  $a_\mu^{\text{HVP;LO}}$  are consistent with those provided above, though the error is larger [19]. If the calculation is supplemented with the cross-section data described above, then this method provides the most precise determination of  $a_\mu^{\text{HVP;LO}}$  as shown in Figure 1.4.

### Hadronic Light-by-Light

The second of these hadronic contribution parts is a higher-order four-photon interaction, termed hadronic light-by-light (HLbL). Diagrams are shown in Figure 1.5. Again perturbation theory is unable to assist in the calculation of these contribu-



Figure 1.4: HVP calculations for  $a_\mu$  from various groups. The green points are lattice calculations, the purple points are dispersive results, and the orange point is a combination approach [19].

tions. The calculation of these diagrams has historically been model dependent, and has therefore been the most contentious part of the SM calculation. In more recent years, there have been efforts to produce results using dispersive [20, 21, 22] and lattice approaches [23, 24, 25]. A cumulative report on recent progress is given by N. Asmussen et al. [26]. The error on this contribution is large in each of these various approaches<sup>3</sup>, comparable to that of the  $a_\mu^{\text{HVP;LO}}$  term, even though the size of this contribution is small. The value of the HLbL contributions to  $a_\mu$  given by Nyffeler [27] comes from model estimates and is

$$a_\mu^{\text{HLbL}} = (9.8 \pm 2.6) \times 10^{-10}. \quad (1.11)$$

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<sup>3</sup>A range of credible models produces a range of results.



(a) The first HLbL diagram, where three photons are exchanged with some virtual hadrons bubble.

(b) A second HLbL diagram, where three photons are exchanged with two virtual hadrons bubbles, that are connected with some virtual charge-less propagator.

Figure 1.5: HLbL diagrams contributing to  $a_\mu$ . B is an external magnetic field. Feynman diagrams made with References [5, 6].

#### 1.2.4 Combined Standard Model value

The sum of the  $a_\mu$  contributions listed here is [7, 8, 9, 10, 17, 27]

$$\begin{aligned} a_\mu^{\text{SM}} &= a_\mu^{\text{QED}} + a_\mu^{\text{EW}} + a_\mu^{\text{Had}}, \\ &= (11659180.26 \pm 3.58) \times 10^{-10}. \end{aligned} \quad (1.12)$$

The relative uncertainty of this result is 307 parts per billion (ppb). Other analyses with different values for the various contributions typically agree well, as shown on the left side of Figure 1.6. In general the consistency of the theory has been stable for almost ten years now. Depending on which calculations are used, the discrepancy between theory and experiment ranges between 3 to 4 standard deviations. The latest experimental result is described in the following section.



Figure 1.6: Various theoretical values for  $a_\mu$  on the left, as compared to the most recent and extrapolated experimental result on the right [17].

### 1.3 Experimental value of $a_\mu$ and discrepancy with $a_\mu^{\text{SM}}$

The theoretical contributions to  $a_\mu$  listed in the previous sections have improved over time as methods have matured and more experimental data has been gathered. Similarly, work on the direct experimental measurement of  $a_\mu$  has been going on for decades, with more precise results being determined over time [28]. The most recent experiment to measure  $g - 2$  was the Brookhaven Muon  $g - 2$  Experiment (E821) held at Brookhaven National Laboratory (BNL), which collected data from 1997 to 2001. That experiment measured a value for  $a_\mu$  of [29, 30]

$$a_\mu^{\text{Exp}} = (11659208.9 \pm 6.3) \times 10^{-10}, \quad (1.13)$$

which corresponds to a 540 ppb relative uncertainty. Note that the uncertainty of the experimental measurement is comparable to that of the theory, necessitating a precise understanding of all the different theoretical parts. The difference between the experimental and theoretical values presented here is

$$a_\mu^{\text{Exp}} - a_\mu^{\text{SM}} = (28.64 \pm 7.25) \times 10^{-10}, \quad (1.14)$$

corresponding to a discrepancy of nearly four standard deviations from zero.

### 1.4 Beyond the Standard Model and the purpose of E989

While the discrepancy between experiment and theory might be attributed to miscalculations in the theory or systematic errors in the E821 experiment, no such errors have been found despite repeated attempts to resolve the two. Indeed the discrepancy has only grown over time as the theoretical calculations have matured. The most intriguing and exciting source of the discrepancy would be physics beyond the standard model (BSM), Figure 1.7. Since the value of  $a_\mu$  receives contributions from

all particles that couple to the muon through virtual loops, unknown particles might be the source of this discrepancy. Specifically, since the contribution to the magnetic moment from heavy virtual particles goes as

$$a \sim \frac{m^2}{\Lambda^2}, \quad (1.15)$$

where  $\Lambda$  is the mass scale of the new physics and  $m$  the mass of the lepton in question, the sensitivity of the muon as compared to the electron to large mass scales is  $m_\mu^2/m_e^2 \approx 43,000$  times greater. For this reason, it is possible that even though the g-factor of the electron has been measured extraordinarily precisely, to .26 parts per trillion (ppt) corresponding to a .23 ppb relative uncertainty in  $a_e$  [31, 30], it has not yielded a definitive difference between theory and experiment, whereas the magnetic moment of the muon might do so<sup>4</sup>.

Because the discrepancy from the previous experiment was not at the five standard deviation level necessary to classify it as a discovery, the E989 experiment was undertaken. Indeed with the lack of new physics results coming out of most other experiments, E989 is especially positioned to uncover something new at a time when there are so few hints of new physics. Because of this, the interest in the E821 experiment and its successor E989 has grown over time. The number of citations for the E821 results has been consistently high for 20 years as shown in Figure 1.8.

The E989 experiment has the goal of measuring  $a_\mu$  to 140 ppb over the course of several years. This would be a factor of four improvement over the E821 result stemming from a twenty times increase in statistics, which was dominate error in the previous experiment. Assuming the same central value for  $a_\mu$  is obtained, this would push the statistical significance of the discrepancy over to approximately seven

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<sup>4</sup>A recent measurement of the fine structure constant  $\alpha$  to the highest precision yet [32] has yielded a new value of  $a_e$  for which the difference between theory and experiment is 2.5 standard deviations, but with the opposite sign as compared to  $a_\mu$ .



Figure 1·7: A general Feynman diagram, where the leptons couple to an external magnetic field  $B$  through some BSM physics. Feynman diagrams made with References [5, 6].



Figure 1·8: The number of citations for the BNL experiment E821 publications as a function of year, as of 10/30/2019 [33].

standard deviations, as shown in Figure 1·6. The data comprising Run 1, gathered between April and July of 2018, is the subject of this thesis, and corresponds to a statistical uncertainty comparable to the E821 result. The possibility is that we will measure something new and exciting, either in Run 1 or the following runs. However, even if we do not, it is valuable in itself to resolve this theoretical and experimental conflict.

## Chapter 2

# Principal Techniques of E989

A particle with non-zero spin in a magnetic field will experience a torque which attempts to line up the magnetic dipole moment of the particle with the external field. Because of this, in a dipole field a particle's spin will turn at the spin precession frequency [34]

$$\vec{\omega}_s = -g \frac{q}{2m} \vec{B} - (1 - \gamma) \frac{q}{\gamma m} \vec{B}, \quad (2.1)$$

where  $m$  is the particle's mass,  $q = \pm e$  where  $e$  is the positive elementary charge,  $g$  is the g-factor,  $\gamma$  is the Lorentz relativistic factor, and  $B$  is an external magnetic field. The first term is the usual Larmor frequency and the second term is a relativistic correction to the precession frequency in an accelerating frame called Thomas precession. Similarly, a particle with some momentum perpendicular to the magnetic field will orbit at the cyclotron frequency

$$\vec{\omega}_c = -\frac{q}{\gamma m} \vec{B}. \quad (2.2)$$

By taking the difference between these two frequencies we arrive at the “spin difference frequency”

$$\vec{\omega}_a = \vec{\omega}_s - \vec{\omega}_c = -\frac{g-2}{2} \frac{q}{m} \vec{B} = -a \frac{q}{m} \vec{B}, \quad (2.3)$$

a frequency that is directly proportional to the anomaly  $a$ . If  $g = 2$ , as in a Dirac theory, then the particle's spin would turn at the same rate as the momentum vector, and  $\omega_a$  would be identically zero. If  $\omega_a$  for a muon and the external magnetic dipole field can be measured, then the anomalous magnetic moment of the muon  $a_\mu$  can be determined.

As will be detailed in Section 2.2, the measurement of the magnetic field is related to the Larmor precession frequency of free protons in water

$$\omega_p = -g_p \frac{e}{2m_p} B, \quad (2.4)$$

where  $g_p$  and  $m_p$  are the g-factor and mass of the proton respectively. Replacing  $B$  and solving for  $a_\mu$ , we arrive at

$$a_\mu = \frac{g_p}{2} \frac{\omega_a}{\omega_p} \frac{m_\mu}{m_p}. \quad (2.5)$$

Using the magnetic moment formulae for the proton, electron, and muon as shown in Equation 1.1, Equation 2.5 can be transformed into

$$a_\mu = \frac{g_e}{2} \frac{\omega_a}{\omega_p} \frac{m_\mu}{m_e} \frac{\mu_p}{\mu_e}, \quad (2.6)$$

where the  $p$ ,  $e$ , and  $\mu$  subscripts stand for the relevant quantities for the proton, electron, and muon respectively. The experimental error on  $a_\mu$  then becomes the quadrature sum of each individual quantity error. As mentioned in Section 1.4 the electron g-factor  $g_e$  has been measured to extremely high precision, 0.26 ppt [30, 31]. The muon-electron mass ratio,  $m_\mu/m_e$ , has been measured to 22 ppb [30, 35]. Finally the proton-electron magnetic moment ratio,  $\mu_p/\mu_e$ , has been measured to 3 ppb [30]. These are small compared to the target statistical error on  $\omega_a$  of 100 ppb, and target systematic errors on  $\omega_a$  and  $\omega_p$ , both at 70 ppb<sup>1</sup>. These errors added in quadrature

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<sup>1</sup>The measurement of  $\omega_p$  has negligible statistical error.

is approximately 140 ppb, which is the target of the E989 experiment.

## 2.1 Measuring $\omega_a$

How can  $\omega_a$  for muons be measured? The answer lies with two key points in the dynamics of muon decay. Positive muons decay to a positron and two neutrinos, as shown in Figure 2·1. The first point is that because of the parity violating nature of the weak interaction, the decay positron will be preferentially emitted right-handed, with its spin directed in the same direction as its momentum [36]. The second key point is that angular momentum must be conserved. Consider the most extreme examples of maximum and minimum energy positrons as shown in Figure 2·2. In the muon rest frame, decay positrons with maximum energy will be emitted opposite to the two neutrinos, both emitted in the same direction. Since neutrinos and anti-neutrinos must be left and right-handed respectively, thus having their spins anti-parallel and parallel to their momentum, by the law of conservation of angular momentum the positron must have its spin be parallel to the spin of the muon at the time of the decay. By the opposite argument, decay positrons emitted with minimum energy such that the neutrinos are ejected opposite to one another must have their spins be anti-parallel to that of the muon at the time of decay. Together, these two points mean that higher energy decay positrons will preferentially be emitted in directions parallel to the muon spin at the time of decay, while lower energy decay positrons will preferentially be emitted in directions anti-parallel to the muon spin at the time of the decay.

This correlation between the direction of an emitted high energy decay positron and the spin of the muon is the signature needed to measure  $\omega_a$ . Muons placed within a magnetic storage ring will orbit at the cyclotron frequency and their spins will precess at the spin precession frequency. As they go around the ring they will



Figure 2·1: Feynman diagram for muon decay.  $\mu^+$  decay through a  $W^+$  boson to a positron, muon anti-neutrino, and an electron neutrino. This process consists of nearly 100% of the muon decay branching ratio, with other decay states including radiative processes.



Figure 2·2: Muon decay in the rest frame for maximum (top) and minimum (bottom) energy decay positrons. Due to the conservation of angular momentum and the single possible helicity states of the decay neutrinos, the spin of the decay positron is exactly parallel to the spin of the muon at the time of the decay for maximum energy decay positrons, or anti-parallel for minimum energy decay positrons.

decay to positrons whose energy and decay directions contain information about the spin of the muon. If the muon ensemble is un-polarized, then the decay distribution for all positrons will be isotropic and in general un-useful. If the muon ensemble is polarized, then the decay distribution for each muon will be the same, and allow for a measurement of  $\omega_a$ , as described in the following text.

The differential decay distribution for positive muons in the muon rest frame is described by [36]

$$dP(y, \theta) \propto N(y)[1 + A(y)\cos(\theta)]dyd\Omega, \quad (2.7)$$

where  $y = E/E_{max}$  is the energy fraction of the positron and  $\theta$  is the angle between the spin of the muon and the momentum of the positron at the time of decay.  $N(y)$  is the number distribution of decay positrons and  $A(y)$  is the so called ‘asymmetry,’ encoding the energy-dependent correlation between the muon spin and the decay positron direction. Here the energy of the positron is assumed to be much greater than its mass. The number distribution and asymmetry are given by [36]

$$N(y) = 2y^2(3 - 2y), \quad (2.8)$$

$$A(y) = \frac{2y - 1}{3 - 2y}, \quad (2.9)$$

and are shown in Figure 2·3a.

In the lab frame, nearly all high energy positrons are emitted parallel to the muon momentum, making it challenging to select purely on the decay angle of the positron. That is not a problem however, as we already know that decay positrons with the highest rest-frame energies will be emitted parallel to the muon spin at the time of decay. Essentially, the energy distribution of detected high energy positrons is modulated by  $\omega_a$ , or  $\theta = \omega_a t + \phi$ . The number of detected positrons at some time and energy in the lab frame for some initial number  $N_0$  of muons can then be described



Figure 2.3: Decay number distribution  $N$  and asymmetry  $A$  in the muon rest frame (left) and in the lab frame (right) as a function of positron energy with a maximum positron energy of 3.1 GeV.  $N$  is multiplied by arbitrary factors in both pictures.

by

$$N_d(t, E) = N_0(E) \cdot e^{-t/\gamma\tau_\mu} \cdot [1 + A(E) \cos(\omega_a t + \phi(E))], \quad (2.10)$$

where the  $d$  subscript stands for ‘detected,’ the muons are decaying with a lifetime of  $\gamma\tau_\mu$ , and all the relevant parameters are energy dependent. Here  $N_0(E)$  and  $A(E)$  have been transformed from Equations 2.8 and 2.9 to the lab frame:

$$N_0(E) \propto (y - 1)(4y^2 - 5y - 5), \quad (2.11)$$

$$A(E) = \frac{-8y^2 + y + 1}{4y^2 - 5y - 5}, \quad (2.12)$$

where as a reminder  $y = E/E_{max}$ . Here the polarization of the muons is assumed to be unity. These relations are shown in Figure 2.3b. All positrons above some energy threshold cut  $E_{th}$  can be taken as the observable,

$$N_d(t, E_{th}) = N_0(E_{th}) \cdot e^{-t/\gamma\tau_\mu} \cdot [1 + A(E_{th}) \cos(\omega_a t + \phi(E_{th}))], \quad (2.13)$$

where the number and asymmetry of the detected positrons is now calculated by



Figure 2.4: The number of detected positrons above some energy threshold ( $y \sim 0.55$ ) as a function of time, where the data plotted come from Run 1 and correspond to nearly  $1 \times 10^9$  counts. The time axis is wrapped around every  $100\text{ }\mu\text{s}$ . See Chapter 5 for the fitting of such a histogram.

simply integrating Equations 2.11 and 2.12 from  $y_{th}$  to 1,

$$N_0(E_{th}) \propto (y_{th} - 1)^2(-y_{th}^2 + y_{th} + 3), \quad (2.14)$$

$$A(E_{th}) = \frac{y_{th}(2y_{th} + 1)}{-y_{th}^2 + y_{th} + 3}, \quad (2.15)$$

where  $y_{th} = E_{th}/E_{max}$ . By counting decay positrons above some energy threshold and fitting the resulting time spectrum with Equation 2.13,  $\omega_a$  can be extracted. A sample of data adhering to such a time spectrum is shown in Figure 2.4.

The statistical error on the  $\omega_a$  measurement, assuming bin errors are Gaussian and a  $\chi^2$  minimization is used with the fit function described in Equation 2.13, is [37]

$$\frac{\sigma_{\omega_a}}{\omega_a} = \frac{\sqrt{2}}{\sqrt{N_{\text{total}}} A \gamma \tau_\mu \omega_a}, \quad (2.16)$$

where  $N_{\text{total}}$  is the total number of counts included in the above-threshold time spectrum. This equation assumes a weighting of one for every count included in the fitted time spectrum. Other weighting schemes exist which slightly improve the statistical precision of the  $\omega_a$  measurement [37], but they are not used in this analysis. What Equation 2.16 reveals is that the statistical precision of  $\omega_a$  is maximized when the quantity  $NA^2$  is at a maximum. It was found for E989 that the optimal energy threshold was about 1.7 GeV as shown in Figure 5·5, which includes detector acceptance effects and corresponds to an asymmetry of about  $A = 0.37$ . Equation 2.16 can be rearranged in order to solve for the number of positrons needed to be collected above threshold for a specific precision goal. For a statistical error of 100 ppb on  $\omega_a$ , the required number of positrons above threshold is approximately  $170 \times 10^9$ , determined from the values  $A = 0.37$ ,  $\gamma = 29.3$  (described later),  $\tau_\mu = 2.2 \mu\text{s}$ , and  $\omega_a = 1.44 \text{ rad}/\mu\text{s}$ . This statistical error of  $\omega_a$  combined with the systematic uncertainties given in Table 5.9, provide the total error on  $\omega_a$ .

## 2.2 Measuring the magnetic field

In order to measure the magnetic moment of the muon to 140 ppb, the field needs to be both highly uniform, and measured to extreme precision. The E989 goal for the field measurement is 70 ppb. As shown in Equation 2.6, the measurement of the magnetic field has equal weight to that of the precession frequency. A cross-section of the magnetic storage ring used in E989 is shown in Figure 2·5. It is an approximately 14 m diameter C magnet, where the muons are stored within a 4.5 cm radius cylindrical storage region at the center of a 1.451 T magnetic field. This corresponds to an approximately  $0.28 \text{ m}^3$  or  $10 \text{ ft}^3$  total volume around the inside of the ring. The magnetic field is made uniform by manipulating many magnetic ‘knobs’ built into the  $g - 2$  storage ring, including the main magnet current, pole



Figure 2·5: Cross-section of the  $g - 2$  magnet. The muons live in the 4.5 cm radius circular storage region shown in red. The main magnetic field is excited by superconducting coils shown in yellow. Pole pieces, wedges, edge shims, top hats, and other magnetic knobs allow for sub-ppm level tuning of the magnetic field.

pieces, wedges, top hats, and thousands of small magnetic shims placed around the storage region, as shown in Figure 2·5. There is also an active feedback system which stabilizes the average magnetic field over time using current carrying coils near the storage region. After a several month shimming campaign by many members of the field team, a precision on the magnetic field of approximately 25 ppm RMS (root mean square) was achieved.

The magnetic field is measured using a nuclear magnetic resonance (NMR) technique, hence the measurement on  $\omega_p$  as shown in Equation 2.4. NMR was chosen as it provides a field measurement precision on the order of 10 ppb with negligible statistical uncertainty [38]. NMR probes consist of pickup coils located around a sample of protons in some fluid, typically water or petroleum jelly. The pickup coils



Figure 2·6: An example FID signal. The current picked up in the coils around the proton sample will oscillate as the spins precess around the main magnetic field, and decay as the spins return to alignment with the external field.

deliver a “ $\pi/2$  pulse” which rotates the proton sample magnetization  $90^\circ$  out of phase from equilibrium. The proton spins will begin precessing at the Larmor frequency ( $\approx 61.79$  MHz). As the spins interact with local magnetic field gradients and inhomogeneities, the magnetization of the proton sample will relax back to equilibrium with the external field, typically on the order of several milliseconds. This so-called free-induction decay signal (FID), an example of which is shown in Figure 2·6, is measured using the same pickup coils that delivered the initial  $\pi/2$  pulse

It is not solely  $\omega_p$  that needs to be measured. What really matters is the average magnetic field that the muons see, namely the time-averaged spatially-weighted magnetic field. The scheme devised to measure this is two-fold. First, the magnetic field in the muon storage region is measured by a trolley which travels around the inside of the ring. This trolley holds 17 NMR probes which measure the field at approximately 6000 locations around the inside of the ring. However, because the trolley cannot be in the storage region when the muons are present in the ring, during data taking it



Figure 2·7: A sample of the azimuthally-averaged magnetic field within the storage region [39]. The contours are normalized to the field value at the center of the storage region. The scale of the field differences is approximately  $\pm 1.5$  ppm. The black dots in the picture correspond to the location of the trolley probes.

is retracted and the field is instead monitored by 378 fixed NMR probes located in the high magnetic field region, just outside the storage region on the outside of the vacuum chambers. The prescription is that the storage ring field is measured every few days by the trolley probes, and the field between trolley runs is interpolated using measurements from the continually-sampling fixed probes. In this way the magnetic field can be mapped over time and over the space in which the muons are stored. A sample of the azimuthally-averaged magnetic field measured with trolley and fixed probes is shown in Figure 2·7.

Lastly, it is the free proton precession frequency in the field that is of interest, but the frequency that the trolley probes measure will be different due to the molecular

properties of the proton sample as well as the material properties of the probe itself. The frequency of the induction signal that the probes measure can be re-cast as

$$\omega_{p,\text{probe}} = \omega_{p,\text{free}}(1 - \sigma(\text{H}_2\text{O}, T) + \delta_b + \delta_p + \delta_s), \quad (2.17)$$

where  $\sigma(\text{H}_2\text{O}, T)$  is the temperature-dependent diamagnetic shielding of protons in a water molecule, and the  $\delta$ 's come from corrections due to the bulk susceptibility of the water sample, paramagnetic impurities in the water sample, and the magnetic effects of the probe itself, respectively [38]. In order to correct for these effects two additional special probes are used, both of which live in a single section of the ring which has been shimmed to extra uniformity. The first is a calibration probe which measures the free proton precession frequency at the center of the storage region, corresponding to the position of the central trolley probe. The calibration probe is made of materials which are designed to reduce the effects listed in Equation 2.17, and has been characterized in a dedicated highly uniform solenoidal magnetic field. The second special probe is called the ‘plunging probe.’ This probe is located inside the vacuum chamber and moves into the storage region to measure the field at each of the 17 trolley probe locations, using a three dimensional motion system. By using these two probes, the calibration to the free proton precession frequency can be transmitted to each of the trolley probes, providing an absolute scale for the measurements inside the muon storage region. The uncertainty on the calibration procedure is estimated to be 35 ppb, about half the total permitted by the experimental error budget of 70 ppb.

Other pieces of the systematic uncertainty include the calibrations of the probes, errors in the trolley measurements, the interpolation with the fixed probes, the uncertainty of the magnetic field relative to the muon distribution, and others such as time dependent external magnetic fields, Table 2.1.

<b>Magnetic Field Measurement Uncertainties</b>	
Source of uncertainty	E989 Goal (ppb)
Absolute calibration of standard probe	35
Calibration of trolley probes	30
Trolley measurements	30
Fixed probe interpolation	30
Muon distribution weighted average	10
Time dependent external fields	5
Others	30
Quadrature sum	70

Table 2.1: Systematic errors in the magnetic field measurement. Unlisted sources of error include the measurement of higher field multipoles, trolley temperature and power supply voltage response effects, and eddy currents from the kicker, among others.

### 2.3 Production of polarized muons

As explained in Section 2.1, in order to measure  $\omega_a$  the muons injected into the storage ring must be polarized. The BNL E821 experiment observed on the order of 10 billion positrons above threshold, and its final result was statistics limited. In order to reach the goal of 140 ppb, approximately 20 times that number of positrons needs to be gathered, and hence a large number of polarized muons produced. Fermilab has the facilities to produce such a large number of polarized muons.

Polarized muon beams are constructed using the physics of pion decay. Using the same parity-violation and spin momentum conservation logic as explained in the discussion of muon decay, it is determined that pion decay produces muons that are 100% polarized in the pion rest frame, due to the pion having zero spin. It is also important to note that pions decay to muons with a  $\sim 99.98\%$  branching ratio due to the parity violating nature of the weak interaction [40]. Thus by producing a large number of polarized pions a large number of polarized muons can be acquired.



Figure 2.8: The layout of accelerator beam-line components Fermilab uses to provide polarized muons to E989. Protons start in the Linac, traverse around the Booster and then Recycler, and are converted to pions at AP0. The pions are gathered and then decay to muons in the Delivery Ring before being sent to the  $g - 2$  storage ring. [38].



Figure 2.9: General timing structure of beam pulses sent to E989.

The production of polarized muons for E989 at the Fermilab accelerator complex involves a number of stages. A map of the various relevant accelerator beam-line components is shown in Figure 2.8. Details of the full accelerator production of polarized muons can be found in Reference [41], and here a summary of the process will be given. First,  $H^-$  ions are produced and accelerated in a linear accelerator. They are stripped down to protons and then transported to a 75 m radius circular storage ring called the “booster,” which accelerates them up to 8 GeV/c and batches them together. A single booster batch contains on the order of  $4 \times 10^{12}$  protons. The protons are then injected into a ring called “recycler,” which separates them into four separate bunches of  $1 \times 10^{12}$  protons, each with a time width of approximately 120 ns. (This is less than the cyclotron period of the storage ring of 149 ns.) This re-bunching process is done in order to reduce the instantaneous rates observed by the detectors, and thus reduce the level of pileup in the E989 detectors; see Section 3.2.1. For a single accelerator supercycle of 1.4 s, E989 receives four booster batches corresponding to sixteen recycler bunches at an average rate of 11.4 Hz, with the time separation between bunches greater than 10 ms. The timing structure is shown in Figure 2.9, where trains of eight bunches are sent to E989 with some time separation between them<sup>2</sup>. Although it remains relatively constant, this timing structure may be modified in response to the requirements of other experiments.

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<sup>2</sup>These eight bunches are sometimes referred to as pulses, and the gathered data is tagged by which bunch or pulse it originates from.

Each bunch is sent to a target hall, where it is directed on to an Inconel target. This Inconel target is made of a nickel-iron alloy, and is optimized for the production of a large number of pions with a small momentum spread, approximately  $1 \times 10^{-5} \pi^+$  per proton on target with  $|dp/p| < 2\%$  [41]. The pions are focused by the magnetic field of a lithium lens, located just downstream of the production target. The lens, a cylinder of lithium 1 cm in radius and 15 cm long, carries a large current down its length which provides a radial focusing effect for particles passing down the cylinder [42]. A pulsed magnet just after the lithium lens is then used to select pions and residual particles, mainly protons, centered at 3.115 GeV.

The resultant pion beam and these residual particles are then injected into another ring called the “delivery ring”<sup>3</sup>. Here it should be noted that in a pion beam the highest and lowest energy decay muons are forwards and backwards polarized, respectively. The delivery ring is used to momentum select polarized muons and separate out the non-desired particles which will have differing velocities compared to the muons, reducing the contamination in the final polarized muon beam [41]. Forward emitted polarized muons are momentum selected at their decay energies of 3.094 GeV with  $\Delta p/p = 2\%$ . Particles that fall outside the momentum acceptance are lost, and the residual protons are kicked into a beam dump. The polarized muon beam is then sent to the E989 building where it passes through four magnetic quadrupole focusing magnets before being injected into the E989 storage ring.

## 2.4 Injection of muons

The injection of the muon beam into the E989 storage ring is a specialized process. In order to measure the magnetic field to the precision described in Section 2.2, the storage ring must be a single monolithic magnet with no end effects. This prohibits

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<sup>3</sup>By the time the pions have gotten to the delivery ring, most of them have decayed to muons.

the usual design of separated magnetic elements through which the muons might be injected. Therefore the beam must be injected through the storage ring magnet yoke. This introduces the design constraint where the main magnet field must be eliminated within the injection tunnel, such that muons are not lost due to deflection into the magnet itself. Any solution to this problem must have it's cancellation field localized to the injection tunnel, such that there is no residual fringe field that contaminates the main storage ring dipole field.

A specialized magnet called the “Superconducting Inflector” magnet, or just inflector, is used to solve these design constraints. This inflector is placed just after a bored out tunnel in the storage ring magnet yoke, on the inside of the C shape. See Figure 2·10 for a view of the injection point. The inflector has an 18 mm wide by 56 mm high aperture through which the muons must pass down its 1.7 m length. The inflector is made up of superconducting coils wrapped in a truncated double cosine theta design around an aluminum mandrel [43]. See Figure 2·11. This design serves to contain the majority of the inflector magnet field, while eliminating the the storage ring field for the muons passing down its length. The inflector is contained within a superconducting shield which traps the fringe field of the inflector such that the storage ring magnet field is unaffected. As shown in Figure 2·11, windings cover both ends of the inflector such that an appreciable fraction of muons are lost due before being injected into the ring. In order to increase the muon flux for future runs of E989, a new inflector magnet is being designed with open ends [38].

## 2.5 Storage of muons

Once the muons have been injected into the ring, they will begin orbiting clockwise around the ring, decaying with a lifetime of 64.4  $\mu\text{s}$ . The E989 experiment and storage ring are shown in Figure 2·12. By necessity, the inflector must be out of the stored



Figure 2.10: A plan view of the inflector and injection point into the storage ring [43].



Figure 2.11: The inflector magnet (left) and a cross section view of the inflector windings and associated shield (right) [43].

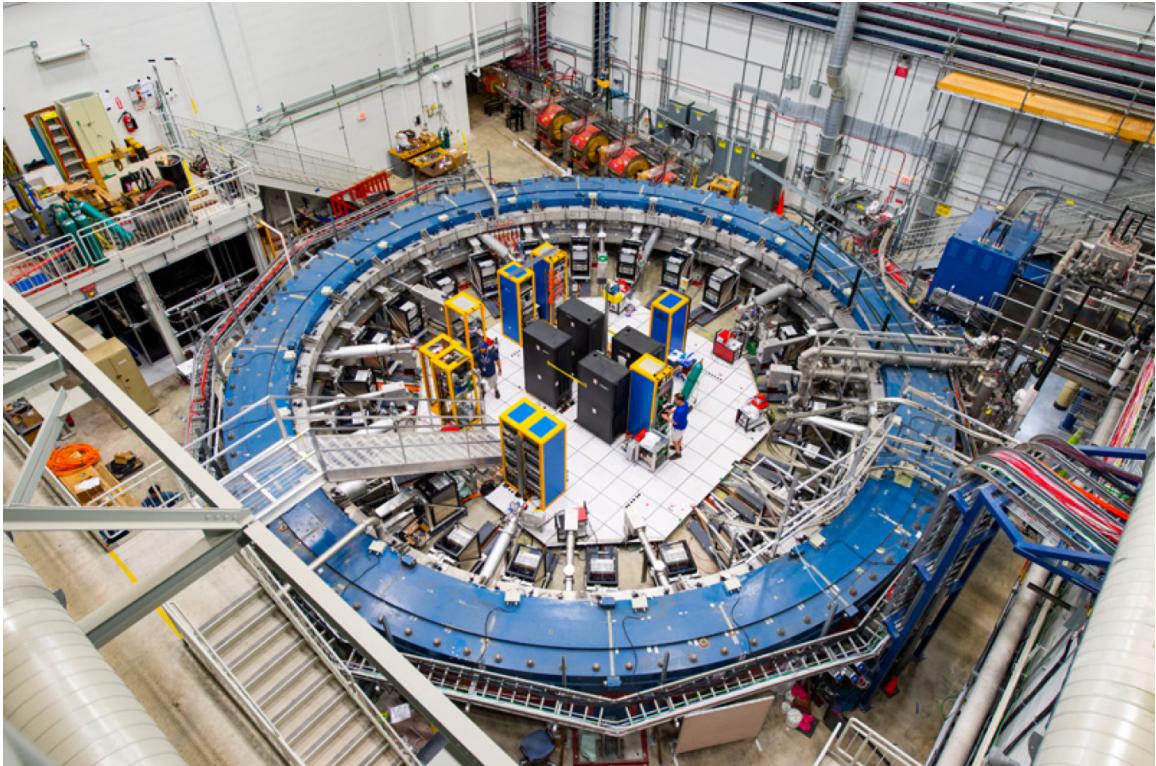


Figure 2.12: The E989 experiment. The blue storage ring can be seen to surround a variety of detectors and electronics. Muons from the accelerator enter at the top of the picture, and are transported to the storage ring through a series of magnetic quadrupoles. Muons are injected through the inflector into the ring, where they orbit in a clockwise direction. The ring is approximately 16 m in diameter to the outside edge of the ring, and about 3 m high. People in the center of the ring give a sense of scale to the picture.

muon beam path, otherwise a large fraction of the muons would be lost upon the return to the injection point as the muons would strike the inflector. Therefore the muon beam must be kicked to move the beam path from the injection orbit onto the central orbit of the storage ring. The muon beam must also be contained vertically. To perform the former, a magnetic “kicker” is used to shift the orbits of the muons. To perform the latter, a series of electrostatic quadrupoles focus the beam vertically. Approximately 2% of the injected muons are stored with  $\Delta p/p = 0.1\%$  centered around 3.094 GeV/c, corresponding to a design goal of  $\mathcal{O}(10,000)$  stored muons per fill.

### 2.5.1 Kicker

The kicker is made up of three separate pulsed magnets located 90° from the exit of the inflector, where the injection orbit crosses the central orbit. The placement of the kickers is shown in Figure 2.13. The kicker must be located within the precision magnetic field of the ring, and must therefore contain no magnetic elements in the hardware itself. For this reason each kicker magnet is made up of two thin 1.27 m long aluminum plates, separated by 10 cm, which carry the current used to create the kicking magnetic field. Due to the bunched nature of the muon beam and the short cyclotron period of 149 ns, ideally the kicker moves all injected muons onto the central orbit and then turns off quickly such that by the time the muons orbit back around to the kicker there is no residual kick to the beam. Any residual eddy currents must die away quickly enough such that the magnetic field seen by the stored muons is unperturbed. The design deflection of the beam is approximately 10 mrad using a vertical pulsed field of around 300 Gauss (corresponding to kicker plate voltages of  $\mathcal{O}(155 \text{ kV})$ ) with a pulse length of about 120 ns [38]. The operational kicker performance in Run 1 was less, as described in Section 2.8.



Figure 2·13: A map of the vacuum chambers in E989. K1-K3 show the locations of the kicker magnets, while Q1-Q4 show the locations of the electrostatic quadrupoles. Also shown is the location of the inflector, the two fiber monitors, and one of the tracker stations.

### 2.5.2 Electrostatic quadrupoles

There are four electrostatic quadrupoles located around the ring as shown in Figure 2.13, which provide vertical focusing for the beam. Although the quadrupoles are defocussing in the horizontal plane, the main dipole magnetic field provides net horizontal focusing. Just as with the kickers, the quadrupoles must be operated in vacuum. E989 uses electrostatic focusing elements instead of magnetic ones in order to avoid magnetic field gradients which would limit the precision of the magnetic field measurement. The quadrupoles occupy 43% of the ring circumference, with four quads having been chosen in order to maximize the symmetry of the beam motion, reduce its amplitude, and leave space for other experimental elements around the ring [38]. Each quadrupole is made up of two segments, a short segment of 13° and a long segment of 26° corresponding to 1.61 m and 2.62 m respectively, with each segment consisting of four plates. To minimize multiple scattering of incident decay positrons, the quadrupoles are made as thin as possible. A picture of the quadrupoles installed into one of the vacuum chambers is shown in Figure 2.14. A calculation of the equipotential lines of the quadrupoles is shown in Figure 2.15. The original design of the quadrupoles is detailed in Reference [44].

Some of the stored muons will be lost during data taking that can adversely affect the measurement of the spin difference frequency  $\omega_a$ . See Section 5.3.3. In order to reduce the number of lost muons, a procedure called “scraping” is used to remove those muons sitting at the edge of the storage region that are more likely to be lost at later times. This scraping procedure involves powering the quadrupole voltages in an asymmetric way such that the beam is pushed to the outside of the storage region, where the edges of the beam will intersect copper collimators. These collimators have a radius of 45 mm and define the storage region aperture. Muons which hit the collimators will lose energy and be lost as they spiral out of the ring. The scraping



Figure 2·14: Electrostatic quadrupoles installed into a vacuum chamber [44]. There are four plates mounted to the chamber through insulator standoffs, with the distance between opposing quadrupole plates equal to 10cm. Also shown are the rails on which the magnetic field trolley rides. Note that as the trolley makes its way through the storage volume, it must pass between both the various sets of quadrupole and kicker electrodes.



Figure 2.15: An OPERA model of the quadrupoles and their equipotential contours [38]. The top and bottom plates sit at positive voltage while the left and right plates are at negative voltage. The muon storage region is shown by the blue circle.

procedure is performed early in the fill and ends at  $8\text{ }\mu\text{s}$ , such that by  $30\text{ }\mu\text{s}$  the beam is stable and centered due to the characteristic RC time constant of the system. The operational performance of the quadrupoles in Run 1 is described in Section 2.8, where it was found that the quads had a longer RC time constant than the design.

## 2.6 Muon beam dynamics

Muons injected into the storage ring will occupy a region in phase space of momenta and positions defined by the injection and collimator apertures. Individual muons will undergo betatron motion within the storage ring in both the vertical and horizontal directions. The horizontal ( $x$ ) and vertical ( $y$ ) equations of motion, including the

effects of the discrete quadrupoles, are given by

$$x = x_e + A_x(s) \cos(\nu_x \frac{s}{R_0} + \phi_x), \quad (2.18)$$

$$y = A_y(s) \cos(\nu_y \frac{s}{R_0} + \phi_y), \quad (2.19)$$

where  $x_e$  is the radial equilibrium orbit of the beam relative to  $R_0$ ,  $A_x(s)$  and  $A_y(s)$  are the amplitudes of the motions containing the effects of the discreteness of the quadrupoles, and  $s$  is the arc length of the trajectory. Here  $\nu_x$  and  $\nu_y$  are the so-called horizontal and vertical “tunes” of the beam motion, which are ratios of the betatron frequencies to the cyclotron frequency  $f_c$ :

$$\begin{aligned} \nu_x &= f_{x_{BO}}/f_c = \sqrt{1-n} \\ \nu_y &= f_{y_{BO}}/f_c = \sqrt{n} \end{aligned} \quad (2.20)$$

These are related to the field index  $n$ , where the field index characterizes the strength of the electrostatic focusing in relation to the magnetic field strength:

$$n = \frac{\kappa R_0}{\beta B_0}, \quad (2.21)$$

where  $\kappa$  is the electric quadrupole gradient,  $B_0$  is the magnetic field strength,  $R_0$  is the central storage ring radius, and  $\beta \cdot c$  is the velocity of the muon beam. Technically  $n$  is the average field index around the ring, where this approximation is justified due to the four-fold symmetry of the discrete quadrupoles and the fact that the betatron oscillations have wavelengths much greater than the length of the quads. A table of the important frequencies in E989 is shown in Table 2.2. Lastly, the maximum angular acceptance of the ring can be determined from the betatron oscillations and

Muon Beam Frequencies				
Name	Symbol	Expression	Frequency (MHz)	Period
$g - 2$	$f_a$	$a_\mu Be/2\pi mc$	0.23	4.365 $\mu$ s
cyclotron	$f_c$	$v/\pi R_0$	6.71	149.2 ns
horizontal betatron	$f_{x_{BO}}$	$\sqrt{1-n}f_c$	6.34	158 ns
vertical betatron	$f_{y_{BO}}$	$\sqrt{n}f_c$	2.21	452 ns
coherent betatron	$f_{CBO}$	$f_c - f_{x_{BO}}$	0.37	2.703 $\mu$ s
vertical waist	$f_{VW}$	$f_c - 2f_{y_{BO}}$	2.31	433 ns

Table 2.2: Frequencies seen in the  $g - 2$  experiment due to beam motion. Parameter values are from a subset of Run 1 corresponding to an  $n$  value of 0.108 or a quadrupole voltage of 18.3 kV.

the field index as

$$\begin{aligned}\psi_{x_{max}} &= \frac{x_{max}\sqrt{1-n}}{R_0}, \\ \psi_{y_{max}} &= \frac{y_{max}\sqrt{n}}{R_0},\end{aligned}\tag{2.22}$$

where  $x_{max}$  and  $y_{max}$  are both equal to the radius of the storage region aperture at 45 mm.

As the muon beam goes around the ring, the muons will experience local field gradients and inhomogeneities. The muons will inevitably pass through the perturbations many times. However, if they don't pass through the perturbations at the same phases of their betatron motion, the amplitude of the would-be resonant oscillation won't continue to grow. The tunes are chosen to avoid these resonances by having the muons sample the entire azimuth of the ring equally, thus keeping the beam stored. The general resonance condition is [45]

$$a\nu_x + b\nu_y = c,\tag{2.23}$$

where  $a$ ,  $b$ , and  $c$  are integers. We know from Equation 2.20 that

$$\nu_x^2 + \nu_y^2 = 1,\tag{2.24}$$



Figure 2.16: The tune plane, with the  $v_x^2 + v_y^2 = 1$  constraint in red. The chosen value of  $n$  lies on this circle. The original design goals for E989 were the  $n$  values as shown by the red points, but due to hardware issues smaller  $n$  values of 0.108 and 0.120 were chosen as described in Section 2.8.

which constrains the available  $n$  values that can be chosen. Figure 2.16 shows the intersections of the resonant lines of Equation 2.23 along with the circular arc of Equation 2.24 in the tune plane, for which a chosen value of  $n$  will lie on a resonance. The operational  $n$  values and corresponding quadrupole voltages in Run 1 are described in Section 2.8.

### 2.6.1 Coherent betatron oscillation

The muon beam consists of many muons, each individually undergoing betatron oscillations. If the phase of the oscillations of the individual muons are incoherent, then the beam can be thought of as a static entity, constant in time around the ring. However, due to injection and kicker effects which induce a particular phase space distribution on the injected beam, the beam itself can be said to oscillate. The beam

is then described by a width and mean dependent on the injection process, and the strength and phase of the kicker pulse, such that the phase distribution of the beam oscillates coherently every betatron wavelength<sup>4</sup>. Due to the mismatching of betatron wavelengths to the ring circumference in order to avoid resonances, a singular time slice of the distribution can be said to move around the ring over time. Individual detectors around the ring measure the beam in discrete pieces based on their individual azimuthal acceptances, where these acceptances depend on the radial and vertical characteristics of the beam. Because the radial betatron frequency is larger than half the cyclotron frequency, there is an aliasing effect such that the radial betatron motion of the beam is instead observed as an apparent slow-moving oscillation. We call the measurable signal of this coherent radial motion coherent betatron oscillation (CBO). See Figure 2.17 for a pictorial view of this phenomenon. Since the acceptance of the calorimeters depend on the beam properties, the CBO will modulate the  $\omega_a$  signal.

The frequency of the CBO is just the beat frequency between the cyclotron frequency and the horizontal betatron frequency

$$f_{CBO} = f_c - f_{x_{BO}}. \quad (2.25)$$

There is also a vertical CBO effect, but the non-aliased rate of oscillation is fast enough such that the effect tends to average out. However, the detectors are sensitive to oscillations of the vertical width of the beam, which is aliased in a similar way to the radial oscillation. Though the principles are the same, we call this effect the vertical waist (VW),

$$f_{VW} = f_c - 2f_{y_{BO}}, \quad (2.26)$$

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<sup>4</sup>The four-fold quadrupole symmetry was chosen in order to minimize this beam ‘breathing.’



Figure 2.17: Marked by the black vertical lines are integer steps in the circumference of the ring, corresponding to the cyclotron wavelength  $\lambda_c$ . The blue line shows the motion of the beam due to the betatron oscillations  $\lambda_x$ . Since  $\lambda_x > \lambda_c/2$ , there is an aliasing effect in the observed signal, which is identified by the red line. To a single detector the beam appears to move slowly back and forth with  $f_{CBO}$ .

where the term *waist* refers to the minimum vertical width. Both of these frequencies are included in Table 2.2. The phase of the CBO signal varies systematically by detector, from 0 to  $2\pi$  around the ring. When adding all of the detector signals together, the CBO oscillations tend to cancel out. However, due to acceptance differences between the different detectors, the CBO oscillations are still observable in the data. When fitting the data to extract  $\omega_a$ , these effects must be accounted for in the fit function, as will be discussed in Section 5.3.1.

### 2.6.2 Beam debunching

As described in Sections 2.3 and 2.4, the muon beam is injected into the ring with a time spread of 120 ns and a range of momenta. At early times the beam will occupy a portion of the ring less than the whole since the cyclotron period is  $\sim 149$  ns. At initial injection, detectors located at discrete points around the ring see a high intensity as the beam passes by and low intensity when it is on the other side of the ring. The cyclotron period of each muon is determined by its particular momentum which gives its radius of orbit. Higher momentum muons have a larger radius and therefore larger

distance to travel, thus taking longer to circulate the ring than lower momentum muons. This leads to the low momentum muons in the head of the bunch catching up with the high momentum muons in the tail, such that the gaps in between each passing of the beam bunch are reduced. By 30  $\mu\text{s}$  the muon beam has gone around the ring two hundred times and the gaps with no beam have reduced so much that the intensity is near-uniform. The varying intensity from the beam bunching is referred to as the “fast rotation” signal. This debunching signal is seen in the data as shown in Figure 2.18. When dealing with the data and attempting to extract  $\omega_a$ , the typical procedure is to both bin out the fast rotation in periods of the cyclotron frequency, and to randomize each hit time by  $\pm T_c/2$  where  $T_c$  is the cyclotron period. In this way the fast rotation is removed entirely and the effect can be ignored in the fitting procedure for  $\omega_a$ .

## 2.7 Corrections to $\omega_a$

Equation 2.3 is an idealized version of the spin difference frequency which ignores two important beam dynamics effects: torques exerted by the electric field and changes in the rest frame magnetic field resulting from the vertical pitching motion of the muons. Including practical experimental concerns, these two corrections must be applied to  $\omega_a$ .

### 2.7.1 Electric field correction

In the presence of an electric field, the spin difference frequency is altered to

$$\vec{\omega}_a = -\frac{q}{m}[a\vec{B} - \left(a - \frac{1}{\gamma^2 - 1}\right)(\vec{\beta} \times \vec{E})], \quad (2.27)$$

where now there is an extra term dependent on the electric field strength and the momentum of the particles. This extra term originates from the motional magnetic



Figure 2.18: In the data, the fast rotation signal is a fast (149.2 ns) modulation on top of the slower (4.365  $\mu$ s)  $\omega_a$  frequency. The plots here are from a subset of data from Run 1. In individual calorimeters at early times the fast rotation signal is seen to be very large, as shown on the top left. As time passes and the beam debunches, the amplitude of the fast rotation signal diminishes as shown on the top right. When adding all calorimeters together, the signal is reduced as shown in the bottom two plots, due to the fast rotation signal being unaligned in phase for the different detectors.

field  $\vec{\beta} \times \vec{E}$  that relativistic particles experience in an electric field. This is necessary to include since we use electrostatic quadrupoles for vertical focusing as described above. The second term cancels to first order for a specific momentum or value of  $\gamma$ . This “magic momentum” can be understood as the momentum at which a relativistic particle moving through an electric field has its spin exactly follow its momentum. This magic momentum is 3.094 GeV for muons, hence the momentum value of the injected muons. This value sets the energy and time scales of the experiment and has driven many of the design constraints, including the size of the storage ring, choice of the magnetic field magnitude, etc.

Not all muons will have the magic momentum however as described in the Section 2.6.2, and therefore a correction to the measured  $\omega_a$  frequency needs to be applied. Approximating the storage ring as having an electric field applied over the whole azimuth of the ring, the spin difference frequency for muons with momentum  $p \neq p_m$  (where  $p_m$  is the magic momentum) becomes

$$\omega'_a = \omega_a \left[ 1 - \beta \frac{E_r}{cB_y} \left( 1 - \frac{1}{a\beta^2\gamma^2} \right) \right]. \quad (2.28)$$

Here the motion of the beam is assumed purely azimuthal. This additional term is the electric field correction that then serves to lower the measured  $\omega_a$  frequency. Using the relation  $p = \beta\gamma m = (p_m + \Delta p)$ , after a little bit of simplification the electric field correction can be written as

$$C_E = \frac{\Delta\omega_a}{\omega_a} = -2 \frac{\beta E_r}{cB_y} \frac{\Delta p}{p_m}. \quad (2.29)$$

The last fraction can be related to the field index described in Equation 2.21 by

$$\frac{\Delta p}{p_m} = (1 - n) \frac{\Delta R}{R_0} = (1 - n) \frac{x_e}{R_0}, \quad (2.30)$$

since we know that the magic momentum muons are at the central radius  $R_0$  of

the storage ring. In this equation  $x_e = \Delta R$  is the equilibrium radius of the muon relative to the central storage radius. Noting that the radial electric field strength for a quadrupole is

$$E = \kappa x = \frac{n\beta c B_y}{R_0} x, \quad (2.31)$$

and assuming that it is perfectly radial, the electric field correction reduces to

$$C_E = -2n(1-n)\beta^2 \frac{xx_e}{R_0^2}. \quad (2.32)$$

Taking the time average of the beam motion, where  $x$  is simply equal to  $x_e$ , the correction becomes

$$C_E = -2n(1-n)\beta^2 \frac{\langle x_e^2 \rangle}{R_0^2}. \quad (2.33)$$

Since the equilibrium radius of the beam is set by the momentum distribution of the muons, this electric field correction can be determined by a measurement of the momentum spread of the beam which comes from an analysis of the fast rotation [46, 47]. For the precision goal of E989, the assumptions made in this derivation are acceptable [38] and results will be cross-checked with spin-tracking simulations as was done in E821 [29]. In E821 the electric field correction was approximately 500 ppb on  $\omega_a$  [29]. In E989 the scale of the correction will be the same considering the experimental principles are identical, Section 5.5.8.

### 2.7.2 Pitch correction

Particles injected into the  $g - 2$  storage ring will have a vertical component of momentum which is parallel to the magnetic field vector (hence the need for vertically focusing electrostatic quadrupoles). This will slightly reduce the magnetic field seen by the muons in their rest frame. Including this motion into the spin difference



Figure 2.19: Beam motion  $\beta$  relative to the vertical and azimuthal axes Y and Z respectively.  $\psi$  is the pitch angle of the beam, and the dashed lines represent the parallel and perpendicular motions of the beam.

frequency,  $\omega_a$  becomes

$$\vec{\omega}_a = -\frac{q}{m}[a\vec{B} - a\left(\frac{\gamma}{\gamma+1}\right)(\vec{\beta} \cdot \vec{B})\vec{\beta}], \quad (2.34)$$

where now there is an extra term dependent on the vertical betatron motion of the beam. Similar to the electric field case, this term can be neglected to first order as the muon momentum is nearly all perpendicular to the field, but a correction again needs to be applied to  $\omega_a$  to account for this effect.

Since the muons in the storage ring will be oscillating vertically as they are focused by the quadrupoles, their momentum vectors will be pitching up and down relative to the azimuthal motion. This pitch angle will oscillate as

$$\psi = \psi_0 \cos(\omega_y t), \quad (2.35)$$

where  $\psi_0$  is the amplitude of the oscillation and  $\omega_y$  is the vertical betatron frequency. Shown in Figure 2.19 is an exaggerated example of the beam motion relative to the vertical and azimuthal axes. Assuming that the field is purely vertical,  $\vec{B} = B_y \hat{y}$  and

that the beam motion is in the vertical-azimuthal plane,

$$\vec{\beta} = \beta_y \hat{y} + \beta_z \hat{z} = \beta \sin(\psi) \hat{y} + \beta \cos(\psi) \hat{z}, \quad (2.36)$$

then  $\omega_a$  becomes

$$\vec{\omega}_a = -\frac{q}{m} [a B_y \hat{y} - a \left( \frac{\gamma}{\gamma+1} \right) \beta_y B_y (\beta \sin(\psi) \hat{y} + \beta \cos(\psi) \hat{z})]. \quad (2.37)$$

Using the small angle approximation such that  $\cos(\psi) \approx 1$  and  $\sin(\psi) \approx \psi$ ,  $\vec{\omega}_a$  can be separated into its vertical and azimuthal components

$$\omega_{ay} = \omega_a \left[ 1 - \left( \frac{\gamma-1}{\gamma} \right) \psi^2 \right], \quad (2.38)$$

$$\omega_{az} = -\omega_a \left( \frac{\gamma-1}{\gamma} \right) \psi. \quad (2.39)$$

Looking at Figure 2.19 again, it can be seen that the spin difference frequency can be resolved into its parallel and perpendicular components  $\omega_{\parallel}$  and  $\omega_{\perp}$  respectively. As the pitch angle of the beam motion oscillates about the azimuthal axes at a frequency much greater than the  $g - 2$  frequency, it can be seen that the parallel component averages to 0 over time. We then only care about the perpendicular oscillation of the beam, which can be determined with a simple rotation matrix such that

$$\omega_a \approx \omega_{\perp} = \omega_{ay} \cos(\psi) - \omega_{az} \sin(\psi) \approx \omega_a \left[ 1 - \frac{\psi^2}{2} \right], \quad (2.40)$$

where in the last approximation the small angle approximation was used once again, but this time with  $\cos(\psi) \approx 1 - \psi^2/2$ . The pitch correction then is the additional term which serves to lower the measured spin difference frequency. Taking the time average,

$$C_P = \frac{\Delta \omega_a}{\omega_a} = -\frac{\langle \psi^2 \rangle}{2}. \quad (2.41)$$

The pitch angle of the beam cannot be measured directly, however we know from Equation 2.22 that the angle of the beam can be related to the vertical distribution of the beam, such that

$$C_P = -\frac{n}{2} \frac{\langle y^2 \rangle}{R_0^2}, \quad (2.42)$$

where once again  $n$  is the field index,  $R_0$  is the radius of the ring at the center of the storage region, and  $\langle y^2 \rangle$  is the vertical width of the beam. The first two are known and the last can be measured experimentally by the straw tracking detectors. Just as in the case of the electric field correction, the assumptions made in this derivation are acceptable for the precision goal of E989, and results will be cross-checked with spin-tracking simulations. In E821 the pitch correction was approximately 300 ppb on  $\omega_a$  [29], and the scale for the E989 correction is about half that, Section 5.5.8.

## 2.8 Run 1 in E989

Run 1 for E989 was conducted in the first half of 2018. Production data were gathered from March 22nd through June 29th. Because of accelerator, experimental, and practical concerns common to the early stages of production running, production data taking was interrupted at various dates. Due to hardware issues both kicker and quadrupole voltages were originally lowered from their design values. Various voltage set points for both systems were identified and used in separate periods of the data taking, depending on the stabilities of the systems. The primary precession frequency analysis datasets gathered by E989 and their associated running parameters are shown in Table 2.3. Other datasets were gathered but they were not used to calculate the precession frequency in this analysis<sup>5</sup>.

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<sup>5</sup>Specifically it is expected to include the LowKick dataset in future analyses, which has approximately 80% of the statistics of the 60h dataset.

Run 1 Precession Frequency Analysis Datasets					
Name	Date Acquired	Number $e^+ > E_{\text{Th}}$	Quad Voltage (kV)	$n$ Value	Kicker Voltage (kV)
60h	4/22/18–4/25/18	$9.34 \times 10^8$	18.3	0.108	128–132
HighKick	4/26/18–5/02/18	$8.70 \times 10^8$	20.4	0.120	136–138
9d	5/04/18–5/12/18	$2.13 \times 10^9$	20.4	0.120	128–132
Endgame	6/06/18–6/29/18	$4.10 \times 10^9$	18.3	0.108	122–127
Total Positrons Above Threshold		$8.03 \times 10^9$			

Table 2.3: The primary datasets acquired during Run 1 of E989 in 2018 and their associated parameters [48]. Other datasets were gathered but they were not used in the calculation of the precession frequency in this analysis.  $E_{\text{Th}} = 1.7 \text{ GeV}$ .

Due to the lower kicker voltages, the muon beam was stored on a central radius  $\sim 7$  mm offset from the central orbit of the storage ring. The associated number of stored muons per fill for Run 1 as a result of the lower kicker voltage was  $\mathcal{O}(4,000)$ , down from the  $\mathcal{O}(10,000)$  design goal. The chosen quadrupole voltages were 18.3 and 20.4 kV, corresponding to  $n$  values of 0.108 and 0.120 respectively [49]<sup>6</sup>. During Run 1 it was discovered that some of the quadrupole resistors were damaged. They had longer RC time constants such that the quadrupole voltages had not reached nominal values at the beginning of the designated analysis portion of the data at 30  $\mu$ s, and were still changing over the course of a fill, see Figure 2.20. The muon beam was seen to move as a function of time in-fill as a consequence of this. See Section 4.4 for a summary of the muon beam characteristics for Run 1. As described in various sections in this chapter and in Section 4.4, these muon beam characteristics fold into the measurement of  $g - 2$  in a variety of ways.

All data listed in Table 2.3 were quality checked. If run conditions were found to be unsatisfactory, the associated data was flagged and ignored in the analysis. A summary of the data quality control procedure is given in Reference [50], and the exact data quality parameters for the 60h dataset are detailed in Reference [51]. The total number of detected positrons above threshold that passed the data quality control was approximately  $1 \times 10^{10}$ , corresponding to a statistical error on  $\omega_a$  of  $\sim 400$  ppb, as calculated with Equation 2.16.

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<sup>6</sup>The associated betatron wavelengths are 1.06 and 3.04 times the circumference of the storage ring respectively.



Figure 2-20: Traces for some quadrupole plate potentials as a function of time. The different plates are identified in the legend, where for instance Q1LT stands for the top long plate of quadrupole 1. The muon beam is injected into the ring at the black line and the scraping procedure occurs until the quadrupole voltages reach their design values, in this case 20.4 kV. As shown some of the high voltage traces do not behave in a smooth and fast exponential manner, due to damaged quadrupole resistors. The traces in orange and brown are for good quadrupole resistors, while the traces in red and blue show the results from damaged quadrupole resistors. The kink in blue is the switch from the scraping procedure to storage.

# Chapter 3

# Detector Systems

There is a variety of different detector systems used in E989. The primary detectors are the calorimeters which measure the  $\omega_a$  signal. In support of that measurement there are several auxiliary systems used for monitoring injection and beam dynamics. These include the T0, IBMS, and fiber harps. There is also a straw tracking detector which measures decay positron trajectories which can be related to muon beam dynamics and calorimeter measurements. Each of these systems will be described in the following sections.

## 3.1 Auxiliary detectors

### 3.1.1 T0

A reference time must be chosen so as to align all different detector systems in time. Without this functionality, analyzing the data from different systems and comparing them would be impossible. Similarly, decay positron spectra must be aligned in phase from fill to fill, otherwise the  $\omega_a$  signal would wash out. To do this a “T0” counter is placed on the outside of the ring just before the inflector. It is made up of a scintillating paddle connected to two photo-multiplier tubes (PMTs) [52, 53]. See Figure 3-1a. PMT A has a 1% neutral density filter which results in a low photo-electron statistics measurement, and acts primarily as a timing measurement. PMT B has a 10% neutral density filter which results in a higher statistics measurement, and acts as a shape counter and proxy for fill intensity. (In general the signals are



(a) The T0 counter is made up of a scintillator in the middle shown in green, which connects with light guides to PMTs on the left and right.

(b) Time profiles for the two PMTs in the T0 counter for one of the eight bunches. Each profile is an average of 100 such profiles. The x axis is in clock ticks, where  $1 \text{ ct} = 1.25 \text{ ns}$ .

Figure 3.1: A model of the T0 detector (left) and a sample of the signal it measures every fill (right).

very similar.) Together they provide a measure of the injected beam profile in time, from which the timing alignments can be made. Some measured T0 pulses are shown in Figure 3.1b. The shape of the incoming pulses has a somewhat trident-like shape, with a very large spike in the middle of the time profile and two spikes on the outside edges. This shape owes itself to the accelerator complex, and varies somewhat from bunch to bunch.

### 3.1.2 Inflector Beam Monitoring System

While the T0 provides timing and intensity measurements, the inflector beam monitoring system (IBMS) provides measurements of the beam position properties as it passes through the inflector. This is useful because the injection aperture is so tight, and incoming beam parameters are tightly constrained. The IBMS system serves to provide a direct diagnostic handle on the phase space matching between the last accelerator components and the storage ring, helping to optimize the number of stored



Figure 3·2: Models of the IBMS 1 and 2 detectors. Scintillating fibers form an array which detect particles as they pass through them.

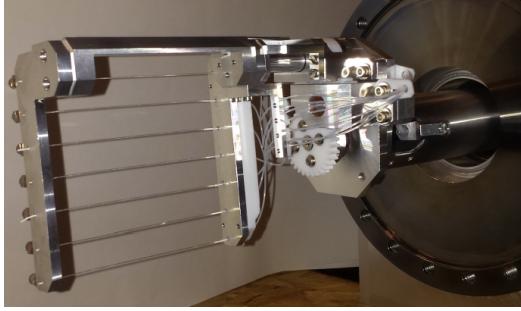


Figure 3·3: The positions of IBMS 1, 2, and the planned 3rd device are shown with respect to the vacuum chamber and inflector.

muons per fill [54]. The IBMS is made up of two scintillating fiber detectors, shown in Figure 3·2. These devices are placed at the outside of the magnet yoke before injection into the back hole of the magnet, and at the entrance to the inflector [55]. A third device is planned to be at or near the downstream end of the inflector. See Figure 3·3.

### 3.1.3 Fiber harps

The last auxiliary detector is the “fiber harp” system. It is made up of four scintillating fiber detectors that serve to measure the beam intensity as a function of time



(a) Picture of one of the fiber harps. A row of seven scintillating fibers measures the beam intensity as a function of time at vertical or horizontal positions depending on which fiber harp is inserted.

(b) Shown are fiber harp beam intensity measurements for a horizontal fiber harp. Each spectra is from a single fiber, with the spectra at the top being at the greatest radial position. The fast oscillations of the cyclotron frequency can be seen along with the slower oscillations of the CBO.

Figure 3·4: Fiber harp detector (left) and a sample of fiber harp measurements (right).

and position within the storage ring [56]. Two of the detectors measure the radial components of the beam, and two measure the vertical components. Each pair of fiber harps are located  $180^\circ$  and  $270^\circ$  in azimuth clockwise from the inflector. One of these devices is shown in Figure 3·4a. The fiber can measure beam properties throughout the fill, providing a diagnostic tool which is sensitive to the scraping procedure and the CBO properties of the beam. An example of fiber harp measurements is shown in Figure 3·4b. However, because the fiber harps are a destructive measurement of the beam due to multiple scattering in the fibers, the system was designed to be retractable. They are inserted during special diagnostic or systematic runs, and are pulled out of the beam path during production data taking.

## 3.2 Calorimeters

Electromagnetic calorimeters are the primary detector in the E989 experiment, responsible for the  $\omega_a$  measurement. They measure arrival times and energies of the decay positrons as they curl inward from the storage region.

### 3.2.1 Requirements and systematic effects

In order to determine  $\omega_a$  to the precision goal, there are specific requirements on the performance of the calorimeters. First, they must have a relative energy resolution of better than 5% at 2 GeV, in order for proper event selection above threshold [38]. The energy resolution requirement is modest but not strict because the important observable is the number of detected positrons above some energy threshold, where the optimal energy threshold can be determined empirically.

Second, they must have a timing resolution of better than 100 ps for positrons with energy greater than 100 MeV [38]. The calorimeters must be able to resolve multiple incoming hits through temporal or spatial separation at 100% efficiency for time separations of greater than 5 ns, or 66% of hits for time separations less than 5 ns, in order to reduce the pileup systematic effect. Pileup is the term used for when multiple particles impact the detector within the deadtime of the detector such that they are registered as a single hit. Unresolved pileup means that the number of detected positrons above threshold is mis-measured. Since pileup is a time-dependent effect, where pileup decays with a lifetime approximately equal to half the muon lifetime, this leads to a mis-measurement of  $\omega_a$ , see Section 5.2.1. The requirements stated here assist in reducing the pileup systematic error below a target goal of 40 ppb.

Third, the energy response of calorimeters for measured hits must be stable to  $< 0.1\%$  over a fill (700  $\mu$ s) [38]. The long term energy response stability over a time period of order seconds must be  $< 1\%$ . The energy response of a detector as a function

of time is typically referred to as the gain of the detector, where technically the gain refers to the amount of current output per detected hit. The gain depends on the temperature stability and hit rate. After a hit, the measured energy fraction of a following hit drops sharply and then rises exponentially back to one. This occurs as a result of pixels in the detector being unable to fire consecutively within short time frames,  $\mathcal{O}(\text{ns})$ , and because of charge depletion in the calorimeter system's electronics capacitative components,  $\mathcal{O}(\mu\text{s})$ . The former is referred to as the short-term double pulse (STDP) effect [57]. The latter is referred to as the in-fill gain (IFG) effect [58], where the effect is due to the high rate of incident particles at injection combined with secondaries which overcomes the recovery response of the electronics at early times in the fill<sup>1</sup>. Hits with mis-measured energies due to these effects can thus be excluded from the fitted histogram if their energies drop below threshold, leading to another systematic effect and subsequent error in the  $\omega_a$  measurement. (Temperature changes vary over time scales greater than a fill, so they don't contribute to the systematic error.) The requirements stated here, along with the use of a laser calibration system (Section 3.2.3), assist in reducing the gain systematic error below a target goal of 20 ppb.

### 3.2.2 Hardware

There are 24 calorimeters located symmetrically around the inside of the ring placed flush to the vacuum chamber wall, as shown in Figure 3.5a. Indeed the shape of the vacuum chambers were designed so as to reduce multiple scattering of the decay positrons before entering the calorimeters. Each calorimeter sits on a board extending out from a cart which contains the electronics that power the calorimeters and read out the data. The carts help to relocate magnetic materials away from the field region

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<sup>1</sup>The large of amount of particles and secondaries at the beginning of a fill is referred to as the 'flash'.



(a) Close up view of a model of the storage ring and calorimeter detector systems. In black are carts which house the electronics for the calorimeters. They hold up a board upon which the calorimeter rests, identified by the long gray box. The calorimeter is pushed flush to the vacuum chamber on the left.

(b) Shown is the backside of a calorimeter. The clear blocks are PbF<sub>2</sub> crystals, each of which has a SiPM mounted onto the end in order to collect emitted Cerenkov light from incident particles.

Figure 3.5: Calorimeter detector systems relative to the vacuum chamber (left) and an individual calorimeter (right).

to avoid perturbing the magnetic field, and provide easy access to the electronics while removing them from the positron decay path region.

Each calorimeter consists of 54 channels of PbF<sub>2</sub> crystals in a 6 high by 9 wide array, for a total of 1296 channels. Each crystal is  $2.5 \times 2.5 \times 14$  cm<sup>3</sup> and is wrapped in black Tedlar® foil. The PbF<sub>2</sub> material has an index of refraction of 1.8, and emits Cerenkov light as incident positrons with energy greater than 100 keV pass through the crystals [59]. Cerenkov light is naturally fast which improves the timing resolution of the incoming hits. The high density of the PbF<sub>2</sub> (7.77 g/cm<sup>3</sup>) gives it a short radiation length (9.3 mm), such that the energy deposition from the incident positrons is nearly 100% over the length of the crystal. The black foil is used to eliminate light transmission between crystals to improve position reconstruction, as well as reduce internal reflection to reduce pulse width and improve timing resolution

[60]. The energy deposition from the incident positrons is typically restricted to only one or two crystals. The segmentation of the calorimeter combined with the black wrapping helps the spatial and temporal resolution of the detected pulses. See Figure 3.5b for a picture of the calorimeter crystals.

Each crystal is paired with a large area silicon photo-multiplier (SiPM) sensor which detects the emitted Cerenkov light. SiPMs are compact, operable in high magnetic field regions, have very linear responses for operation at MHz rates, are suited to measuring Cerenkov light due to their high photo-detection efficiency in the wavelengths of interest, and have a high degree of gain stability [60]. The SiPMs used are designed by Hamamatsu and detailed in Reference [60]. They sit on printed circuitry boards devoid of magnetic materials, which are designed to preserve the fast SiPM pulse shape. The combined properties of the chosen SiPMs, their electronic boards, and the  $\text{PbF}_2$  crystals results in an energy resolution of  $(4.6 \pm 0.3)\%/\sqrt{E/\text{GeV}}$ , a timing resolution of 20 ps, and a pulse width of 5 ns, satisfying the stated requirements earlier [59, 60]. Details of the pulse fitting algorithms are shared in Section 5.1. The SiPMs can be seen in Figure 3.5b. Finally, 12-bit waveform digitizers sample each SiPM channel at a rate of 800 mega-samples per second with a 1 Gb memory buffer and the data are transferred to a bank of GPU processors for on-line data processing [61]. The timing resolution of these waveform digitizers is  $< 50$  ps for most pulse amplitudes.

### 3.2.3 Laser calibration system

In order to satisfy the gain requirements of the calorimeter detectors, a laser calibration system is used. This system monitors the SiPM responses over short and long time scales to  $< 0.04\%$  [62]. The system consists of six different lasers and a suite of optical devices. The light from the six lasers is piped through optical fibers to a board mounted on the front face of each calorimeter. This board contains right-angle

prisms which deflect the laser pulses directly into each calorimeter crystal, for all 1296 channels. The lasers can be pulsed at various intensities, both in-fill to monitor the STDP or IFG effects, and out-of-fill to monitor for long term drifts due to changing temperatures. The SiPM measured response is compared to local known source monitors in order to calibrate the SiPM energy response. Corrections to the gain of the calorimeters can thus be determined and applied to the hit energies. Simultaneously the laser system allows for measuring the timing resolution of the SiPMs, and in general performing diagnostic tests with the calorimeter. Lastly, the laser system is used to time align the different calorimeter channels by outputting a sync pulse to each channel at the beginning of every fill.

### 3.3 Straw trackers

As described in Section 2.6, the muon beam moves as a whole within the storage ring. As explained in Section 2.2, the muon beam distribution ties into the measurement of the average magnetic field. The primary detector system used to measure this behavior and determine the muon beam distribution is the “straw tracker” system. The straw trackers characterize the beam in a non-destructive fashion by measuring decay positron trajectories and extrapolating them back into the storage ring. They serve to provide the direct measurement of the pitch correction as described in Section 2.7.2, determine the momentum distribution of the beam, and characterize parameters of the CBO which tie into the calorimeter  $\omega_a$  analysis. Decay positron trajectories can also be extrapolated forwards into the calorimeter, in order to cross-check calorimeter measurements to help resolve pileup. Finally, the trackers also serve as a measuring device to search for a possible muon electric dipole moment. The existence of such a thing would tilt the precession plane of the muons and subsequent decay positron trajectories which the trackers are sensitive to. Details of the beginnings of such a search

are given in Reference [63]. The details of the track reconstruction and analysis will be given in Chapter 4. Here will be given a description of the detectors themselves.

In general, straw tracking detectors work by measuring hits in gas filled straws [64]. Each straw is made up of a cylindrical piece of thin material, typically Mylar with a conductive aluminum coating, and contains a wire at the center of the straw held at high voltage,  $\mathcal{O}(1000\text{ V})$ . The minimal amount of material in straw trackers serves to reduce multiple scattering of incident particles, and was the reason a straw tracker system was chosen over other types of trackers. Fast moving charged particles ionize the gas as they pass through it, and the resulting ions are drawn to the wire and straw surface (positive and negative charges respectively). As the ions move to the wire, they enter a high electric field region that causes them to speed up, hit other gas molecules, and create more ions. This produces an avalanche gain effect which amplifies the original signal. Once the ions reach the wire and straw surface, an electrical signal is read out telling what the drift time of the ions was, which can be related to the radius or distance of closest approach (DCA) at which the particle passed relative to the wire. The straws therefore measure drift circles in a plane perpendicular to their physical orientation. The calculations of these DCA's and their errors is done using a simulation program for gaseous detectors called Garfield [65, 66]. By combining several such measurements in time and space, tracks of incident particles can be reconstructed. See Figure 3·6.

The straws used in E989 are 5 mm in diameter and contain a 50:50 mixture of argon-ethane gas [67]. The argon component serves as the gas to be ionized. The fast moving ions near the wire in addition to producing more ions, will incite excited states of the gas, which emit photons when de-excited. These photons can restart the whole ionization and avalanche process over again as they can escape the avalanche region, leading to a break-down of the system. The ethane therefore serves to absorb the



Figure 3.6: Diagrams showing the determination of a drift time  $t_d$  and hit time  $t_h$  from an incident ionizing particle (left) and the combination of several such hits to produce a track (right). Diagram on the left courtesy of Saskia Charity.

photons with its large number of molecular degrees of freedom and photo-absorption characteristics [67]. The wire has a radius of  $25\text{ }\mu\text{m}$  and is made up of gold-plated tungsten. The Mylar walls have a width of  $15\text{ }\mu\text{m}$  and are wound in a double spiral shape in order to improve the electrostatic shielding of the wire and reduce the gas leak rate of the straws [67]. The DCA resolution of hits within the straws is approximately  $120 - 150\text{ }\mu\text{m}$ .

A tracker module is shown in Figure 3.7. Each tracker module consists of 4 layers of 32 straws each with stereo angles of  $\pm 7.5^\circ$ , for a total of 128 straws per module. The first two straw layers are designated as “U” layers, and are oriented with the tops of the straws at greater radial positions than the bottoms. The second two layers are designated as “V” layers, and are oriented with the bottoms of the straws at greater radial positions than the tops. Both U or V layers are collectively referred to as the U and V views of the module respectively. See Figure 3.8. The two layers in



Figure 3.7: A straw tracker module. The first layer of silver Mylar straws with a stereo angle of  $7.5^\circ$  can be seen, with the other three straw layers hiding behind it. In black on the left is the carbon fiber post which holds the end of the module in a fixed position. Electronics live in the top and bottom manifolds above and below the straws, and cables connect from those electronics through small apertures to external electronics which plug in on the right.

each view are shifted half a straw diameter horizontally from one another in order to help resolve which side of the wire any incident particles passed, Section 4.2.6. The two types of layers are non-parallel to each other in order to resolve incident tracks in 2D space. The small stereo angle both improves the straw measurement area as electronics can be kept out of the positron decay path region, and improves the radial momentum resolution of the fitted tracks, since both views measure mostly in the horizontal plane. The active straw measurement area is approximately 10 cm high by 20 cm wide. A carbon fiber post sits at the outside end of the module to provide structural rigidity to the module, and keep the straw wires under tension.

There are two tracker stations, each consisting of eight tracker modules. The two stations are located at positions in front of calorimeters 13 and 19, or at approximately  $180^\circ$  and  $270^\circ$  clockwise from the inflector. These are designated as Tracker

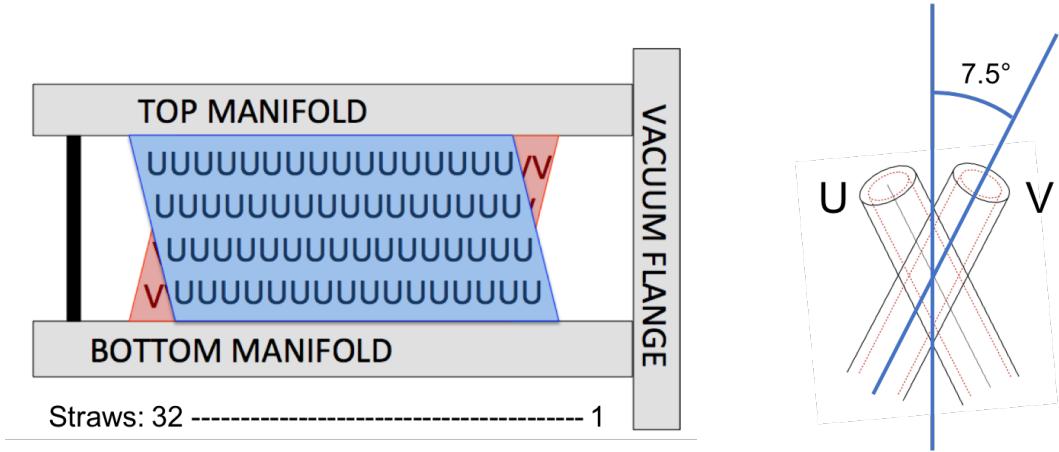


Figure 3.8: Orientation of U and V straws in the tracker module.

Station 12 and Tracker Station 18 respectively, where the numbers come from which vacuum chamber the trackers reside in. There is also a third station just after the inflector, which currently sits empty of any tracker modules. The straw tracker modules sit inside the vacuum chamber in specially modified vacuum chamber sections in a staircase-like pattern that follows the curvature of the ring. See Figure 3.9 and Figure 3.10. The number of modules and their orientation in each station was chosen to provide a long measurement arm for precise momentum measurement of the incident tracks. The modules sit inside the vacuum in order to eliminate multiple scattering in air and produce better reconstructed tracks. Due to their proximity to the storage region, the tracker modules are located in a region of high field non-uniformity. Though the acceptances between the trackers and calorimeters are not identical, their close proximity facilitates comparison between the two measurement devices.

### 3.3.1 Tracker readout electronics

The readout electronics for the system are split into two groups, the front-end and back-end. The front-end electronics were built and tested in the Boston University Electronic Design Facility. They start with ASDQ (Amplifier Shaper Discriminator with charge (Q)) chips on boards that plug down directly onto the straws and pro-



Figure 3·9: Birds eye view of a model of a vacuum chamber containing a tracker station, and the associated calorimeter. Each tracking station consists of eight tracker modules.



Figure 3·10: View of the tracker modules from within the storage ring. Muons move forward into the page, some of which decay to positrons which then pass through the trackers arrayed on the right. The carbon fiber post for each module can also be seen.

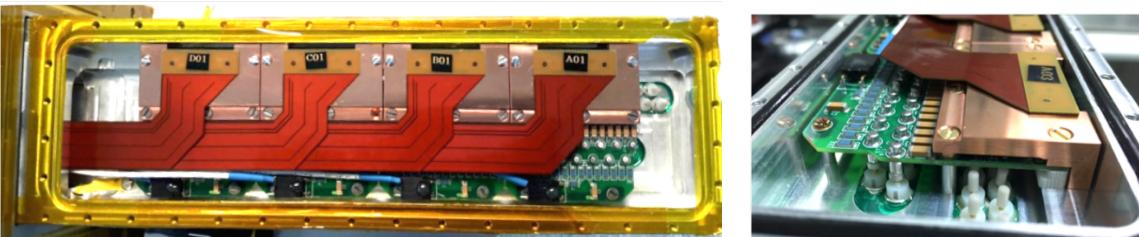
vide the main signal, Figure 3·11a. Each ASDQ plugs onto sixteen straws. These ASDQ boards are application specific integrated circuits (ASICs) which read out the signal from one end of the straws and shape and discriminate that signal, as well as provide some baseline restoration and tail cancellation. The ASDQs and their associated components are contained within thin aluminum manifolds above and below the straws. The physical footprint of these boards and components was minimized in order to increase the straw measurement area. The signals from the ASDQs are passed through Flexi Cables to Time to Digital Converter (TDC) boards which time stamp the signals with 625 ps precision [67], Figure 3·11c. The Flexi Cables are flexible and thin such that the cables can be passed through the thin aperture shown on the right of Figure 3·7. High voltage is provided to the sense wires with high voltage boards, which pass the voltage through the ASDQs. A feedthrough board provides the interface between the TDCs and Flexi Cables, as well as the high voltage signal, Figure 3·12. They serve a dual purpose of providing the gas seal. Finally, there are logic boards that serve as the interface between the TDCs and the back-end electronics. They manage the clock and controls for the TDCs and store data onto FPGAs, which are piped out through a high throughput optical fiber connection. The logic boards, high voltage boards, and TDCs are all housed within an aluminum box which provides RF shielding, Figure 3·13. In each module there are eight ASDQs, four TDCs, two feedthrough boards, two high voltage boards, and two logic boards. An overview diagram of the front-end readout chain is shown in Figure 3·14. The back-end electronics consist of FC7s, one per tracker station, and a single AMC13 for all tracker stations. These modules provide clock and DAQ services to the whole tracker system, and ultimately pipe out the data to where it can be saved on disk.



(a) ASDQs contained within the aluminum manifold (left) that plug down onto the straws (right).



(b) Copper cooling bars (left) keep the ASDQs cool, and a high voltage spacer (right in white) keep the high voltage components isolated.



(c) Orange Flexi Cables plug down onto the ASDQ boards and run left through a small aperture in the tracking detector to the TDCs.

Figure 3-11: ASDQs and associated components which read out the signals from the straws.

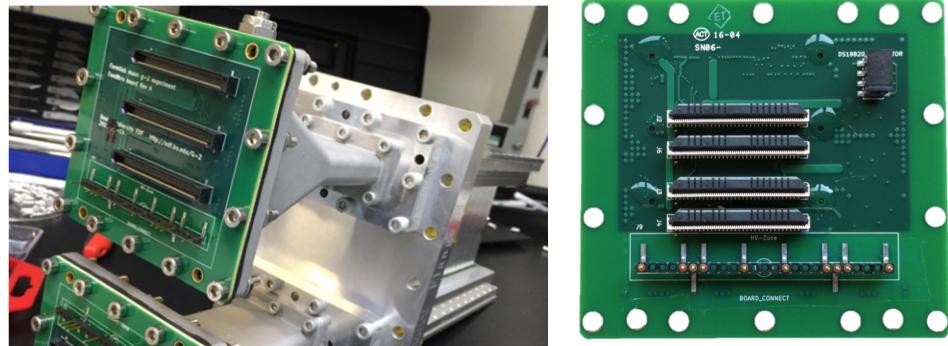
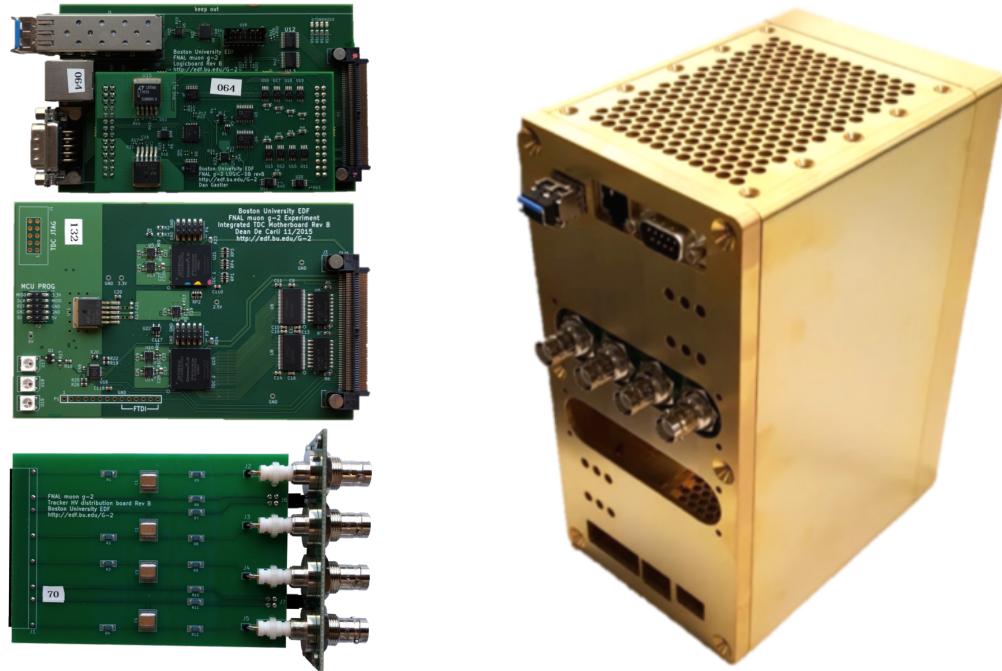


Figure 3.12: Tracker feedthrough board. The feedthrough board plugs onto the end of the tracker snout to form a gas seal (left). Outside of the gas seal, a logic board, two TDCs, and a high voltage board all plug into the feedthrough board. Flexi Cables plug into the back side of the feedthrough board (right).



(a) A logic board (top), TDC (middle), and high voltage board (bottom). (b) The aluminum housing which houses the front-end readout boards.

Figure 3.13: The front-end readout boards (left) and their housing (right). The logic and high voltage board connectors can be seen peeking out of their respective slots in the housing.



Figure 3.14: Front-end tracker electronics readout chain. Courtesy of James Mott.

## Chapter 4

# Track Reconstruction and Analysis

As described in section 3.3, the straw trackers are used to provide information about the muon beam, which is important for the calorimeter  $\omega_a$  analysis, calculating the  $\omega_a$  pitch correction, and determining the spatially weighted magnetic field seen by the muons. The track reconstruction is performed in three stages. First, individual hits in the tracker are grouped into tracks in the finding stage. Second, a best estimate trajectory is fit to grouped hits in the fitting stage. Third, the best fit trajectory is extrapolated back to the storage region or forwards to the calorimeter in the extrapolation stage. A fourth refinement stage is planned but not yet implemented, which would add or remove hits in the finding stage based on the results of the fitting and extrapolation stages. This refinement is expected to improve statistics but not considerably improve performance.

As a brief aside, every stage of the track reconstruction is performed in the event-processing framework known as *art* [68]. The *art* framework is a collection of modularized stages in a C++ framework useful for reading, reconstructing, filtering, analyzing, and writing data, among other things. Most Fermilab experiments now use *art*, including E989.

### 4.1 Track finding

The track finding stage consists of pattern recognition routines in order to group individual hits into separate sets corresponding to individual incident tracks. The general

implementation of these pattern recognition routines is relatively straightforward [69, 70]. Hits across all modules are grouped in time windows called time islands, with an average width of 40 ns and a max width of 100 ns<sup>1</sup>. Within those time islands hits are then grouped into clusters. Clusters consist of one or two hits for each U or V view per module. As a reminder the U and V views of a module consist of the two U or V layers in that module, Section 3.3. Hits are only clustered if they lie in close proximity in time and space to one another. The spatial constraint is defined as the difference in hit straw numbers, from 0 to 31 for the 32 straws per layer, which by default is limited to  $\leq 4$ . Neighboring hit clusters in the same module are then grouped to form seeds, one per module. Finally, seeds are grouped together module by module from one end of the tracker to the other to form what are called track candidates. The seeds are formed and grouped into track candidates again only if they lie close in time and space to one another. The entire track candidate formation process occurs for all hits in a time island to find as many real tracks as possible. See Figure 4.1.

After a track candidate has been formed a number of checks are made before passing it on to the fitting stage. If hits, clusters, or seeds are found to be shared among multiple track candidates, the candidates are dropped. Likewise, a track candidate is dropped if it is made from seeds consisting of only one type of view, or if the track candidate has less than six hits. There are also various small geometry and timing algorithms to improve the track candidates, such as removing hits from secondaries [71]. The  $t_0$  time for the track candidate is calculated as the mean time of all hits, with some fixed offset. The track candidate is supplied with an original momentum and position guess at the start of the track by fitting a circle to the hit straw wires in the horizontal plane. The final track candidates are then passed on to the fitting stage.

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<sup>1</sup>This time scale is set by the characteristic drift time.



Figure 4·1: Hits in a tracker station in a single time island. The black dots indicate the position of the straw wires, while the blue and red points indicate hits. In blue is the first formed track candidate in the island, formed from separate seeds in different modules. The track finding algorithms will move onto the remaining hits in the time island to attempt to form other track candidates, one of which is easily observable by eye.

## 4.2 Track fitting

The track fitting stage takes the track candidates from the track finding stage, and outputs a best fit trajectory to those candidates. This includes optimal state vectors and error matrices for the track at each measurement plane and at a fictitious starting plane at the entrance to the straw tracking detector. The track fitting routines can roughly be split into two parts, error propagation and the actual fitting and improvement of the track. The implementation of these parts go hand in hand, and will be described in turn. Details of the track fitting code itself is described in Reference [72].

### 4.2.1 Error propagation and coordinate systems

The process of error propagation involves taking track parameters and error matrices (which describe the spread in those track parameters) and transporting them along discrete steps from one point to another, accounting for changes due to any present magnetic fields or material along the step paths. There is a set of error propagation routines originally written in Fortran by the EMC collaboration, called “Geometry and Error Propagation” or Geane [73]. Geane works by propagating particles along their average trajectories neglecting the effects of discrete processes, using a helix equation along small enough steps where the change in the magnetic field is small. These routines were used in the E821 experiment as well as the PANDA and FINUDA experiments with some success [74]. The Geane routines were at one point converted to C++ and added to Geant. The strength of using Geane within a Geant simulation lies in its direct access to the Geant geometry and field. This is crucially important for the E989 track fitting because the trackers live in a region of high field non-uniformity. Shown in Figure 4.2 is the location of the tracker with respect to the radial and vertical fields as calculated in Opera2D and included in the E989 Geant4

simulation. As shown the radial field in the tracker region rises from 0 T at the outer ends to roughly 0.3 T at the inner top and bottom ends, and the vertical field drops approximately 50% from the storage dipole field of 1.451 T. These large field gradients over the tracking measurement space must be handled appropriately, which Geane does effectively.

Predicted average track parameters in Geane are a function of path length

$$\tilde{p}_l = F_{l,l_0}(p_0), \quad (4.1)$$

where  $p_0$  are some original tracking parameters and  $\tilde{p}_l$  are the updated average ones. The path length  $l$  can be defined or limited how one wishes, and typically corresponds to a single step in the Geant4 simulation. The track parameter vectors  $p$  are defined in some coordinate system. In the Geane routines these track parameters are  $5 \times 1$  vectors either defined in the “free” (curvilinear) system

$$\frac{1}{p}, \lambda, \phi, y_\perp, z_\perp, \quad (4.2)$$

or the “surface” (detector) system

$$\frac{1}{p}, \frac{p_v}{p_u}, \frac{p_w}{p_u}, v, w. \quad (4.3)$$

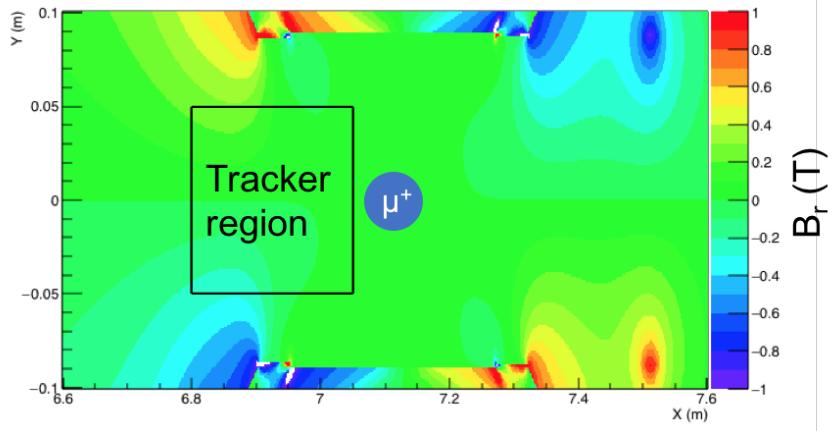
In the free system, the  $\lambda$  and  $\phi$  parameters are the dip ( $\pi/2 - \theta$ ) and azimuthal angles respectively, while the  $y_\perp$  and  $z_\perp$  parameters are defined as being in the  $XY$  or  $XZ$  global Geant4 planes and orthogonal to  $x_\perp$ , where  $x_\perp$  is defined as being along the momentum vector of the particle. See Figure 4.3. In the surface system, the UVW coordinates are defined with any two orthogonal vectors V and W<sup>2</sup>. The surface system is most usefully defined in the tracker reference frame, where the modules are

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<sup>2</sup>For clarification, the UVW surface system has nothing to do with the UV orientations of the straws at this time.



(a) Vertical magnetic field



(b) Radial magnetic field

Figure 4.2: Shown are the vertical (top) and radial (bottom) magnetic fields of the storage ring magnet in and around the storage region as calculated in Opera 2D. The horizontal and vertical axes are the radial and vertical coordinates of the ring respectively. The center of the storage region lies at 7.112 m along the horizontal axis. The contours represent the strengths of the vertical and radial magnetic fields. The black box shows the rough location of the tracker with respect to the ring. It can be seen that there is a large field non-uniformity within the tracker space.

staggered in a local Z coordinate, the local Y coordinate is vertical, and the local X coordinate increases with straw number. See Figure 4.4. The surface system is then defined as

$$\frac{1}{p}, \frac{p_x}{p_z}, \frac{p_y}{p_z}, x, y. \quad (4.4)$$

In both free and surface systems the track is represented by one momentum parameter, two directional parameters, and two position parameters. Needing six independent parameters to describe a particle in space and momentum (three momentum and three position parameters), one parameter is left out and taken as a known variable. For Geane this is taken either as a known path length in the free system, or a known U coordinate in the surface system (or known Z coordinate in our tracker reference frame). In our tracker reference frame, the 32 straw layers corresponding to a tracking station are defined at known local Z coordinates. The path lengths for steps in Geane can be set to be equal to the distance for a track to travel between between detector planes, and therefore the track parameter dependence on the path length can instead be replaced by a dependence on plane number. The number of degrees of freedom per track is the number of measurement planes it hits,  $N$ , minus 5 for the number of track parameters.

The  $5 \times 5$  error matrix on a plane calculated in Geane describing the expected distribution in true parameters about the average ones is defined as

$$\sigma_N^{ij} = \langle p_N^i p_N^j \rangle - \langle p_N^i \rangle \cdot \langle p_N^j \rangle, \quad (4.5)$$

where i and j are track parameter indices,  $N$  is some plane number, and  $p_N$  are ‘true’ track parameters on plane  $N$ . This error matrix will include effects from multiple scattering, delta ray production, ionization, and bremsstrahlung [73, 74, 75]. These matrices are transported from plane to plane by what are called transport matrices,



Figure 4.3: Free (left) and surface (right) tracking coordinate systems [74].



Figure 4.4: Shown is a model view of a tracker station in relation to the magnetic ring. Tracker modules are shown in white. Around the tracker measurement area is defined a coordinate system called the tracker reference frame. In that frame, the X coordinate is directed outward along the straws nearly radially, the Y coordinate is directed vertically up, and the Z coordinate is directed along the direction that the tracker modules are staggered.

where the  $5 \times 5$  transport matrix elements between two planes are defined as

$$T_{N,N-1}^{i,j} = \frac{\partial \tilde{p}_N^i}{\partial p_{N-1}^j}. \quad (4.6)$$

The transport matrix  $T$  is a Jacobian between planes which expresses the infinitesimal changes in parameters at some plane (or path length) with respect to the parameters at some previous plane (or previous path length):

$$\delta \tilde{p}_N = T_{N,N-1} \delta p_{N-1} \quad (4.7)$$

Note that the transport matrix does not propagate the track parameters themselves as does an equation of motion. The error matrix is propagated forward from one plane to another by

$$\sigma_N = T_{N,N-1} \sigma_{N-1} T_{N,N-1}^T + \sigma_{\text{material}}, \quad (4.8)$$

where  $\sigma_{\text{material}}$  is the added error due to material effects between the planes. See Figure 4·5. The calculation of the transport matrices themselves is done within the Geane routines in the free system on a step by step basis, where the derivation of the transport matrix elements is given in Reference [76]. It should briefly be pointed out that the transport matrix between any two planes (or number of steps) is the multiple of all intermediate transport matrices,

$$T_{N,N-2} = T_{N,N-1} T_{N-1,N-2}, \quad (4.9)$$

regardless of what reference system the matrices are defined in (as long as they are all consistent). Geane can convert the transport matrices between the free system and the surface system using further Jacobians, also derived in Reference [76]. When



Figure 4.5: Transport and error matrices for straw tracker planes. Transport matrices are defined between straw planes, and error matrices are defined on the planes.

converting a transport matrix from one reference system to another,

$$T_{N,N-1}^s = A_N T_{N,N-1}^f A_{N-1}^{-1}, \quad (4.10)$$

where the  $s$  and  $f$  superscripts stand for the surface and free reference systems respectively, and  $A$  is the Jacobian between reference frames which is defined at a specific point or plane ( $A_N \neq A_{N-1}$ ). The error matrices are converted between reference frames in the usual way,

$$\sigma_N^s = A_N \sigma_N^f A_N^T. \quad (4.11)$$

Finally, while the tracker reference frame is nominally defined in the local XYZ coordinates as described previously, the straws themselves don't measure in that frame directly. As described in Section 3.3, the straws measure drift circles in planes perpendicular to the straws themselves. The measurements from U and V straws therefore lie on the U and V measurement axes shown in Figure 4.6, where the measurement of the drift circle is instead taken as a U or V coordinate to the left or right of the straw wire. To first order the U or V coordinate is the DCA of the hit, which can be corrected with the angle of the track to get a better estimate, as shown in Appendix A.1. It's important to note that out of the five track parameters each straw only measures a single U or V position. The new coordinate system is defined

as

$$\frac{1}{p}, \frac{p_u}{p_z}, \frac{p_v}{p_z}, u, v, \quad (4.12)$$

where this  $Z$  variable is the tracker reference frame  $Z$ , and the U and V coordinates here are non-orthogonal and different to those in Equation 4.3. The transformation between the  $XYZ$  and  $UVZ$  systems is given by

$$p^{UV} = J_5 p^{XY} \quad (4.13)$$

where  $J_5$  is a  $5 \times 5$  matrix defined by

$$J_5 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & J_2 & 0 \\ 0 & 0 & J_2 \end{pmatrix} \quad (4.14)$$

and  $J_2$  is a  $2 \times 2$  matrix given by

$$\begin{pmatrix} u \\ v \end{pmatrix} = J_2 \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}. \quad (4.15)$$

$J_2$  is readily determined by inspection of Figure 4.6. In order to transform the transport or error matrices from the tracker reference frame to the tracker measurement frame, the same relations as in Equations 4.10 and 4.11 apply,

$$T_{N,N-1}^{UV} = J_5 T_{N,N-1}^{XY} J_5^{-1} \quad (4.16)$$

$$\sigma_N^{UV} = J_5 \sigma_N^{XY} J_5^T \quad (4.17)$$

where the superscripts of  $XY$  or  $UV$  identify which coordinate system the objects belong to.



Figure 4.6: The straw tracker measurement reference system. The XYZ system here is the straw tracker reference frame.  $\theta$  is the same angle as the stereo angle of the straws, at  $7.5^\circ$ . U straws measure along the U axis and V straws measure along the V axis.

#### 4.2.2 $\chi^2$ minimization

The method for fitting and improving the track is a global  $\chi^2$  minimization algorithm that uses the transport and error matrices as described previously [73, 77]. This straightforward global fitting algorithm works because of the minimal amount of material contained within the tracker and the resulting small correlations between planes. For denser detectors with more scattering and therefore greater correlations, other fitting algorithms such as a Kalman filter should be used [74]. The following derivations and minimization assume measurements on planes in the tracker measurement frame described by Equation 4.12, but it should be noted that the results apply to any reference frame. A derivation for a  $\chi^2$  including no material correlations is presented followed by one which includes material correlations.

The  $\chi^2$  for a track is defined as the residuals between predicted and measured parameters on a measurement plane, divided by their errors, summed over all hit planes:

$$\chi^2 = \sum_{i=1}^N [(\tilde{p}_i(p_s) - x_i)^T (\sigma_i^{-1})(\tilde{p}_i(p_s) - x_i)] \quad (4.18)$$

$x_i$  are vectors of the measured track parameters on plane  $i$ ,  $\tilde{p}_i$  are vectors of the average predicted track parameters which stem from some general starting parameters  $p_s$ , and  $\sigma_i$  are the  $5 \times 5$  error matrices on the planes. To first order the error matrices consist only of the measurement errors on the U and V parameters and exclude the effects of random material processes. These errors are located in the U and V diagonal elements (3, 3) and (4, 4) respectively, with corresponding resolutions of approximately  $150 \mu\text{m}$  as described in Section 3.3. At second order the material error matrices as calculated by Geane are added to the measurement errors. Because the measured parameters consist of solely U or V measurements, the  $x_i$  vectors are  $5 \times 1$  objects where only the (3) or (4) elements have any meaning respectively<sup>3</sup>. The errors on the non-measured parameters in the diagonals of the error matrix are taken as infinite, such that when the error matrix is inverted all corresponding rows and columns of the final matrix calculation on each plane reduce to zero and contribute nothing to the  $\chi^2$ .

By minimizing this  $\chi^2$  with respect to the starting parameters  $p_s$ , and evaluating it at the target best starting guesses  $p'_0$ , which are the parameters of interest, the track can be fit:

$$\begin{aligned} \frac{\partial \chi^2}{\partial p_s} \Big|_{p_s=p'_0} = 0 &= \sum_{i=1}^N \left[ \left( \frac{\partial \tilde{p}_i(p_s)}{\partial p_s} \Big|_{p_s=p'_0} \right)^T (\sigma_i^{-1}) (\tilde{p}_i(p'_0) - x_i) \right. \\ &\quad + (\tilde{p}_i(p'_0) - x_i)^T \left( \frac{\partial (\sigma_i^{-1})}{\partial p_s} \Big|_{p=p'_0} \right) (\tilde{p}_i(p'_0) - x_i) \\ &\quad \left. + (\tilde{p}_i(p'_0) - x_i)^T (\sigma_i^{-1}) \left( \frac{\partial \tilde{p}_i(p_s)}{\partial p_s} \Big|_{p_s=p'_0} \right) \right] \end{aligned} \quad (4.19)$$

The middle term is small and can be neglected assuming that the error matrix doesn't change much with respect to the choice of starting parameters. This is true as the part of the error matrix that depends on the starting parameters is already

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<sup>3</sup>A straw tracker module as a whole can be approximated as measuring in 2D space, but this leads to correlations between measured parameters which must be taken into account, as compared to the natural tracker measurement frame in 1D space of U or V for which there are no measurement correlations [78].

small due to the low amount of material in the tracker. In tandem the error matrix doesn't change much from one fitting iteration to the next as long as the path length through the material remains about the same. The first and third terms are identical in value, and so must therefore both separately be equal to zero. Equation 4.19 is therefore reduced to

$$0 = \sum_{i=1}^N T_{i0}^T \sigma_i^{-1} (\tilde{p}_i(p'_0) - x_i), \quad (4.20)$$

where  $T_{i0}$  is the transport matrix between the point at which the starting parameters are defined and plane  $i$  given by Equation 4.6:

$$T_{i0} = \frac{\partial \tilde{p}_i(p_s)}{\partial p_s} \Big|_{p_s=p'_0} \quad (4.21)$$

In minimizing the  $\chi^2$  the desire is to update some original set of starting track parameters  $p_0$  to the new best ones  $p'_0$ . This difference,  $\Delta p_0$ , can be determined by substituting the following into Equation 4.20,

$$\tilde{p}_i(p'_0) = \tilde{p}_i(p_0) + T_{i0}\Delta p_0, \quad (4.22)$$

which follows from Equation 4.7. After simplifying one arrives at

$$\Delta p_0 = \sigma_{p_0} \sum_{i=1}^N T_{i0}^T (\sigma_i^{-1}) (x_i - \tilde{p}_i(p_0)), \quad (4.23)$$

where

$$\sigma_{p_0} = \left[ \sum_{i=1}^N T_{i0}^T (\sigma_i^{-1}) T_{i0} \right]^{-1}. \quad (4.24)$$

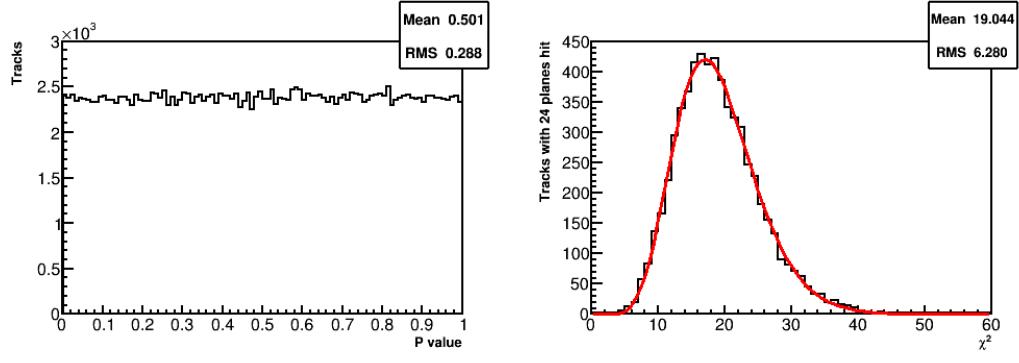
$\sigma_{p_0}$  is a  $5 \times 5$  covariance matrix of the starting fit parameters, where the diagonals describe the fit errors in the 5 track parameters at that point.

To summarize, an initial set of starting parameters  $p_0$  are propagated forwards

in Geane to produce predicted track parameters, transport matrices, and error matrices. These objects along with the measured parameters are plugged into the  $\chi^2$  minimization algorithm detailed here to provide a  $\chi^2$  describing the goodness of the fit corresponding to those original starting parameters, an improvement  $\Delta p_0$  to those starting track parameters, and the errors  $\sigma_{p_0}$  on those starting parameters. This consists of a single iteration of the track fitting. In order to determine the predicted parameters of the track corresponding to the improved starting parameters, the error propagation part of the procedure needs to be repeated. The track fitting is iterated until the  $\chi^2$  no longer improves, at which point the track fitting is said to have converged. Typically three or four iterations are enough to get a best fit track. As a reminder the initial set of starting parameters is given by a circle fit to the hit straws as described at the end of Section 4.1. The starting parameters for a track are defined on a virtual 0 plane parallel to the measurement planes, where the placement of the 0 plane is defined based on a track by track basis and is placed at a point 1 cm in front of the first straw tracker module that was hit. Note that there is remarkable robustness with respect to the initial starting parameters in fitting the track. Of course if the initial starting parameters are too poor, then the fit will not converge.

#### 4.2.3 Fits to idealized tracks in vacuum

The tracking algorithm was built and tested in the full E989 Geant4 simulation in *art*. Hits were generated from decay positrons from muons in the storage region. In the initial tests of the track fitting material was turned off and the measured hits were defined as the truth hits with some known Gaussian smearing. (The truth hits are accessible within the Geant4 simulation by defined “dummy plane” detectors which record hits at the straw measurement planes.) Plots showing the goodness-of-fit for the fitted tracks are shown in Figure 4.7. Beyond the goodness-of-fit plots, the other measure of how good the track fitting is performing is to plot the truth pulls of the fit



(a) p-value distribution for all tracks. (b)  $\chi^2$  distribution for tracks which hit 24 planes.

Figure 4.7: Goodness-of-fit distributions for fitted tracks in vacuum, with Gaussian measurement errors and no material effects. The p-value distribution is flat, and the  $\chi^2$  distribution matches a normalized  $\chi^2$  pdf for 19 degrees of freedom which is overlayed in red. ( $\chi^2$  distributions for tracks which hit other numbers of planes are very similar.)

parameters. The truth pulls are defined as the residual between the fitted parameter and the truth parameter, divided by the fit error on that parameter:

$$\frac{\Delta p_0^i}{\sigma_{p_0}^{ii}} = \frac{p_{0,\text{fit}}^i - p_{0,\text{true}}^i}{\sigma_{p_0}^{ii}} \quad (4.25)$$

Since the  $\chi^2$  minimization returns fit parameters and errors on the starting plane, this is where the truth pulls are defined. Plots of the truth pulls for the five track parameters are shown in Figure 4.8, where each pull is a unit Gaussian as they should be for idealized results.

#### 4.2.4 Material correlations

While the trackers consist of relatively little material, it is desirable to include the full effects of material for improved tracking results. Random processes due to material contribute to the error matrix in Equation 4.5 as described in Section 4.2.1. The random scattering of a particle trajectory at one plane means that there is an extra



Figure 4.8: Truth pulls for the five fitted track parameters at the starting plane of the track, for tracks in vacuum with Gaussian measurement smearing and no material effects. The plots are shown on a log scale, and each are fit to a Gaussian in red. Each fit is consistent to a unit Gaussian with a mean of zero and an RMS of one, showing that the track fitting is working properly.

correlated error in all further planes, see Figure 4.9. Equation 4.18 does not take into account these material correlations between measurement planes when fitting the track. While it provides a decent approximation of the best fit track in the low material tracker, the  $\chi^2$  distribution is noticeably off as shown in Figure 4.10. To calculate a better estimate of the trajectory, a more general version of the  $\chi^2$  equation is used:

$$\chi^2 = (\vec{p} - \vec{x})^T (\sigma^{-1}) (\vec{p} - \vec{x}) \quad (4.26)$$

Here  $\vec{x}$  and  $\vec{p}$  are a  $5N \times 1$  vectors of the measured and predicted track parameters respectively, where  $N$  is the number of planes hit, and these objects are the combined vectors of the  $5 \times 1$  counterparts. Similarly,  $\sigma$  is a  $5N \times 5N$  matrix, where the  $5 \times 5$  diagonal block matrices are the individual plane error matrices described before, calculated between plane 0 and  $N$ . Calculating the  $\chi^2$  is now re-casted from a sum over measurement planes into a single large linear algebra equation. At this point both calculations of the  $\chi^2$ 's are equivalent.

The new format however allows for the material correlations between planes to be included, where these correlations are added as  $5 \times 5$  matrices in the off-diagonal blocks of the new large error matrix. The upper diagonals are given by<sup>4</sup>

$$\sigma_{MN} = T_{MN}\sigma_N, \quad (4.27)$$

where  $\sigma_{MN}$  is the material correlation matrix between plane  $M$  and plane  $N$ ,  $T_{MN}$  is the transport matrix between the two planes, and  $\sigma_N$  is the ordinary material error matrix as calculated from the 0 starting plane. See Appendix A.2 for the derivation of Equation 4.27. The  $\chi^2$  is minimized in the same way as was done in the previous section such that the improvement to the starting track parameters  $\Delta p_0$  remains a

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<sup>4</sup>The lower diagonals are just the transpose.



Figure 4·9: RMS error between the true track position and the average track position as a function of distance through the tracker. This error increases as a particle passes through more and more material. Each black point indicates the location of a straw measurement layer. If a track goes through the whole detector, on average the true position has an extra uncertainty of nearly 200  $\mu\text{m}$  compared to the average one.

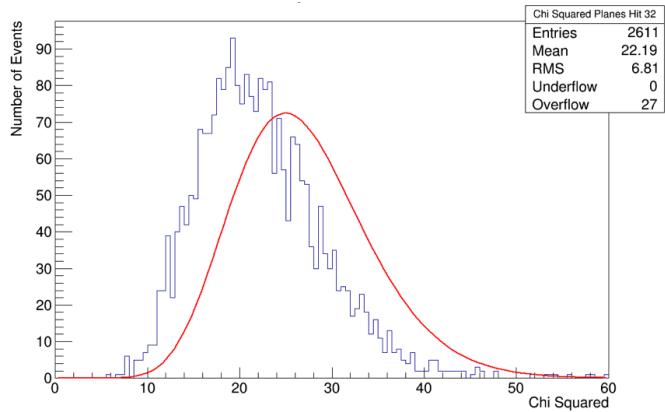


Figure 4·10: The  $\chi^2$  distribution for fitted tracks including material effects in the error matrices but excluding correlations in the calculation. The  $\chi^2$  distribution is noticeably different from the correct curve in red.

$5 \times 1$  vector and is given by

$$\Delta p_0 = \sigma_{p_0} \tau^T \sigma^{-1} (\vec{x} - \vec{p}), \quad (4.28)$$

$$\sigma_{p_0} = [\tau^T \sigma^{-1} \tau]^{-1}, \quad (4.29)$$

where  $\tau$  is a  $5N \times 5$  object of the individual transport matrices combined together.

Because  $\sigma$  is such a large matrix,  $5N \times 5N$  where  $N$  ranges from 6 to 32, inverting it is a slow process. The tracking must have a certain speed for the data to be efficiently processed and fit, which makes these inversions unfeasible. Note however that the diagonal errors of infinity values for non-measured parameters would reduce all the corresponding rows and columns to 0 after the inversion, in the same way as described before. This fact can be taken advantage of by removing said rows and columns that would contribute nothing to the  $\chi^2$  anyway, and thus reducing the  $5N \times 5N$  size to  $N \times N$ . The corresponding rows and columns of the unmeasured parameters in the combined transport matrix  $\tau$  and residual vector are also removed, resulting in an  $N \times 5$  matrix for  $\tau$  and an  $N \times 1$  vector for the residuals. The covariance matrix  $\sigma_{p_0}$  remains a  $5 \times 5$  matrix. This improves the speed of the  $\chi^2$  calculation dramatically, while leaving the final calculation unaffected<sup>5</sup>. All pieces of the  $\chi^2$  calculation and minimization is done with a C++ linear algebra library optimized for speed called Eigen3 [79]. The tracking time is now dominated by the error propagation in Geant4, not the linear algebra.

#### 4.2.5 Fits to simulated tracks including material effects

Once the material correlations are properly included, the  $\chi^2$  distribution is repaired, as shown in Figure 4.11. The plots in this section show the results of the track

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<sup>5</sup>Note that these element removals are done just before the final calculation of the  $\chi^2$  and fit to the track and not at the beginning of the algebra, otherwise the plane material correlations are not properly included.

fitting in the full  $g - 2$  Geant4 simulation with material effects turned on. Truth measurements with 150  $\mu\text{m}$  Gaussian smearing are once again used as the measured hits, and a cut of 3 MeV on the true simulated energy loss is made to remove kinked tracks which would result in poor fits from hard energy loss physics processes. The comparison between the simulated and reconstructed energy loss for fitted tracks is shown in Figure 4.12. While there is a noticeable difference, the energy loss in general is small so it is acceptable. Truth pulls for the tracks are shown in Figure 4.13. It can be seen that there is a slight spread in results due to the material effects. This is to be expected given the non-Gaussian nature of the processes, and the vast majority of tracks still fit well. The number of iterations it takes to fit a track is shown in Figure 4.14. The number of planes a track hits and the corresponding momentum dependence is shown in Figure 4.15. The total momentum distribution and residuals to truth are shown in Figure 4.16. The track fitting has a momentum resolution of approximately 2% with just a slight dependence on the momentum of the track.

After the track fitting has determined the best fit parameters in the  $UV$  space, the returned parameters can be turned back into the tracker reference frame coordinates. Plots for the fitted vertical and horizontal momenta, positions, and corresponding residuals are shown in Figures 4.17 and 4.18. The forward momenta plots are omitted since the majority of the total momenta is in the forward direction. More extensive plots than what is shown here can be found in Reference [72].



(a) p-value distribution for all tracks. (b)  $\chi^2$  distribution for tracks which hit 10 planes.

Figure 4.11: Goodness-of-fit distributions for fitted tracks in the full  $g - 2$  Geant4 simulation with material effects included. The p-value distribution is flat, and the  $\chi^2$  distribution matches a normalized  $\chi^2$  pdf for 5 degrees of freedom which is overlayed in red.

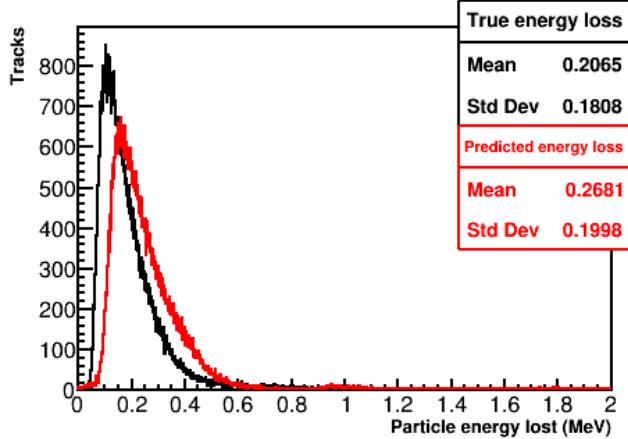


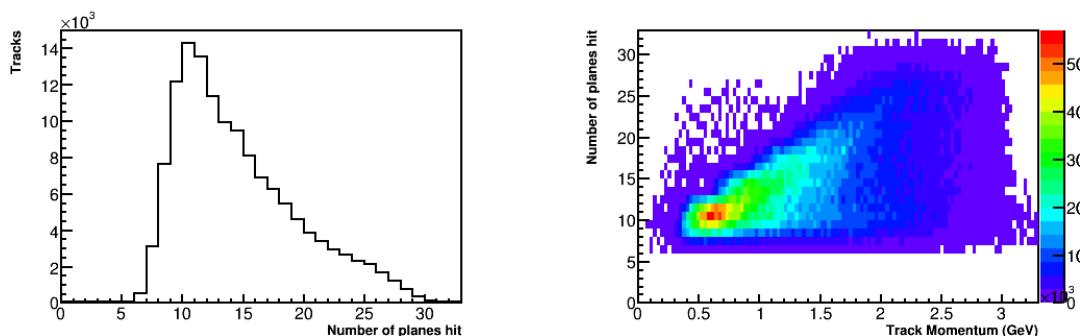
Figure 4.12: Simulated true energy loss (black) vs Geant4 predicted energy loss (red) for fitted tracks. As shown there is a mismatch between the two. This is acceptable as the energy loss is in general very small compared to the total momentum of each track,  $200\text{keV} \ll 2\text{GeV}$ . Sources of energy loss come from ionization and bremsstrahlung processes, which account for the long Landau tail running off to infinity. Originally the Geant4 physics calculations were taking too much energy away due to bremsstrahlung processes in our low material tracker, so the energy loss calculations were modified slightly [72].



Figure 4.13: Truth pulls for the five fitted track parameters at the starting plane of the track for fitted tracks in the full  $g - 2$  Geant4 simulation with material effects included. The plots are shown on a log scale, and each are fit to a Gaussian in red. Each fit is close to a unit Gaussian with a mean of zero and an RMS of 1, but there are tracks which lie outside the Gaussian due to material effects and imperfect resulting fits.



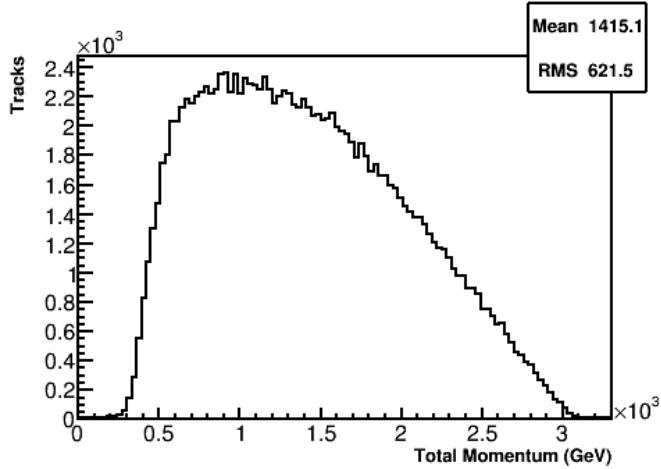
Figure 4.14: Number of iterations for the track fitting to converge per track. The track fitting does not take less than three iterations, and the number of iterations is capped at ten.



(a) The number of planes hit peaks at 10, and falls off to the maximum number of planes at 32.

(b) The number of planes hit vs the momentum of the track. Tracks with higher momentum in general hit more planes.

Figure 4.15: Number of planes hit per track (left) and the momentum dependence (right).



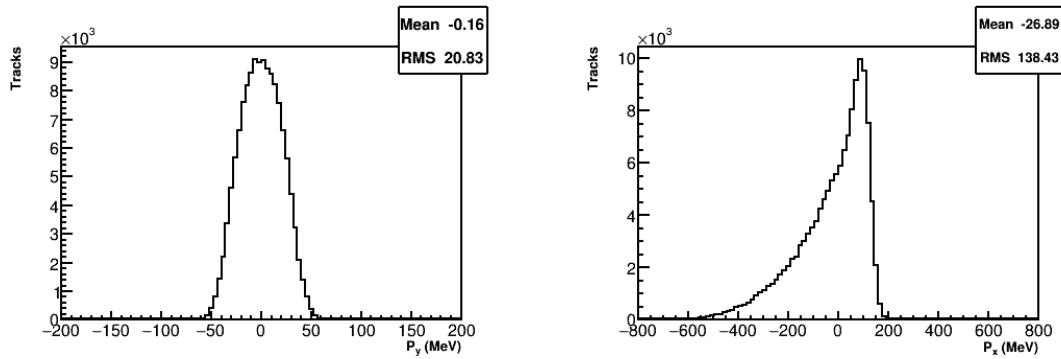
(a) The fitted momentum distribution for tracks from decay positrons which tails off at approximately the magic momentum of 3.094 GeV.



(b) The residual between the reconstructed and true total momentum. The RMS is approximately 40 MeV, though there are some tails which spread out from the distribution.

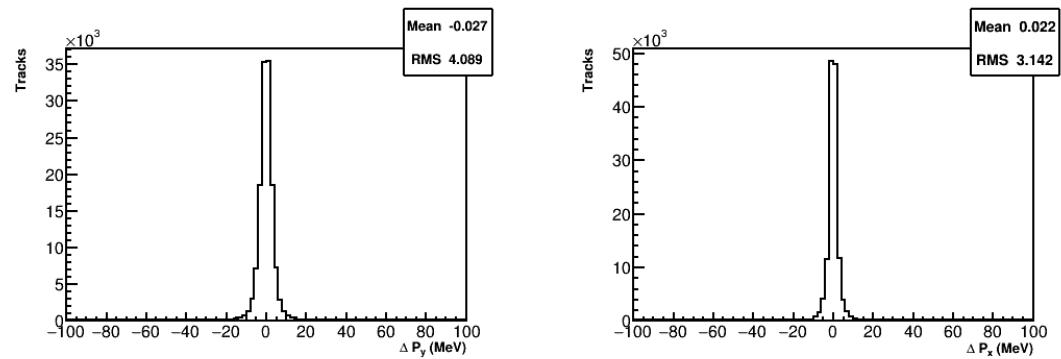
(c) The relative residual between the reconstructed and true total momentum. The RMS is approximately 2.4 %. This plot includes tracks of all momenta; in general the resolution of the total momentum is around 2% for all energies.

Figure 4.16: Fitted track momentum distribution and corresponding residuals to truth.



(a) Vertical momentum distribution. This distribution is bounded by the vertical angle acceptance of the tracker.

(b) Horizontal momentum distribution. The shape of the distribution is due to the acceptance of the tracker. It is perhaps more informative to look at the radial momentum distribution, which is shown in Figure 4.21.



(c) The vertical momentum residual, which has a width of about 4 MeV.

(d) The horizontal momentum residual, which has a width of about 3 MeV.

Figure 4.17: Shown are the fitted vertical and horizontal momentum components (top) for tracks at the entrance to the tracker, and their residuals to truth (bottom).



Figure 4.18: Shown are the fitted vertical and horizontal position components (top) for tracks at the entrance to the tracker, and their residuals to truth (bottom).

#### 4.2.6 Left-right ambiguity and fit modes

Before actual data can be fitted, the left-right ambiguity problem needs to be dealt with. Since straws only measure drift circles and not actual U or V positions, and since tracks are in general passing forwards through the detector, there is an ambiguity as to which side of the wire a particle passed for each hit. In general if even a single left-right choice for a hit is wrong then a track will fit poorly. There are a couple of different track fitting modes used to deal with this problem, which are detailed in Reference [72] and summarized here and in Reference [80].

##### `wireFit`

The first fit to any set of tracks in data is to do a fit to the wire centers, which requires no left-right or drift-time information. The errors on the measured hits are set as the RMS of a uniform distribution with a width of a straw diameter. After a wire fit, an approximation of the best fit track is acquired which provides some left-right information. The number of fitting iterations is capped at three for the wire fit.

##### `mainFit`

The primary fit mode for analyzing data is called `mainFit`. After the wire fit is completed, the predicted track parameters at each measurement plane are compared to the locations of the wires for the straws that were hit. The left-right choices for each hit are set depending on which side of the wire the predicted track passed. The measured hit positions and errors are updated from the wire values to the angular-corrected DCA values based on the the left-right choices of each hit, and the fitting is repeated. At each iteration of the fitting, the left-right choices for each hit are updated based on where the previous predicted track parameters ended up. This is the fastest fit mode and with this method about 66% of tracks fit well. The advantages

in processing speed outweigh the lost track fraction as the tracking analyses are not statistics limited.

### `fullSeqFit`

The secondary fit mode is called `fullSeqFit`. This fit mode does a better job of determining the left-right choices for the hits in a track, but is much slower than `mainFit`. Here is given a short summary, see Reference [72] for more technical details.

After a wire fit, the geometry of the straw layers is used to resolve some of the left-right choices for a hit on a track. The left-right ambiguity is partially resolved through the shift in layers for each view as detailed in Section 3.3, where if a particle passes straight through both layers in a view it can be seen to go to the left of one wire and the right of the other. This implies that for curved tracks with certain angles through tracker modules, the left-right choices for some hits are known precisely [81]. For these hit planes the left-right choices are locked in and taken as known for the rest of the fitting.

For the remaining hits where the left-right choices are unknown, an approximate  $\chi^2$  is calculated for each combination of left-right choices,  $2^{N_{\text{unknown}}}$ . This  $\chi^2$  is calculated using the same formula as in the full fitting with the measured parameters set to the different left-right combinations, Equation 4.26, but using the Geane transport and error matrices calculated by the wire fit instead of from a fit to the left-right choice measured parameters. Calculating this  $\chi^2$  for each combination is slow, so this process is sped up by only checking the left-right combinations for the U or V hit layers at a time respectively,  $2^{N_{\text{U,unknown}}} + 2^{N_{\text{V,unknown}}} \ll 2^{N_{\text{unknown}}}$ . This  $\chi^2$  calculation allows for a measure of how good a left-right combination is to be determined without having to propagate every track in Geane for the full fitting. The tracks with the best combinations and smallest approximate  $\chi^2$ 's are then fit with the full Geane propagation and  $\chi^2$  minimization algorithm. Typically the best 10 to 15 combinations

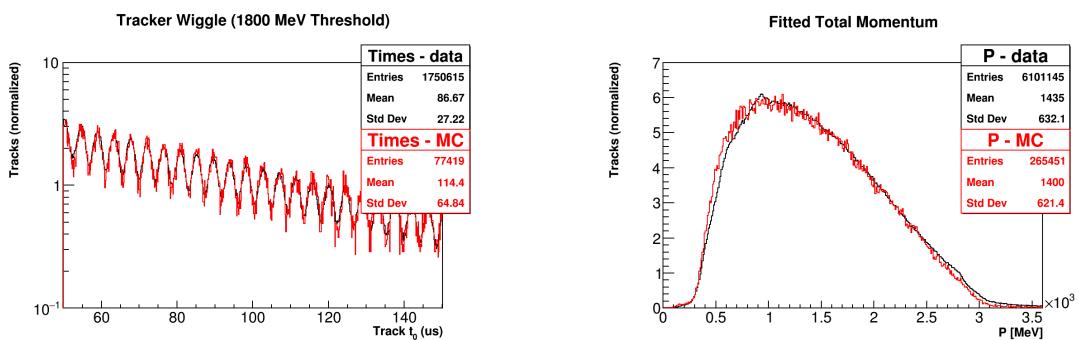
are fit, and the final track with the best  $\chi^2$  is taken as the best fit track. This gets the left-right choices for the fitted tracks right about 99% of the time with simulated data, but is very slow due to the combination checking and the Geane propagation for each of the best track left-right combinations. If the track fitting can be sped up significantly or a majority of the left-right choices constrained upstream somehow, then this fit mode would become the default.

#### **4.2.7 Fits to tracks from data and comparison with Monte-Carlo**

Fits to data are done with `mainFit`. Due to imperfect left-right assignments and the use of real DCA and DCA error measurements, the tracks from data are naturally more messy. See Figure 4·19 for a sample fitted track and a plot of the goodness-of-fit for the tracks. While not all tracks are fit perfectly, enough of them are fit well in order to pass on the results to the extrapolation stage. Fitted tracks to Run 1 data are compared to results from Monte-Carlo. Track times and fitted total momenta are shown in Figure 4·20. Fitted track radial and vertical momentum and position distributions are shown in Figure 4·21. In general the results are very similar between the two. There are mismatches in the distributions because the idealized simulation did not reflect the real run conditions at the time of this track fitting analysis. As described in Section 2.8 and explored later in Section 4.4, the muon beam was imperfectly stored in the ring leading to the measured distributions here.



Figure 4.19: Sample fitted track (left) and p-value distribution (right) in data. A cut has been made in the p-value distribution at 1% to remove tracks which have entirely failed the fitting. The distribution can be seen to rise towards zero where tracks have been fit but imperfectly.



(a) Track times for tracks with energy greater than 1.8 GeV. The  $g - 2$  frequency can be seen in both the data and simulation.

(b) Fitted track total momentum, there is a very slight mismatch between data and Monte-Carlo.

Figure 4.20: Fitted track results in data (black) vs Monte-Carlo (red). The number of entries in each are normalized to each other so that they can be compared. Shape differences between the two are due to a mismatch between simulation conditions and the real experiment, and not any problem with the track fitting.



Figure 4.21: Fitted track results in data (black) vs Monte-Carlo (red). The amount of entries in each are normalized to each other so that they can be compared. Shape differences between the two are due to a mismatch between simulation conditions and the real experiment, and not any problem with the track fitting.

### 4.3 Track extrapolation

The last stage of the track reconstruction is the track extrapolation. The extrapolation takes the fitted track results and either extrapolates them back to the storage region to the approximate position of the muon decay point, or forwards to the face of the calorimeter sitting right behind the tracker. The extrapolation stage utilizes a fourth order Runge-Kutta Nyström algorithm [63] which discretely steps a trajectory through the magnetic field in the full  $g - 2$  Geant4 simulation, similar in some respects to the error propagation. At each step of the extrapolation, the updated track position and covariance matrix are compared to physical volumes in the simulation to flag tracks which have been reconstructed as likely originating from outside the storage region [63, 82]. Because there is no fixed interaction point in the storage region, tracks are extrapolated backwards to the point of tangency where the radial momentum is equal to zero. Studies were done to verify that this approximation for the muon decay point was sufficient using Monte-Carlo, and it was found that a simple 1.1 mm correction to the radial decay position could be applied regardless of the momentum of the track [63]. A schematic diagram explaining this behavior is shown in Figure 4.22. The vertical extrapolated distribution was found to have no biases. A birds eye view for tracks extrapolated back into the storage ring is shown in Figure 4.23.

### 4.4 Muon beam measurements

All stages of the track reconstruction have ultimately lead to the goal of determining the characteristics of the muon beam. As described in Section 2.6, the muon beam is not a fixed entity. Where the muons live and how they move in the storage region ultimately affect the calculation of the E-field and pitch corrections, the weighting of the muon distribution with respect to the measured field, and the calorimeter preces-



Figure 4.22: Tracks are extrapolated backwards into the storage region to the point of radial tangency. While for a single decay positron the extrapolated muon decay point will be different than the truth, the degeneracy of inwards and outwards decays leads to a correct average extrapolated position. A radial offset of 1.1 mm needs to be applied as shown. High and low momentum tracks are emitted at smaller and larger angles respectively. Because high momentum tracks travel farther than low momentum tracks before reaching radial tangency, the offset is momentum-independent on average. Plot courtesy of Saskia Charity [63].

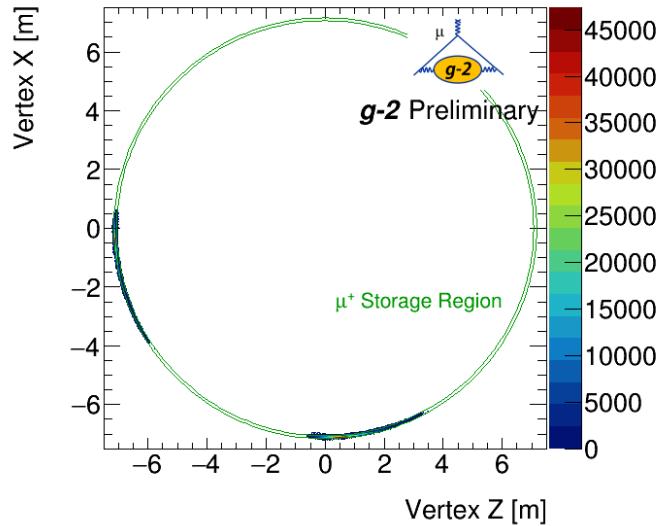


Figure 4.23: A birds eye view of the extrapolation results in the storage ring. The two distributions of extrapolated tracks can be seen at the left and bottom of the figure, where the trackers sit at the head of each distribution. It can be seen that some tracks are extrapolated multiple meters back through the storage region.

sion frequency analysis. Presented here are plots for the 60h dataset which describe the muon beam in the storage region, for those muons which decayed to positrons which then passed through the tracking detectors. There is still on-going work in regards to analyzing the acceptance differences between the tracker and calorimeter, in order to make sure that the correct muon beam dynamics are described in regards to those positrons which hit the calorimeters and are responsible for the measurement of  $\omega_a$ . Tracker-calorimeter matching studies will not be detailed here. A preliminary analysis is given in Reference [67]. A set of quality cuts are applied to each track in order to remove failed or poorly fitted tracks. By applying these cuts a clean sample of the extrapolated muon decay positions is extracted without biasing the results. The following conservative cuts are applied [83, 84]:

- Non-failed track or vertex
- No material hit during extrapolation back to vertex
- Number of tracking planes hit  $\geq 12$
- p-value  $> 5\%$
- Vertical extrapolation uncertainty:  $0.5 < \sigma_y < 3.5$  mm
- Radial extrapolation uncertainty:  $0.5 < \sigma_r < 5$  mm
- Horizontal entrance into tracker:  $60 < X < 150$  mm
- Vertical entrance into tracker:  $-40 < Y < 40$  mm
- Drift time:  $0 < t_d < 70$  ns
- Track UV residuals  $< 500$   $\mu$ m
- Fraction of missed planes  $< 30\%$

- $|\text{Number of U hits} - \text{number of V hits}| \leq 4$

First, a radial slice of the extrapolated beam distribution is shown in Figure 4·24. As shown the beam is off-center due to the lower than nominal kicker voltage used in Run 1 as described in Section 2.8. The radial and vertical projections of this cross-section are shown in Figure 4·25. The vertical distribution is very similar to that from the track fitting results, while the radial distribution of the stored muons has a non-Gaussian shape due to the storage procedure of the muons. Because the beam is moving in time and these distributions include tracks of all times, these two distributions in general are wider than they actually are at a single point in time. Table 4.1 gives the measured values of the beam parameters corresponding to the various datasets for all times. Figure 4·26 shows the radial and vertical distributions of the muon beam as a function of time in-fill, with the means plotted in Figure 4·27. Because of the bad quadrupole resistors as described in Section 2.8, the beam was stored at a greater vertical position than the center of the storage ring, and slowly moved downwards over the course of a fill. Similarly, the vertical width of the beam was seen to decrease slowly over the course of a fill. These trends are shown in Figure 4·28.

The observed oscillation in the radial mean is the CBO as described in Section 2.6.1. Using the tracking results shown here, the CBO frequency and amplitude can be characterized [86]. Because of the bad quadrupole resistors, not only does the average beam vertical position change over the course of a fill, but the CBO frequency changes as well. The CBO frequency and amplitude as a function of time in-fill are shown in Figure 4·29. It can be seen that over a course of a fill, the CBO amplitude and frequency decay away exponentially (where the frequency asymptotes to its static value). Both of these characteristics of the beam motion are folded into the precession frequency analysis as will be described in Chapter 5.

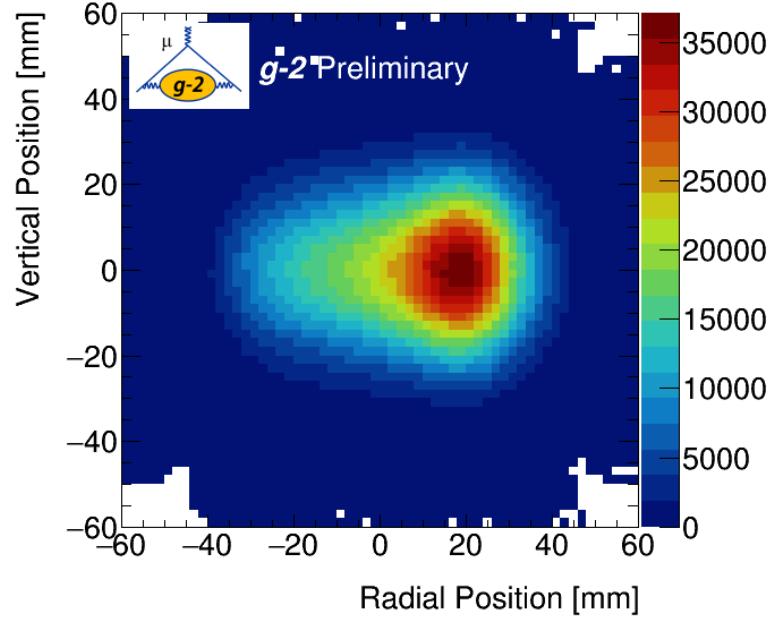
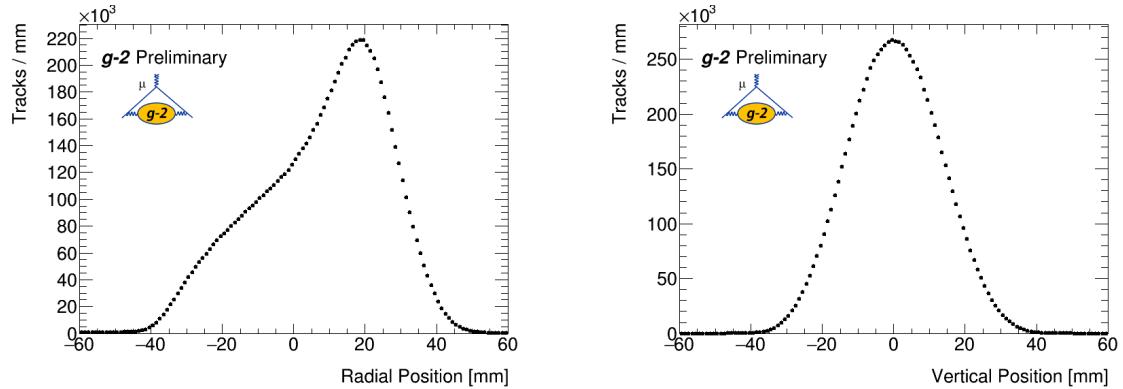


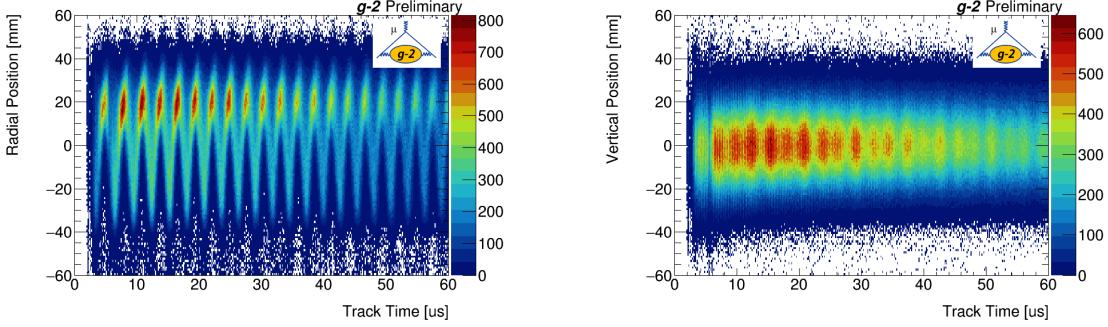
Figure 4.24: Shown is a radial slice of the extrapolated muon distribution or beam spot. The beam is localized off center of the storage ring due to the kicker settings used, Section 2.8.



(a) The radial position of 0 mm corresponds to the magic radius of 7.112 m. The radial distribution is peaked radial outward from the center.

(b) The vertical distribution looks closely centered in the storage region, which has a width of 90 mm.

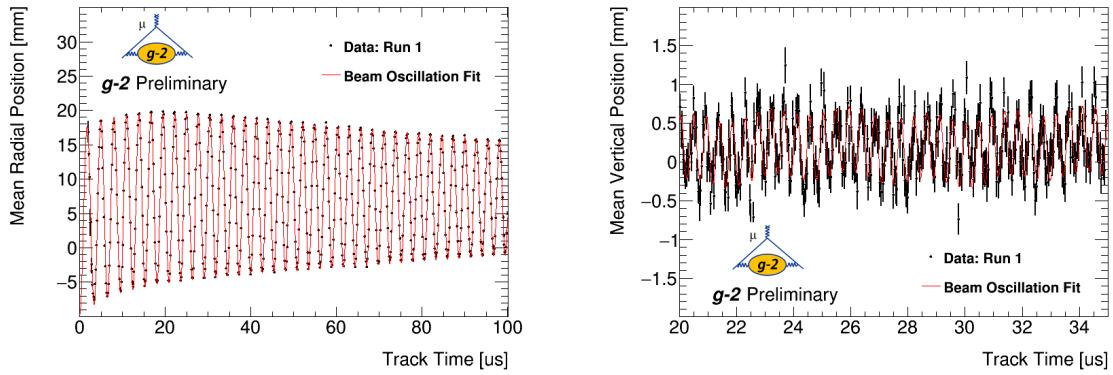
Figure 4.25: Shown are the muon beam radial (left) and vertical (right) distributions for all track times.



(a) The radial distribution of the muons can be seen to both oscillate radially inward and outward as a function of time, and bunch and spread out as a function of time. This behavior is the CBO of the beam.

(b) The vertical distribution of the muons is relatively stable as a function of time, but there is an oscillation that can be seen.

Figure 4.26: Shown are the muon beam radial (left) and vertical (right) distributions as a function of time in-fill for tracks in station 12. In general the distributions for station 18 are very similar.



(a) The radial distribution mean oscillates at the CBO frequency. It can be seen to rise at early times in station 12 due to the scraping procedure, and then slowly relax away once the scraping is turned off and the muons decay away.

(b) The vertical distribution mean oscillates at the vertical betatron frequency, as described in Table 2.2. As shown for these early track times the beam is actually stored slightly higher than the center of the storage region.

Figure 4.27: Shown are the muon beam radial (left) and vertical (right) distribution means as a function of time in-fill for tracks in station 12. In general the distributions for station 18 are very similar.

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**Run 1 Dataset Muon Beam Parameters**


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	Radial			Vertical	
Dataset	Mean	Peak	Width	Mean	Width
60h	7.61	18.41	18.43	0.15	12.88
HighKick	6.89	15.95	17.74	0.11	12.92
9d	8.03	19.31	18.37	0.40	12.71
Endgame	7.40	18.11	18.01	0.95	12.68

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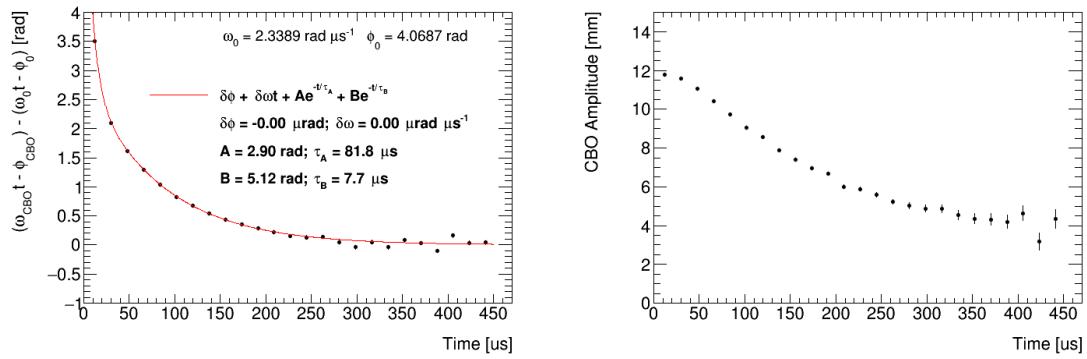
Table 4.1: Muon beam widths and means for the Run 1 precession frequency analysis datasets. Both radial and vertical values are defined with respect to the center of the storage region. The values listed here are from the distributions including tracks in both tracker stations. The radial peak values were determined by fitting Gaussians around the peaks of the respective radial distributions. All values are in units of mm.



(a) Average vertical position.

(b) Average vertical width.

Figure 4.28: Average vertical position (left) and width (right) as a function of time in-fill. The data points are binned such that the higher frequency betatron motion is binned out. Stations 12 and 18 are shown in the blue and red respectively. The sets of points correspond to the means and widths directly from histograms at specific time points. Gradients are calculated by fitting the points to a straight line from 50 to 350 μs. The vertical position of the beam drops over the course of the fill on the order of 1 mm/ms, depending on dataset, while the vertical width decreases on the order of 0.5 mm/ms. The differences between the two stations are due to the closed orbit. Plots courtesy of Joe Price. Fits were also done to means and widths calculated from Gaussian fits to the individual histograms with similar results [85].



(a) The CBO frequency change is due to the slowly changing quad voltages over each fill. Plotted is the difference between the cosine argument for the changing frequency and the value it asymptotes to at late times. The frequency fit model includes two exponentials, one with a longer time constant and one with a shorter time constant.

(b) The CBO effect decays away exponentially with a lifetime on the order of 200 – 300 μs. Reasonable CBO envelopes include a pure exponential or an exponential plus a constant.

Figure 4.29: CBO frequency (left) and amplitude (right) as a function of time in-fill.

## Chapter 5

# $\omega_a$ Measurement

The measurement of  $\omega_a$  is determined by counting the number of detected positrons in the calorimeters above some energy threshold versus time, as described in Section 2.1. Doing so results in a histogram of counts that is modulated by  $\omega_a$ , Figure 2.4. Fitting for the frequency allows  $\omega_a$  to be extracted. The  $\omega_a$  analysis therefore consists of the steps needed to construct the histogram of counts, the fitting of that histogram, and the necessary systematic error studies.

### 5.1 Reconstruction of decay positron hits

The calorimeters measure hit times and energies of incident particles, where these hit times and energies are determined from the raw SiPM signals and a reconstruction procedure. In E989 there are two overall separate reconstruction algorithms, `ReconWest` and `ReconEast`, both written in the *art* framework similar to the tracking reconstruction. Each of these reconstruction algorithms is modularized, and the steps of the reconstruction process can be switched in and out at will. Using separate reconstruction methods gives confidence in any final results by removing single points of failure. The reconstruction method used in this analysis is `ReconWest`. A summary of its details will be presented here. A more thorough description is detailed by A. Fienberg [87].

The raw data are digitized waveforms, which are voltage traces from each SiPM for each calorimeter crystal. Due to the incredible amount of data coming in with the

high sampling rate, only those pulses which exceed some threshold are saved to disk. An online processing system checks the traces against this pre-configured threshold by passing all of the data through a GPU farm [88]. If any trace is found above threshold, then the data are saved from every SiPM in the hit calorimeter, for a time range around the over-threshold trace. This time range is called a time island, similar to that in the tracking reconstruction, and typically has a width of 40 ns [87].

The traces are fit with templates in order to extract the area and peak times of any present pulses. Each SiPM has its own templates, one for positrons and one for laser pulses. These templates are extracted from data, where each template is determined by collecting many single pulse traces from a SiPM, normalizing by pulse area, aligning in time, and averaging them. These templates were checked against many systematic effects in order to make sure that the constructed templates did not bias the energy or time measurements, such as hit angle, energy (pulse size), position, and rate, as well as aging effects [60, 87]. Each trace is fit using a  $\chi^2$  minimization algorithm with the corresponding SiPM templates in order to determine the time and energy of the hit. In order to fit for multiple pulses in a single time island, the fitting procedure first fits with a single template, and then checks the residuals for any remaining peaks. If peaks exist above some threshold, then the fitting is repeated until all pulses have been fit. The time measurement performance in the pulse finding was found to be unaffected by the number of pulses in a time island, and there is 100% pileup separation at 5 ns. See Figure 5.1 for a typical single template fit to a SiPM trace.

Once a pulse has been fit with a template, the pulse area needs to be converted to real energy units using an energy calibration procedure. Several different techniques exist that can be used, including a method that counts photo-statistics seen in the SiPMs. The default method used is a comparison of lost muon energy signatures in

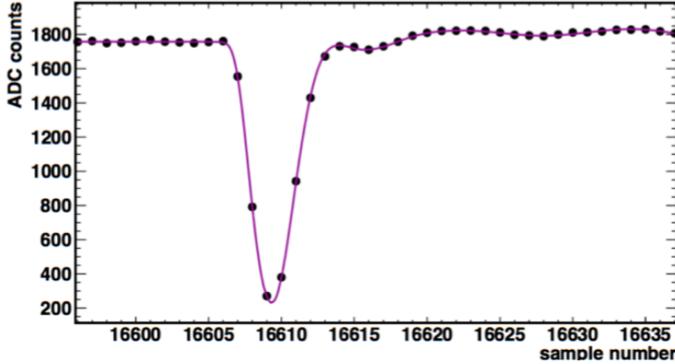


Figure 5.1: A template fit in purple to a SiPM trace delineated by the black points which is in units of ADC counts [87].

the calorimeters. As described in Reference [89], muons lost from the storage ring can spiral inward and hit consecutive calorimeters with a specific time separation between calorimeter hits. These lost muons are minimum-ionizing particles, and thus leave a low energy signature in the crystals, see Section 5.3.3. Selecting on the time signature allows hits corresponding to lost muons to be isolated, and the energy signature can be used to determine the appropriate conversions from area to energy<sup>1</sup>.

The energy calibration for positron hits as compared to lost muon hits then needs to be determined. Again there are several different techniques, including a comparison of endpoint energies for high energy positrons which tail off at the magic momentum of 3.094 GeV, and comparison with simulation. The default technique is to calibrate the energies such that the optimal energy threshold for the  $\omega_a$  analysis is near 1.7 GeV. Ultimately the accuracy of the energy calibration is relatively unimportant because it is not the energy units that really matter. Rather, the quantity of interest is the number of positrons above some energy threshold, where that threshold can be optimized empirically. Indeed the entire  $\omega_a$  analysis could be done only considering the area of the SiPM pulses as a proxy for the energy of the incident positrons.

Each pulse now has an associated energy and time. The measurement of  $\omega_a$  de-

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<sup>1</sup>Different channels can also be equalized based on the energy signatures.



Figure 5.2: In-fill gain function fit for a single calorimeter crystal (top) and fit residuals (bottom). Each crystal has its own in-fill and STDP gain function parameters. Plot courtesy of Matthias Smith.

pends heavily on the time reconstruction since the analysis is a frequency extraction, thus the pulse times must be corrected for various effects in order to reach the precision goal. The fitted times for each pulse need to be aligned on a fill-by-fill basis relative to the injection time of the beam, corrected for any channel differences due to differing pulse shapes or optical fiber lengths, and corrected for any calorimeter time misalignments due to the use of different laser system components. The fill-by-fill alignment is corrected using the T0 detector as described in Section 3.1.1. The channel differences are corrected by aligning calorimeter channels in time using signals from islands with large simultaneous pulses in neighboring crystals. Calorimeters are time aligned using lost muon coincident events as described before. Once the times of the pulse fits or crystal hits have been determined, the energies can be corrected appropriately for gain effects measured by the laser system. As described in Section 3.2.3, the laser calibration system corrects for out-of fill, STDP, and in-fill gain effects, in that order. Figure 5.2 shows an in-fill gain function fit to data for a single calorimeter. Systematic effects for corrected gain effects are studied in Section 5.5.2.

The last part of the calorimeter reconstruction is the clustering. Clustering is the stage which takes the individual template fit results from separate crystals, and turns them into the times, energies, and positions of decay positron impacts. For a time island with a single positron impact, the procedure is straightforward. The energy for the positron hit cluster is the sum of the individual hit crystal energies. The time for the cluster is taken as the time of the maximum energy hit in the island. This works because most of the deposited energy from a hit is localized to a single crystal. The position of the cluster is determined with a logarithmic weighting function between crystal hits, which for a 2 GeV positron in the E989 calorimeters results in a resolution of 2 mm [87]. See Figure 5.3 for a single calorimeter cluster from a positron hit in the calorimeter. For a time island with multiple positron impacts, the individual crystal hits are separated in time, where the time partitioning separates hits that are 2.5 ns apart, and the clustering proceeds as before. For hits which are within this time window, a pileup event has occurred. If the pileup event happens within the same crystal, then the multiple hits are measured as a single hit, and this needs to be corrected for using a pileup subtraction technique, as described in Section 5.2.1. For hits that occur in separate crystals, the pileup can be resolved using the spatial separation of the calorimeters. This is an ongoing area of work, and one technique is described in Reference [87]. For this analysis the spatial separation was disabled. This increases the amount of pileup seen in the data, which then needs to be handled by the pileup subtraction technique. For the precision of the Run 1 analysis result, this was found to be acceptable.

## 5.2 Construction of positron hit energy and time spectra

Once the reconstruction has processed all calorimeter hits into clusters, the energy and time spectra histograms are made. At the very last stage of the reconstruction

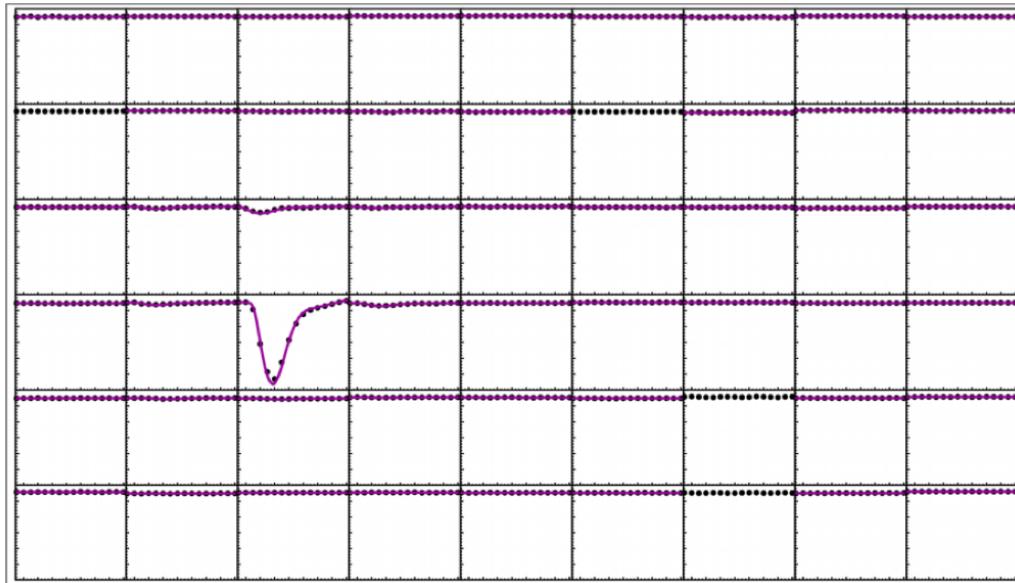


Figure 5·3: A single positron hit in the calorimeter, which resulted in a reconstructed calorimeter cluster [87]. Each box is a crystal in the calorimeter, where the contained trace is the SiPM output fit with a template. The positron hit the crystal three from the left and three from the bottom, where it deposited most of its energy. Some of the energy was deposited in the neighboring crystals.

Time Spectra Parameters	
Parameter	Value
Energy threshold ( $E_{th}$ )	1700 MeV
Bin width ( $T_c$ )	149.2 ns
Artificial dead time (ADT)	5 ns
Shadow dead time (SDT)	5 ns
Shadow gap time (SGT)	10 ns
Pileup energy scaling (C)	1
$g - 2$ period ( $T_a$ ) in Ratio Method	4.365 411 $\mu$ s
Muon lifetime ( $\tau_\mu$ ) in Ratio Method	64.44 $\mu$ s

Table 5.1: Parameter values used in the construction of  $\omega_a$  time spectra.

procedure, an *art* module takes the produced clusters and puts them into **ROOT TTree** formats, where individual data members include the energies, times, calorimeter numbers, etc. of the individual clusters. There is of order 20,000–140,000 cluster data files per dataset, which are combined down to order 200–1,400 **ROOT TTree** files. These **ROOT TTrees** are then passed through a **ROOT** macro to produce **ROOT** files with the histograms defined by the **TH1F** class, one **ROOT** histogram file per tree file.

It should be noted that some of the parameter choices for the constructed histograms were informed by analysis results. All analysis parameters were chosen to be identical between the distinct analyzed datasets, in order to simplify both the comparison and combination of different dataset results. This section describes the justification for the different histogram parameters chosen. A table of the histogram parameters is shown in Table 5.1.

Energy and time histograms are made for each individual calorimeter. These are summed together to form histograms of all hit times and energies. Figure 5.4 shows a sample energy spectrum for the Endgame dataset. An energy threshold is applied to the clusters before filling the time histograms. As described at the end of Section 2.1, the optimal energy threshold is where the quantity  $NA^2$  reaches the

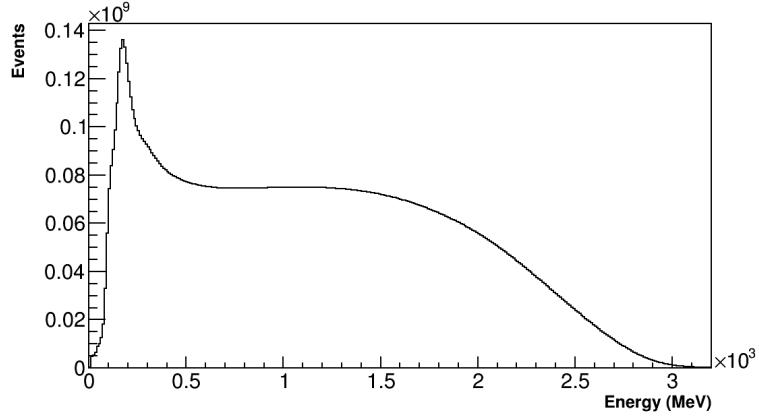


Figure 5.4: Energy spectrum for hits with cluster times greater than 25  $\mu$ s in all calorimeters on a linear scale. The peak at about 170 MeV corresponds to lost muons layed on top of the low energy decay positron part of the spectrum. Data are from the Endgame dataset.

maximum, at least in the case of a five parameter fit<sup>2</sup>. By scanning over the choice of energy threshold and fitting the resulting time spectra with Equation 2.13, the optimal energy threshold can be determined as seen in Figure 5.5. The optimal choice of energy threshold was determined to be 1700 MeV, in accordance with the cluster reconstruction energy calibration.

The optimal bin width for the time histograms was determined to be 149.2 ns, the peak of the cyclotron period distribution determined from fast rotation analyses to the data<sup>3</sup>. As described in Section 2.6.2, this bin width combined with a time randomization on each cluster over a range of  $\pm T_c/2 = 149.2 \text{ ns}/2$  serves to eliminate the fast rotation signal in the data<sup>4</sup>. This randomization is done using ROOT's TRandom3 class. As will be described in Section 5.3.2, the cluster times are also randomized

<sup>2</sup>Using the final fit function and looking at the error directly on the fitted  $\omega_a$  frequency, a slightly better estimate can be found.

<sup>3</sup>A change in the bin width of order 0.1 ns has very little effect on the extracted  $\omega_a$  frequency as seen in Section 5.5.6.

<sup>4</sup>An alternative approach is to randomize all times in a single fill by half the cyclotron period as opposed to each individual pulse.

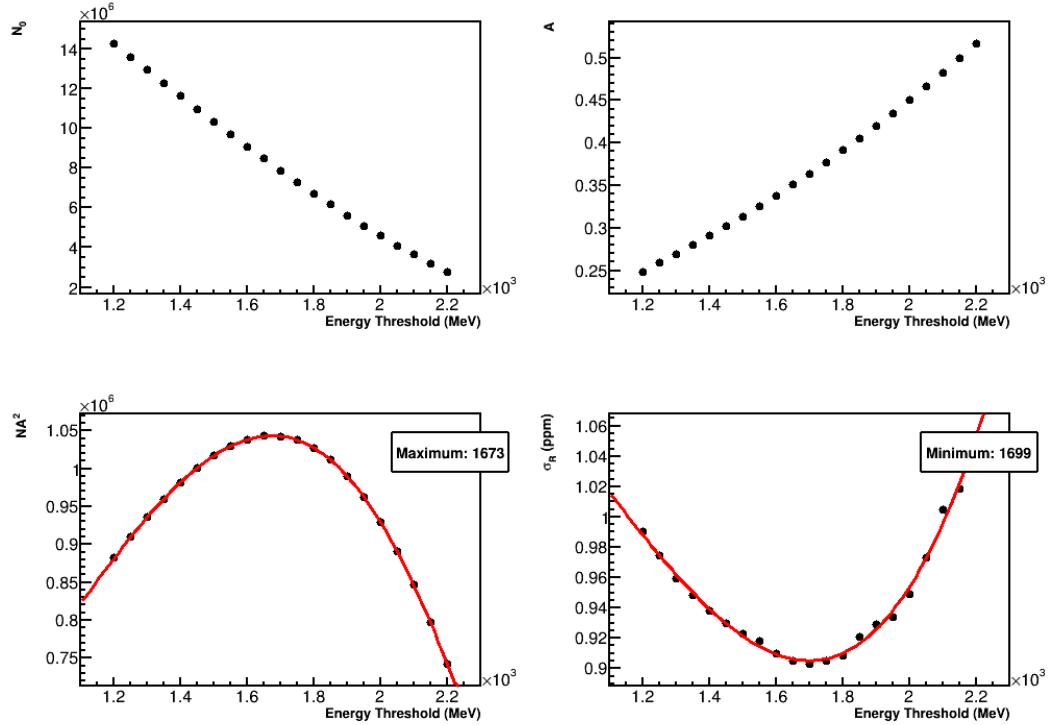


Figure 5.5: The optimal energy threshold can be determined from the  $NA^2$  quantity as described in Section 2.1 from five parameter fits to the data (bottom-left) or from the calculated error with the final fit function (bottom-right). Fitted  $N$  and  $A$  parameters are also shown (top) which satisfy Equations 2.14 and 2.15. The maximum and minimum are determined using seven parameter polynomial fits to the relevant quantities. The optimal threshold varies slightly per calorimeter and per dataset. Since the region of the minimum is relatively flat, a single energy threshold of 1700 MeV was chosen. Data are from the 9d dataset.

by the half the vertical waist period,  $\pm T_{VW}/2$ , where  $T_{VW}$  is determined from the inverse of the VW frequency, Equation 2.26. Putting the frequency just in terms of the  $n$  value and the cyclotron frequency  $f_c$  the VW period is given by

$$T_{VW} = \frac{1}{f_{VW}} = \frac{1}{(1 - 2\sqrt{n}) \cdot f_c}. \quad (5.1)$$

This randomization is done in order to remove the effects of the VW in the data<sup>5</sup>. The default random seed for each histogram ROOT file is the hash of the unique input file name using C++’s standard hash class. Other random seeds can be chosen in order to verify the consistency of fit results. Histograms are defined with a time range of 0–699.8972  $\mu\text{s}$  (the closest integer multiple of the bin width to 700  $\mu\text{s}$ ), corresponding to 4691 bins.

### 5.2.1 Pileup subtraction

As described in Section 3.2, there will be a certain amount of pileup in the detectors. As a reminder pileup is the term for when multiple particles hit a calorimeter within the dead time of the detector such that they are registered as a single hit or cluster. The measured energy and time spectra for all observed clusters will include this pileup distortion. For the time histogram of clusters above energy threshold, the number of counts will be wrong for cases where two below-threshold particles are registered as a single cluster above threshold, and where two above-energy threshold particles are registered as a single cluster. In the former, an extra count is added into the histogram, and in the latter a count is missed. The case where two lower energy positrons are registered as a single higher energy cluster will have a different  $g - 2$  phase than an actual single cluster at the same energy. This is because the lower energy positrons on average decay from muons which have traveled further around the

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<sup>5</sup>Even though the VW frequency was found to be changing over the course of the fill, the change is small enough such that this constant time randomization was found to remove all residual traces of the VW in the data.

ring, due to acceptance effects. These muons which have traveled further around the ring have spent more time in the magnetic field, and thus their spins have precessed more. See Figure 5.6. Clusters which originate from pileup events therefore have a different  $g - 2$  phase than non-pileup events.

If pileup was a constant effect in time, then the phase of the time histogram would be shifted by some constant amount, and the  $\omega_a$  frequency extracted would be unaffected. However, the rate of pileup in the detectors changes over the time of a fill, as muons decay away. The lifetime of double pileup events in the detectors, where the word double indicates cases where two hits are registered as a single cluster, will go approximately as half the lifetime of single hit events<sup>6</sup>, and similarly for triple and higher orders of the pileup effect. Because the rate of hits in the detectors oscillates at the  $g - 2$  frequency, pileup will increase and decrease accordingly leading to oscillations in the pileup time spectra at  $\omega_a$  and  $2\omega_a$ . This time-dependent distortion means that the pileup effect needs to be included in the fit function or subtracted out of the data in order to extract the correct  $\omega_a$  frequency. The former is challenging due to the non-linear nature of the dead time of the detectors, and would include another phase in the argument of the cosine term in the fit function, thus worsening the statistical precision of the extracted  $\omega_a$  frequency. All analyzers thus construct an approximate pileup spectrum and then subtract it from the data before fitting.

There are various methods to construct pileup spectra which are then subtracted from the main time and energy spectra. The method used in this analysis is called the ‘asymmetric shadow method’, used successfully in E821 [90]. This method statistically constructs an approximation of the pileup spectrum from the data by assuming that the probability of observing a pileup pulse is the same as the probability that two pulses will be offset by some small amount of time, such as 10 ns. The method works by looking in time windows after trigger pulses to see if a ‘shadow’ pulse exists.

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<sup>6</sup>It is not exactly half when including the non-linear dead time of the detectors

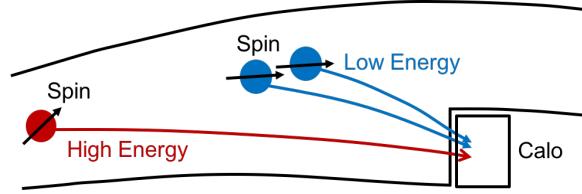


Figure 5·6: Pileup example, where two low energy positrons are registered as a single high energy positron. The black arrows indicate the (exaggerated) direction of the muon spins at the time of decay. Because of acceptance effects the lower energy decay positrons typically come from muons which have traveled further around the ring, and thus the muon spins have precessed more in the magnetic field, leading to a different measured  $g - 2$  phase for pileup events.

If such a pulse exists, then a shadow doublet is created, see Figure 5·7. The pileup spectrum  $P$  is then the sum of the shadow doublets minus the singlets used in their construction,  $P = D - S$ . The width of the time window, and the time offset from the trigger pulse to the window, are called the shadow dead time (SDT) and shadow gap time (SGT) respectively. The times and energies of the constructed pileup doublets are taken as

$$E_{\text{doublet}} = C \cdot (E_1 + E_2), \quad (5.2)$$

$$t_{\text{doublet}} = \frac{t_1 \cdot E_1 + (t_2 - SGT) \cdot E_2}{E_1 + E_2}, \quad (5.3)$$

where the energy of the doublet is the sum of the two singlet pulses  $E_{1,2}$  times some calibration constant  $C$ , with a default value of 1, and the time of the doublet is the energy-weighted time of the two singlets  $t_{1,2}$ . The procedure for constructing the pileup spectra is as follows:

- Put each hit into a vector corresponding to a specific fill and a specific calorimeter
- Time order the hits

- Loop through the hits; for each hit look within a window of width SDT a time SGT later to see if a shadow pulse exists
- If a shadow pulse exists, construct a shadow doublet with energies and times as defined in Equations 5.2 and 5.3
- Randomize  $t_{\text{doublet}}$  over the range  $\pm T_c/2$  (to remove fast rotation as before, Section 2.6.2)
- Per calorimeter, construct pileup energy and time spectra as  $P = D - S$ , where  $D$  is the sum of doublets and  $S$  is the sum of singlets used in the construction of the doublets, with the times of the singlets set as  $t_{\text{doublet}}$ ; when constructing the pileup time spectra, only include those doublets and singlets above the energy threshold

Thus pileup energy and time spectra are constructed for each calorimeter, which can then be subtracted off the calorimeter cluster energy and time histograms. When combining the calorimeter data, the individual pileup histograms are added together before subtracting them from the calorimeter sum histograms.

In order to produce an estimate of the pileup spectra which best matches the data, an artificial dead time (ADT) is applied to the data before time randomization. This is done because the true dead time of the detectors depends on the energies and spatial separation of the incoming hits. While this slightly increases pileup, by applying an artificial deadtime and matching the shadow window time, the pileup estimation is improved and the overall systematic error is reduced. The construction of the artificial pileup is handled in the same way as the construction of the shadow pileup, with SGT set to 0 ns. The constructed artificial doublets replace the singlets in the data. The value for the ADT and SDT is set at 5 ns, the time threshold at which pileup is 100% resolved.

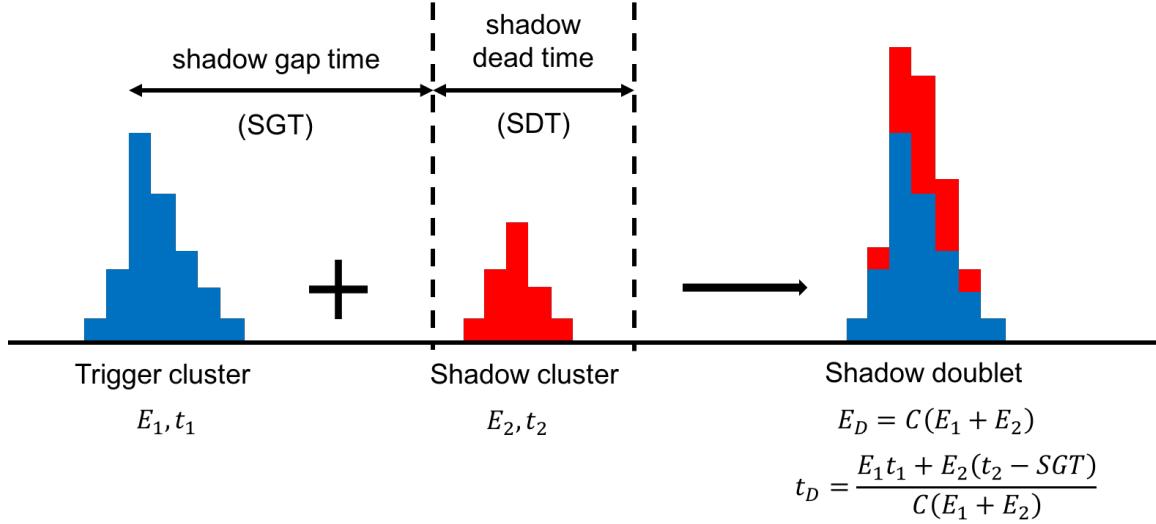


Figure 5·7: The shadow pileup method looks for shadow clusters within a time window SDT, a small time away (SGT) from trigger clusters. If a shadow cluster is found, an artificial doublet is formed and included in the pileup spectra if it exceeds the chosen energy threshold.

**potentially long way from text reference**

The value of the SGT is simply set to twice the SDT, in order to push the shadow window out to times well beyond the dead time of any pileup events, but not so far that an appreciable fraction of muons have decayed. The value of the doublet energy scaling factor C is set to 1, which is a fine approximation as the spatial separation in the reconstruction is turned off<sup>7</sup>. The values for each pileup parameter were shown previously in Table 5.1. See Section 5.5.1 for systematic studies on the effect on  $\omega_a$  due to these chosen parameters.

The pileup energy spectrum and the cluster energy spectrum are shown in Figure 5·8. In general, the two lobes starting at approximately 3 GeV and 6 GeV consist of double and triple pileup events respectively<sup>8</sup>. It can be seen that the shadow method of pileup construction produces a pileup energy spectra which is a good ap-

<sup>7</sup>With the spatial separation turned off, ‘pileup’ events can occur in crystals that are easily separated by eye. While this increases the level of pileup seen in the data, the pileup approximation method also does not consider the spatial separation, and thus handles the level of pileup accordingly.

<sup>8</sup>All orders of pileup fill out the whole energy range, but certain regions consist of mostly one order.

proximation of the cluster energies above the end point of a single decay positron, for cases of double and even triple pileup. The shape difference arises from two factors. First, the shadow method is only written to construct doublets, and does not consider cases of triple or higher orders of pileup. Second, the real pileup in the data contaminates the construction of the shadow pileup spectra, such that a shadow doublet can be constructed from real pileup pulses. While this alleviates the triplet problem slightly, it means that the doublet pileup spectrum is slightly wrong. The corrected energy spectra (cluster energies minus pileup energies), can be seen in Figure 5.9. The shape mismatch is even more apparent as the corrected energy spectrum is high for energies above the expected tail of the true energy distribution, and then goes negative before tailing off to zero.

In order to produce a slightly better estimate of the pileup, a multiplier can be applied to the pileup energy and time spectra. By taking the ratio of cluster energies over pileup energies and fitting the region where the energies are dominated by real pileup doublets, a correction factor of approximately 3-4% is found, as shown in Figure 5.10. The default multipliers for the Run 1 datasets are 1.03210, 1.03413, 1.03387, and 1.03819 for the 60h, HighKick, 9d, and Endgame datasets respectively. Similarly, the cluster times can be examined for cluster energies over 3500 MeV, where the clusters consist purely of pileup pulses, Figure 5.11. By taking the ratio of the pileup corrected times over all times, the level of residual pileup can be determined. Just as in the ratio of the energies, an approximate 3-4% factor is found. When applying this multiplier, the cluster times above 3.5 GeV can be seen to be eliminated as in Figure 5.11. As will be shown in Section 5.5.1, the scale of this multiplier is well within  $1\sigma$  of the pileup multiplier error. The final pileup time spectrum for those pileup pulses above energy threshold is shown in Figure 5.12.

It has been determined that regardless of any residual shape mismatch in the

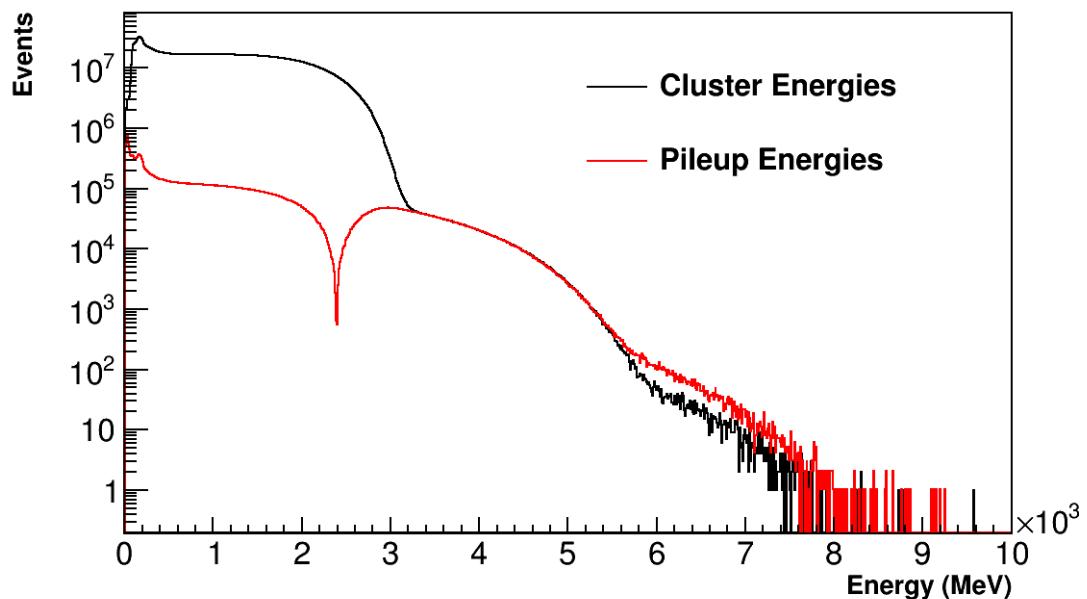
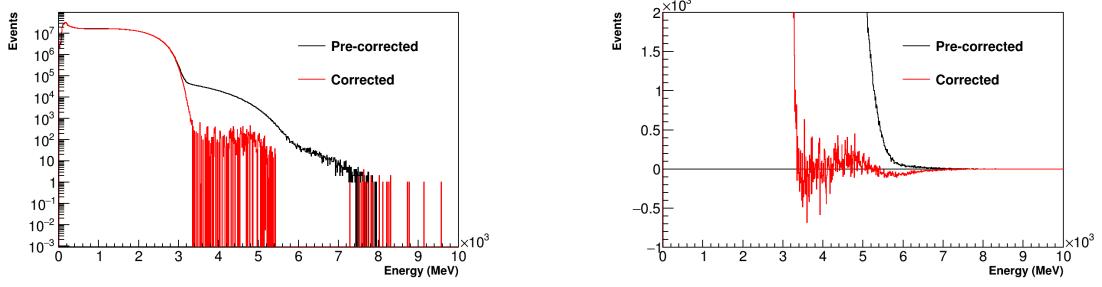


Figure 5.8: Cluster energies in black are plotted vs pileup energies in red, for all calorimeters added together, plotted on a log scale. At energies below about 2.4 GeV the pileup energy spectrum goes negative. In this plot the absolute value of the pileup energies is plotted, and a spike at about 2.4 GeV can be seen as a consequence of this. The shapes do not match perfectly for the constructed pileup spectra, which can be seen at high energies. Data are from the 60h dataset.



(a) Log scale - the corrected energy spectrum goes negative around 5 GeV.

(b) Linear scale - zoomed in to show the shape.

Figure 5.9: Plots for the pre-corrected and corrected energy spectra are shown, all calorimeters added together. Because the triplets and contamination are not accounted for, the corrected energy spectrum does not lie exactly along zero above the energy response of the detectors. Data are from the 60h dataset.

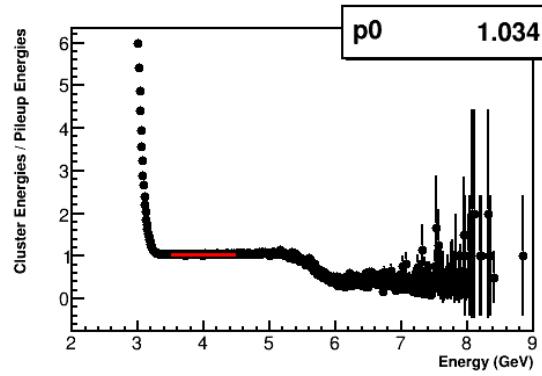


Figure 5.10: Cluster energies divided by pileup energies. A region from 3500–4500 MeV is fit to a straight line, where the doublets dominate the energy distribution. Data are from the 9d dataset.

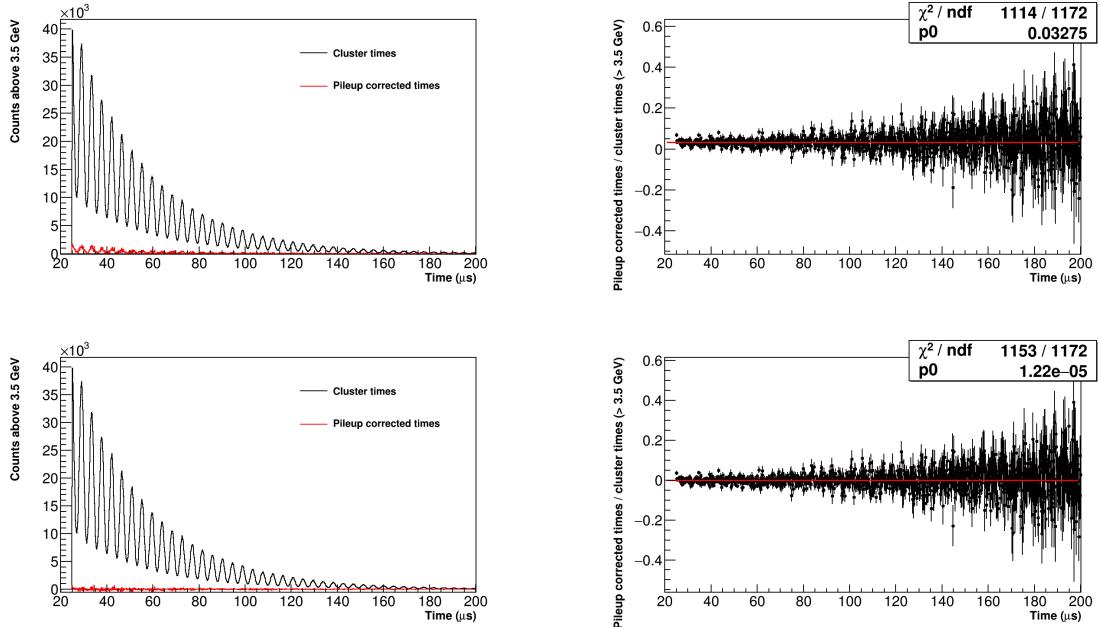


Figure 5.11: Cluster times and pileup corrected times for counts above 3.5 GeV (left) and their ratio (right). The top two plots are used to determine the approximate level of residual pileup left in the data, coming out to about 3%. The bottom two plots show the application of that factor and the resulting removal of the remaining pileup. Data are from the 9d dataset.

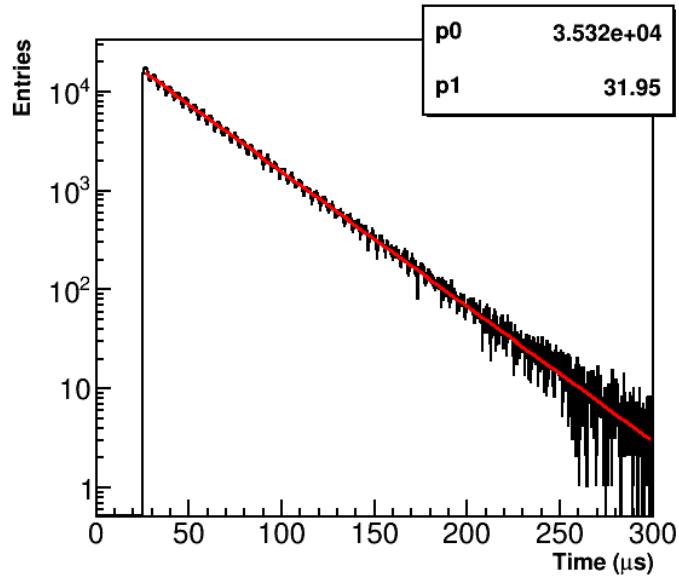


Figure 5.12: Plotted is the constructed pileup time spectrum on a log scale. Oscillations at  $\omega_a$  can be seen by eye. The histogram is fit to a simple two parameter exponential to get an idea of the lifetime of the pileup, calculated here as 31.95  $\mu\text{s}$ , which as expected is close to half of the muon lifetime at about 64.44  $\mu\text{s}$ . Data are from the 60h dataset.  
**potentially long way from text reference**

cluster times below 3.5 GeV, the systematic error on the extracted  $\omega_a$  frequency due to the pileup is within the target uncertainty for the level of statistics in the Run 1 dataset, Section 5.5.1. For analyses after Run 1 where the error budget is reduced, it may be necessary to improve the shadow method to account for triplets and the contamination. Finally, since the pileup is statistically constructed and then subtracted from the data, the errors on the final time histogram are no longer pure Poissonian. The proper calculation of the errors is detailed in Appendix C.

### 5.2.2 Ratio Method

The method used in this analysis to extract  $\omega_a$  is called the “Ratio Method,” or sometimes “R-Method.” It is a technique that modifies the data in such a way that the exponential decay in the time histogram is removed, and slowly varying effects are reduced. It was used successfully in the E821 experiment [91, 92, 93]. A full derivation of the equations in the method is given in Appendix B; here is given a short summary. Figure 5·13 provides a pictorial representation of how the method works.

The method works by randomly dividing the data into four separate datasets, one with the times of all clusters shifted up by half a  $g - 2$  period,  $+T_a/2$ , one with cluster times shifted down by half a  $g - 2$  period,  $-T_a/2$ , and two unchanged. Assuming the data are described by the five parameter function described in Section 2.1 and shown in Figure 5·13a,

$$N_d(t, E_{th}) = N_0(E_{th}) \cdot e^{-t/\gamma\tau_\mu} \cdot [1 + A(E_{th}) \cos(\omega_a t + \phi(E_{th}))], \quad (5.4)$$

and that the data are equally split into four subsets, then the four new datasets are given as<sup>9</sup>:

$$\begin{aligned} u_+(t) &= \frac{1}{4} N_5(t + T/2) \\ u_-(t) &= \frac{1}{4} N_5(t - T/2) \\ v_1(t) &= \frac{1}{4} N_5(t) \\ v_2(t) &= \frac{1}{4} N_5(t) \end{aligned} \quad (5.5)$$

In order to time shift the data as such,  $T_a$  needs to be known a priori to reasonable precision. The value used is taken from the E821 result, and its value is taken as

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<sup>9</sup>When handling the pileup in the ratio method, the pileup time spectra are split into four datasets and time-shifted in the same way as the cluster hit times. Associated doublets and singlets are kept together in the same individual dataset, and the four pileup datasets are subtracted from their respective ratio datasets before forming the ratio.

$1/f_a$ , where  $f_a$  is 0.229 073 5 MHz:

$$T_a \approx 4.365\,411\,\mu\text{s} \quad (5.6)$$

This value for  $f_a$  was determined by averaging column 2 of Table XV of the E821 Final Report [29], which consists of the  $f_a$  results for the different run periods in that experiment. A systematic error on the choice of this parameter is calculated in Section 5.5.5 and is negligible for the precision known.

The datasets are then combined as

$$\begin{aligned} U(t) &= u_+(t) + u_-(t), \\ V(t) &= v_1(t) + v_2(t), \end{aligned} \quad (5.7)$$

both of which are shown in Figure 5·13b. It is immediately apparent that the  $U(t)$  data are shifted 180° out of phase from the  $V(t)$  data. The ratio is then defined as<sup>10</sup>

$$R(t) = \frac{V(t) - U(t)}{V(t) + U(t)} \quad (5.8)$$

where the numerator and denominator are plotted in Figures 5·13c and 5·13d respectively. The numerator is an exponentially decaying cosine, while the denominator is a simple exponential, both of which can be seen as originating from the difference and sum of the  $U(t)$  and  $V(t)$  data respectively. The resulting ratio spectrum can be seen in Figure 5·13e, where the exponential has been eliminated. The fit function is then reduced from five parameters down to three:

$$R(t) \approx A \cos(\omega_a t) - C, \quad (5.9)$$

---

<sup>10</sup>The ratio can also be defined with  $U(t) - V(t)$  in the numerator, however then the phase of the ratio spectrum is shifted 180° from the original  $N_5(t)$  spectrum.

where

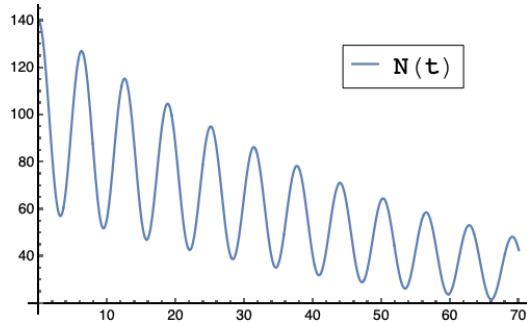
$$C = \frac{1}{16} \left( \frac{T}{\tau} \right)^2 \approx 2.87 * 10^{-4}, \quad (5.10)$$

and these functions have been determined from the time-shifted five parameter function plugged into the  $U(t)$  and  $V(t)$  variables. In addition to the exponential being eliminated, any slow terms in the data get time-shifted and divided as well, such that the amplitude of said slow effects are reduced. For faster effects, the degree of cancellation of the effect is dependent on the frequency. Effects at frequencies which are an odd multiple of  $\omega_a$  are preserved while effects at an even multiple of  $\omega_a$  are completely canceled out. An example is shown in Figure 5.14. While this makes fitting the data easier in some cases, in others it is a downside that effects which still need to be included in the fit function now have their amplitudes reduced, making them harder to fit.

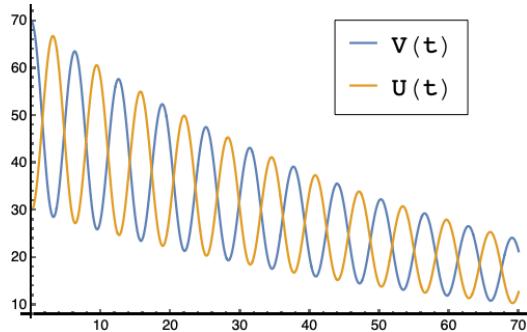
In order to eliminate the constant  $C$  at the end of Equation 5.9, a different weighting scheme is used in this analysis as described in Reference [37]:

$$\begin{aligned} u_+(t) &= \frac{e^{T/2\tau}}{2 + e^{T/2\tau} + e^{-T/2\tau}} N_5(t + T/2) \\ u_-(t) &= \frac{e^{-T/2\tau}}{2 + e^{T/2\tau} + e^{-T/2\tau}} N_5(t - T/2) \\ v_1(t) &= \frac{1}{2 + e^{T/2\tau} + e^{-T/2\tau}} N_5(t) \\ v_2(t) &= \frac{1}{2 + e^{T/2\tau} + e^{-T/2\tau}} N_5(t) \end{aligned} \quad (5.11)$$

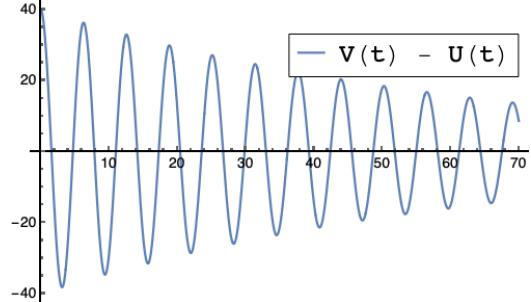
Here  $\tau = \tau_\mu$  where  $\tau_\mu$  is the time-dilated muon lifetime, and the factors out front are each close to 1/4 and account for the degree of muon decay over a time period of  $T_a/2$ . Similar to  $T_a$ , the muon lifetime must be known a priori. Its value is taken as 64.44  $\mu$ s, determined from fits to the data. A systematic study regarding this parameter is described in Section 5.5.5. The ratio spectrum is then almost exactly



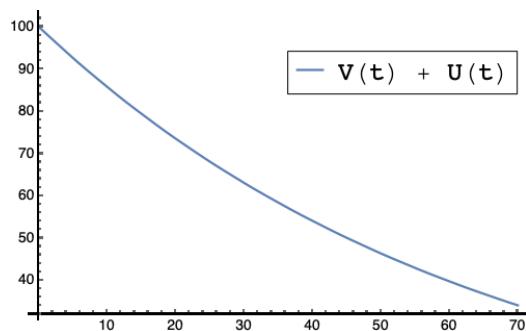
(a) The five parameter function defined in Equation 5.4, which describes the incoming data to first order.



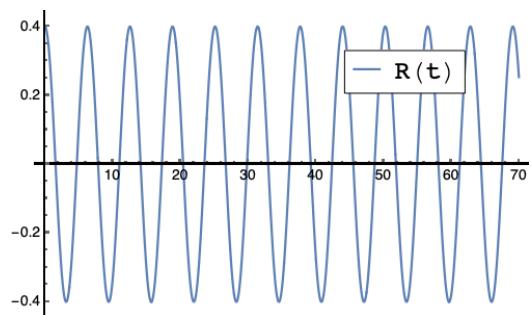
(b)  $U(t)$  and  $V(t)$  functions which describe the time-shifted and un-shifted ratio datasets.



(c) The numerator function in the formed ratio,  $V(t) - U(t)$ . It is an exponentially decaying cosine.



(d) The denominator function in the formed ratio,  $V(t) + U(t)$ . It is a decaying exponential.



(e) The ratio function, which describes the data after it has been handled according to the text. To first order it is a simple cosine.

Figure 5.13: Functions describing the formation of the ratio data. **potentially long way from text reference**

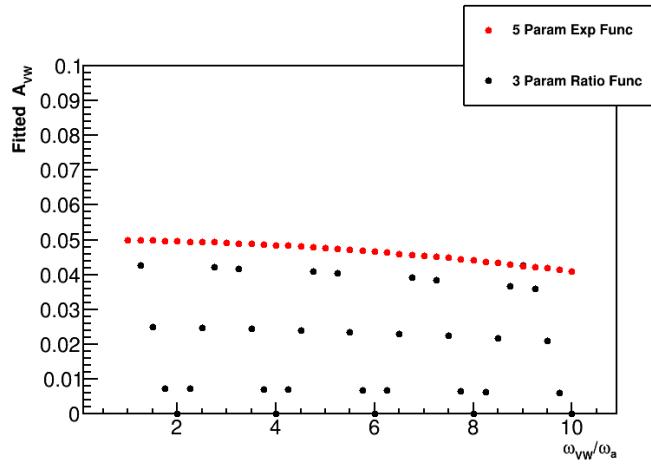


Figure 5.14: Fitted amplitude for a VW effect as a function of frequency in units of  $\omega_a$  in a Toy MC simulation, with a five parameter function in red and a three parameter ratio function in black. The input amplitude is 0.05. The fall off of red points is due to the high frequencies relative to the bin widths leading to an underestimate of the amplitude; performing an integral fit removes this trend. As shown, the amplitude of the effect goes to zero for frequencies which are an even multiple of  $\omega_a$ .

described by just the cosine term,

$$R(t) \approx A \cos(\omega_a t), \quad (5.12)$$

in the absence of other effects in the data.

### 5.3 Fitting the data

The basic five parameter function used to fit the data as described before is given as<sup>11</sup>

$$f(t) = \mathbf{N}_0 \cdot e^{-t/\tau} \cdot (1 + \mathbf{A} \cdot \cos(\omega_a t + \boldsymbol{\phi})), \quad (5.13)$$

where the fit parameter for  $\omega_a$  is recast in terms of a ppm level shift  $\mathbf{R}$  on a reference frequency,

$$\omega_a = 2\pi \cdot 0.2291 \text{ MHz} \cdot (1 + \mathbf{R} \times 10^{-6}). \quad (5.14)$$

This reference frequency of 0.2291 MHz was the same reference frequency used in E821, and  $\mathbf{R}$  is blinded at the hardware and software levels [94, 95]. Fitting the data with Equation 5.13 however is insufficient to properly describe the data. An FFT of the residuals between fit and data is shown in Figure 5.15. There are peaks due to beam dynamics frequencies corresponding to the CBO, VW, and some beat frequencies with  $\omega_a$ . In order to properly account for these effects, additional terms need to be added to the fit function.

Equation 5.13 can be expanded to

$$f(t) = \Lambda(t) \cdot V(t) \cdot N_{cbo}(t) \cdot \mathbf{N}_0 \cdot e^{-t/\tau} \cdot (1 + A_{cbo}(t) \cdot \cos(\omega_a t + \phi_{cbo}(t))), \quad (5.15)$$

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<sup>11</sup>In the current section the actual fit parameters are in bold.

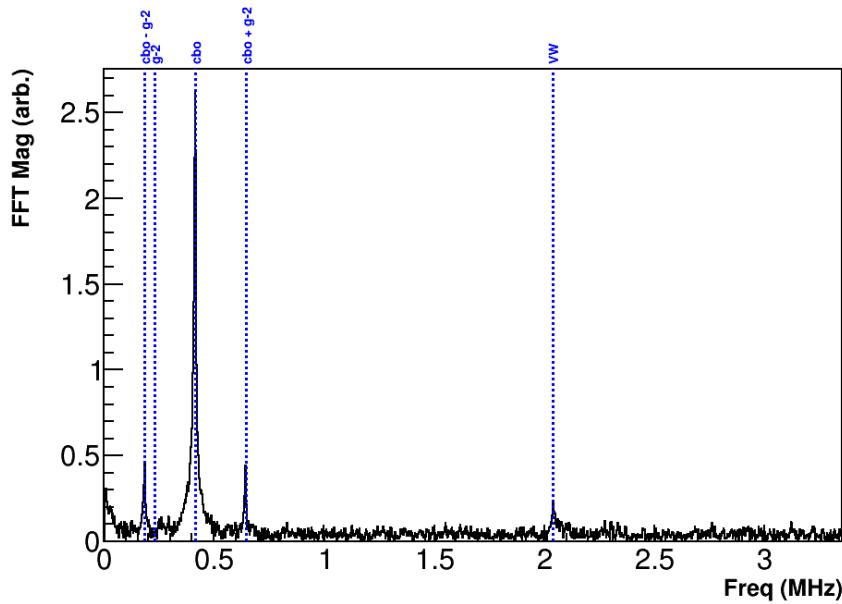


Figure 5.15: FFT of five parameter fit residuals. Peaks corresponding to beam dynamics frequencies of the CBO and VW and some beat frequencies with  $\omega_a$  are readily apparent. A rise at low frequencies corresponds to the effects of the lost muons in the data. The VW peak has already been diminished by the choice of binning and disappears entirely with the additional level of time randomization at that frequency. From the 60h dataset.

where many additional terms have been added in order to account for effects in the data. The various additional terms  $\{N_{cbo}(t), A_{cbo}(t), \phi_{cbo}(t), V(t), \Lambda(t)\}$  are described in the following sections. Fitting the data with this function, referred to as the “Threshold Method” or just “T-Method,” while not the subject of this dissertation, was done in this analysis as a diagnostic and informative tool for the Ratio Method analysis.

In order to fit the ratio time spectra as constructed in Section 5.2.2, a different function is used. While the immediate inclination would be to use an expansion of Equation 5.12 with additional effects included similar to the T-Method fit function, instead the fit function used is a return to the explicit definition of the construction of the ratio time spectra in Equations 5.7 and 5.8. Including the additional effects previously mentioned, the fit function becomes

$$R(t) = \frac{2f(t) - f_+(t) - f_-(t)}{2f(t) + f_+(t) + f_-(t)}, \quad (5.16)$$

$$f_{\pm}(t) = f(t \pm T_a/2), \quad (5.17)$$

$$f(t) = \Lambda(t) \cdot V(t) \cdot N_{cbo}(t) \cdot (1 + A_{cbo}(t) \cdot \cos(\omega_a t + \phi_{cbo}(t))). \quad (5.18)$$

The  $f(t)$  given here differs from that in Equation 5.15 in that the  $N_0 \cdot e^{-t/\tau}$  terms have divided out, thus reducing the number of fit parameters necessary to model the data. Using this function as opposed to an expansion of the three parameter ratio function eliminates any approximations made in that three parameter function derivation, and any fit parameters should be consistent in value between the T-Method and Ratio Method results, barring adjustments due to the application of the Ratio Method.

Because the Ratio Method reduces the sensitivity of the  $\omega_a$  determination to various effects in the data, peaks that appear in the FFT of the five parameter fit residuals do not appear in the FFT of the three parameter ratio fit residuals, Figure 5.16. Indeed unless one looks at the FFT over the early part of the fit (first

30  $\mu\text{s}$ ) or at the shape of the Ratio Method denominator, one might not know the effects even exist in the data at all, barring occasionally poor  $\chi^2$ 's. Nevertheless, those effects typically still need to be included in the fit function for a proper estimation of  $\omega_a$ . Because of the reduction in sensitivity in the ratio fits, there are some parameters which the ratio has trouble fitting by itself. Using the results from a T-Method fit to the data can then be a useful tool for constraining those specific parameters.

### 5.3.1 CBO terms

As described in Section 2.6.1, the CBO effect modulates the  $\omega_a$  oscillation. This shows up as a modification on the five parameter function parameters  $\{N_0, A, \phi\} \rightarrow \{N_0 \cdot N_{cbo}(t), A_{cbo}(t), \phi_{cbo}(t)\}$  where these terms are given to first order as

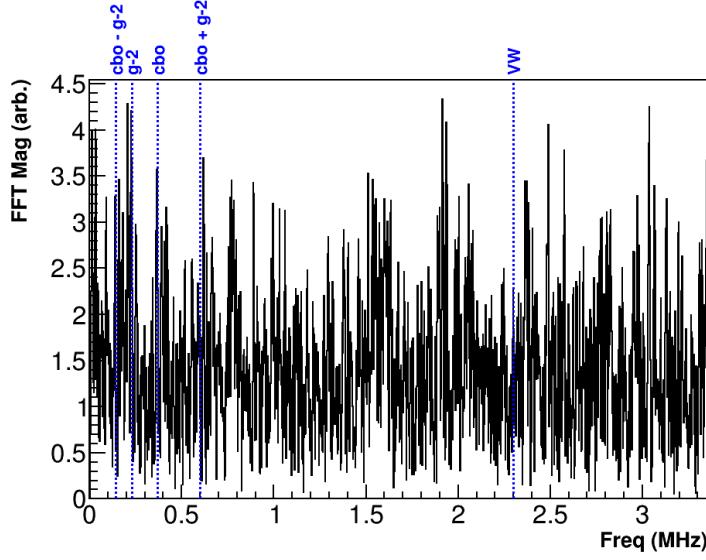
$$N_{cbo}(t) = (1 + \mathbf{A}_{cbo-N} \cdot e^{-t/\tau_{cbo}} \cdot \cos(\omega_{cbo}(t) \cdot t + \phi_{cbo-N})) \quad (5.19)$$

$$A_{cbo}(t) = \mathbf{A} \cdot (1 + \mathbf{A}_{cbo-A} \cdot e^{-t/\tau_{cbo}} \cdot \cos(\omega_{cbo}(t) \cdot t + \phi_{cbo-A})) \quad (5.20)$$

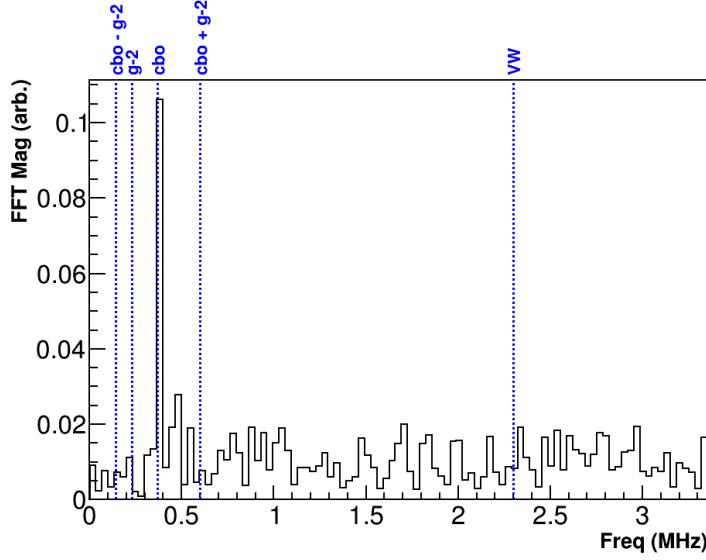
$$\phi_{cbo}(t) = \phi_0 + \mathbf{A}_{cbo-\phi} \cdot e^{-t/\tau_{cbo}} \cdot \cos(\omega_{cbo}(t) \cdot t + \phi_{cbo-\phi}) \quad (5.21)$$

In Equation 5.19 the  $\mathbf{N}_0$  term is left out since the ratio fit includes the  $N_{cbo}(t)$  term but not  $\mathbf{N}_0$ , and in Equation 5.21,  $\phi_{cbo}(t)$  has an additive phase instead of a multiplicative one since  $\phi_0$  is not an amplitude and can be equal to zero. Each of the terms then includes additional fit parameters in an extra amplitude and phase, as well as one shared CBO lifetime and frequency. As described in Section 4.4, the default model for the CBO modulation is assumed as an exponentially decaying envelope. The CBO frequency,  $\omega_{cbo}(t)$ , was time-dependent for Run 1 as found in Section 4.4. The function for the CBO frequency shown in Figure 4.29 is given in the fit function as

$$\omega_{cbo}(t) = \omega_{cbo} \cdot \left( 1 + \frac{Ae^{(-t/\tau_A)}}{\omega_0 t} + \frac{Be^{(-t/\tau_B)}}{\omega_0 t} \right), \quad (5.22)$$



(a) FFT of fit residuals over all times within the fit range. There is no immediately apparent structure for residual effects left out of the fit function.



(b) FFT of fit residuals over the first 30  $\mu$ s. The CBO peak can be seen above the noise, though the corresponding beat frequencies do not appear.

Figure 5.16: FFT of three parameter ratio fit residuals. Dashed blue lines indicate various beam dynamics frequencies and their beat frequencies with  $\omega_a$ . Data are from the 60h dataset.

CBO Frequency Model Parameters						
Dataset	Tracker Station	$\omega_0$ (rad/ $\mu$ s)	$A$ (rad)	$\tau_A$ ( $\mu$ s)	$B$ (rad)	$\tau_B$ ( $\mu$ s)
60h	12	2.3389	2.9	81.8	5.12	7.7
	18	2.3387	2.82	81.1	5.08	8.2
HighKick	12	2.6145	3.27	52.8	6.96	6.6
	18	2.6137	3.23	46.2	6.61	6.8
9d	12	2.6106	2.86	72.8	5.50	8.5
	18	2.6110	2.89	79.2	5.44	9.2
Endgame	12	2.3377	7.43	95.1	4.71	9.0
	18	2.3379	7.44	95.2	4.90	9.2

Table 5.2: Fixed parameters in the CBO frequency model [96, 97]. The larger values of the  $A$  and  $\tau_A$  parameters for the Endgame dataset are a consequence of the degradation of the quadrupole resistors over the course of Run 1.

where  $\omega_{cbo}$  is the free fit parameter, and the model parameters  $\{\omega_0, A, \tau_A, B, \tau_B\}$  are fixed from the tracking analysis. These parameters for the various datasets and two tracker stations are given in Table 5.2.

It should be noted that Equations 5.20 and 5.21 are not necessarily needed in order to get good fits to the data (whereas Equation 5.19 always is). This is typically dataset or random seed dependent. While some datasets had certain parameters with large errors relative to their amplitudes, for this analysis all terms were included in all fits. Fits converged properly with appropriate tuning of the starting parameters, justifying their inclusion. When considering higher order CBO modifications to the fit function, the only term that was found to be necessary was the second order CBO modulation on the  $N_0$  term,  $N_0 \cdot N_{cbo}(t) \rightarrow N_0 \cdot N_{cbo}(t) \cdot N_{2cbo}(t)$ , where

$$N_{2cbo}(t) = (1 + \mathbf{A}_{2cbo-N} \cdot e^{-2t/\tau_{cbo}} \cdot \cos(\omega_{cbo}(t) \cdot t + \phi_{2cbo-N})). \quad (5.23)$$

This stands to reason as the  $N_{cbo}(t)$  is the largest CBO effect. The form is assumed to be the same as the first order CBO terms, except the lifetime of the effect is half

the CBO lifetime,  $\tau_{cbo}/2$ . This is due to the fact that  $N_{2cbo}(t)$  comes from the width of the oscillating beam, as opposed to the oscillating mean. Indeed as will be shown in Section 5.4.1, the inclusion of this term is necessary to get good fits to the per calorimeter data, where the CBO effect is stronger compared to the sum of the data from all calorimeters.

Systematic studies relating to the choice of CBO envelope and choice of fixed parameters in the CBO frequency model are explored in Section 5.5.3. For future runs beyond Run 1, it may be necessary to include the higher order modifications to the  $A_{cbo}(t)$  and  $\phi_{cbo}(t)$  terms.

### 5.3.2 VW term

As mentioned at the end of Section 5.2, the VW effect is time-randomized out of the data, such that  $V(t) = 1$ . This is done due to complications with the Ratio Method. In the 60h and Endgame datasets, the VW frequency was found to be nearly  $10\omega_a$ , on a potential resonance. While to first order this even multiple frequency implies the VW effect should completely cancel out in the Ratio Method, Figure 5.14, the FR effect in combination with the VW leads to a modified envelope for the VW effect in the Ratio Method fits and inflated VW amplitudes [98]. See Figures 5.17 and 5.18.

For the 9d dataset which has a VW frequency that's nearly  $9\omega_a$  and avoids said resonance (and by extension the HighKick), it was found that the Ratio Method flattened out the VW amplitudes as a function of calorimeter number leading to a systematically smaller VW amplitude in the calorimeter sum fit compared to in a T-Method fit. The simplest solution to remove both of these problems was to randomize out the VW effect entirely. Table 5.3 gives the change in asymmetry and corresponding change in the statistical error on R due to this added time randomization for three of the Run 1 datasets. It was found that the added randomization increased the statistical error on R by a negligible amount for the Run 1 analysis. It

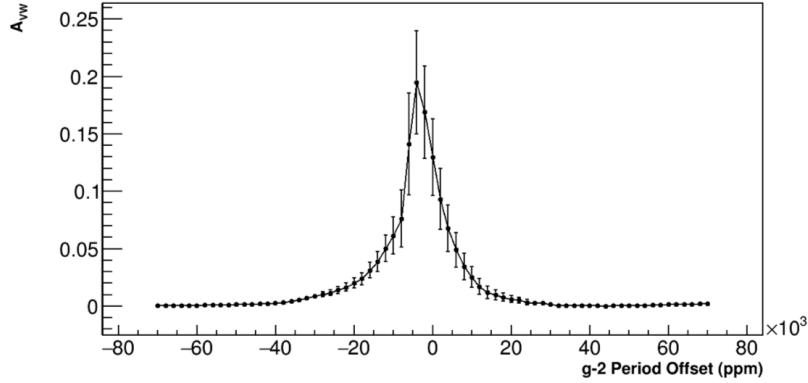


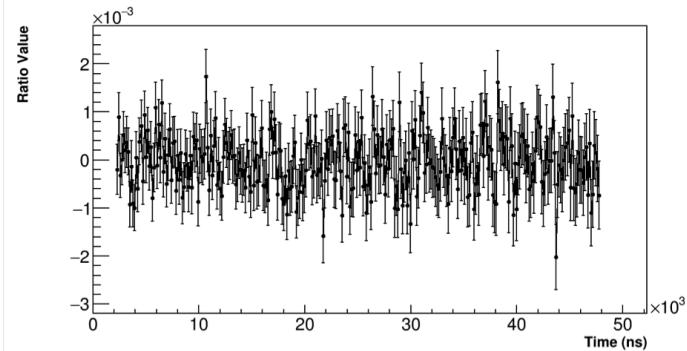
Figure 5.17: Fitted VW amplitude as a function of the choice of offset from  $T_a$  in units of thousands of ppm for the 60h dataset. The default time-shift lies at 0 on this plot, right on the resonance where the VW amplitude blows up. Only by time-shifting by a drastically different amount (which negatively affects R), can the resonance be avoided.

Change in Asymmetry due to VW Randomization				
Dataset	$A$ no randomization	$A$ with randomization	$\Delta A$	$\Delta\sigma_R$ (ppb)
60h	0.3697	0.3637	-0.0060	22.7
HighKick	0.3707	0.3632	-0.0075	29.3
9d	0.3714	0.3639	-0.0075	18.1
Endgame	0.3747	0.3686	-0.0061	10.7

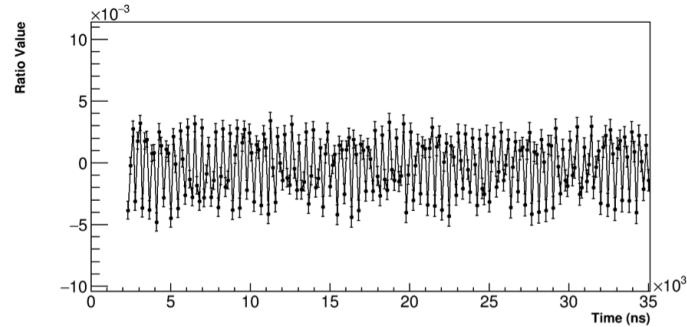
Table 5.3: Asymmetry values in the the Run 1 datasets with and without the VW randomization, and the corresponding change in the statistical error on R. An energy cut of 1700 MeV was applied to the data.

was also found that the added level of time-randomization changed the mean value of R for many random seed fits to the data by a small amount, statistically consistent without the extra randomization [98]. Going forward for the future runs, it may be necessary to include the proper VW envelope in the fits using a functional form of the FR instead of randomizing out the effect.

The form for the VW as it is used in the T-Method fits is taken identically to the



(a) Without the FR effect included.



(b) With the FR effect included.

Figure 5.18: Ratio data with and without the FR effect from a Toy MC simulation, with a VW effect with a frequency  $\omega_{VW} = 10\omega_a$ . The  $\omega_a$  wiggle itself has been removed, and the lifetime of the VW was set to a large number. The top plot shows ratio data which is consistent with 0 after all effects have been removed and the VW has divided out. The bottom plot shows ratio data inconsistent with 0, with oscillations at the VW frequency, and an interesting beating structure. Note the different scales.

CBO terms,

$$V(t) = 1 + \mathbf{A}_{\mathbf{VW}} \cdot e^{-t/\tau_{\mathbf{VW}}} \cos(\omega_{VW}(t) \cdot t + \phi_{\mathbf{VW}}), \quad (5.24)$$

with an exponentially decaying envelope, and additional amplitude and phase parameters. The VW frequency  $\omega_{VW}$  is given in Equation 2.26, where it is seen to be dependent on the cyclotron frequency and vertical betatron frequency. Using Equations 2.20 and 2.25, the dependence on the CBO frequency is seen as

$$\begin{aligned} \omega_{VW}(t) &= 2\pi(f_c - 2f_{y_{BO}}), \\ &= 2\pi\left(f_c - 2f_{cbo}(t)\sqrt{2f_c/f_{cbo}(t) - 1}\right), \end{aligned} \quad (5.25)$$

where  $f_{cbo}(t) = \omega_{cbo}(t)/2\pi$  is determined in the tracking analysis as described in Section 5.3.1 and given by Equation 5.22. While Equation 5.25 is the theoretical frequency for the VW effect in the continuous quadrupole approximation, it was found in the tracking analysis that including an adjustment factor on the CBO frequency  $f_{cbo} \rightarrow \kappa f_{cbo}$  resulted in better agreement with the directly measured VW frequency [86, 99]. In the fitting function itself, the VW frequency is then taken as

$$\omega_{VW}(t) = 2\pi\left(f_c - 2 \cdot \kappa_{\mathbf{VW}} \cdot f_{cbo}(t)\sqrt{2f_c/(\kappa_{\mathbf{VW}} \cdot f_{cbo}(t)) - 1}\right), \quad (5.26)$$

where now the VW frequency fit parameter is  $\kappa_{\mathbf{VW}}$ . This adjustment factor ends up being on the order of about a percent in the fits to the data.

### 5.3.3 Lost muons

Muons lost from the storage ring during the frequency analysis portion of each fill will distort the observed decay positron spectrum. These hits show up as a rise at low frequencies in the FFT of the fit residuals due to the slow nature of the effect, Figure 5.15. These muon losses typically originate from those muons with large



Figure 5.19:  $\Delta t$  and energy deposition distributions for lost muons passing through adjacent calorimeters, where counts have been included from hits in both calos. A typical flight time is 6.5 ns and energy deposition is 170 MeV. Data are from the Endgame dataset.

betatron amplitudes which hit material near the edge of the storage ring, or those muons which experience local field perturbations one too many times. In both cases the muons will lose energy and spiral inward out of the ring, a fraction of which will then pass through multiple calorimeters. Because lost muons are minimum-ionizing particles (MIPs), they are relatively easily identified by their small energy deposition signature in hit calorimeters, around 170 MeV as shown in the left peak in Figure 5.4. These lost muons typically have a flight time between adjacent calorimeters of  $\Delta t_{12} = 6.5$  ns [89, 100]. The  $\Delta t$  and energy deposition distributions are shown in Figure 5.19. By looking for coincidences between three adjacent calorimeters, or triples, and then applying cuts and subtracting backgrounds, a pure sample of lost muons can be constructed<sup>12</sup>. This lost muon spectrum  $L(t)$  can then be implemented into the fit function in order to account for the positrons that would have been observed later in the fill in the absence of losses.

The cuts used for the lost muon selection are given in Table 5.4. Triple coincidences are only included where every cluster consists of three or less crystals hit, with

<sup>12</sup>Using triples rather than doubles helps reduce the accidental background while keeping enough statistics.

<b>Lost Muon Cuts</b>	
Parameter	Value or Range
Cluster size	$\leq 3$ crystals
Cluster energy fraction	$\geq 0.8$ in main crystal
Time of flight between adjacent calorimeters	$5 \text{ ns} \leq \Delta t_{12,23} \leq 7.5 \text{ ns}$
Energy deposition	$100 \text{ MeV} \leq E_{1,2,3} \leq 250 \text{ MeV}$
Time of flight between separated calorimeters	$\Delta t_{13} \leq 14.4 \text{ ns}$

Table 5.4: Lost muon selection cuts. In a triple coincidence the subscripts of 1, 2, and 3 correspond to the three calorimeters hit clockwise around the ring.

80% of the energy deposited in one crystal.  $\Delta t$  and energy deposition ranges are taken as  $5 \text{ ns} \leq \Delta t_{12,23} \leq 7.5 \text{ ns}$  and  $100 \text{ MeV} \leq E_{1,2,3} \leq 250 \text{ MeV}$ , where these ranges come from inspection of Figure 5.19. The  $\Delta t_{12}$  distribution as a function of time in-fill is shown in Figure 5.20. By examining this distribution in the range  $2 \text{ ns} \leq \Delta t_{12} \leq 4 \text{ ns}$ , and averaging the contained counts, an approximation for the accidental background can be determined and subtracted off the triples spectrum. The accidental background typically comes from either double coincidences and a real positron hit, or a particle shower induced by an incident positron which hits an adjacent calorimeter.

Also shown in Figure 5.20 are two bands of stable beam contaminants corresponding to stored deuterons and protons. These particles have different times of flights between calorimeters due to their larger masses. By looking at the  $\Delta t_{13}$  distribution for times greater than  $300 \mu\text{s}$  so that most muons have already decayed, the deuteron population is easily isolated, Figure 5.21. While the deuteron population is mostly removed by the  $\Delta t_{12}$  cut, an additional cut of  $\Delta t_{13} \leq 14.4 \text{ ns}$  helps remove any remaining deuteron contamination. The proton population, due to its similarity to the real lost muon population, is harder to remove. The simplest solution is to simply cut on the negative side of the  $\Delta t_{12}$  or  $\Delta t_{13}$  distributions. See Section 5.5.4 for the results using this additional cut. It was found that the proton contamination makes



Figure 5.20:  $\Delta t_{12}$  distribution as a function of time in-fill before any cuts. Note the log color scale. Lost muons have a  $\Delta t_{12}$  distribution centered at 6.5 ns. The accidental background can be seen as counts out at  $\Delta t$ 's far from the center of the distribution. Color striations in the core of the distribution correspond to CBO periods. There are two bands of hits that do not fall off as severely with time as the lost muons do. The band contained mostly between 7 and 8 ns corresponds to deuterons, while the band contained mostly between 6 and 7 ns corresponds to protons. Data are from the Endgame dataset.



Figure 5·21:  $\Delta t_{13}$  distribution as a function of energy for times greater than 300  $\mu$ s. The lost muons can be seen centered at 170 MeV and  $\Delta t_{13} \approx 12.5$  ns, while the deuterons can be seen at  $\Delta t_{13} \geq 14.4$  ns. While the deuterons have a preferentially larger energy deposition, they can be seen to extend to low energies, making cutting on energy unrealistic. Though not easily separated by eye, the stored protons are contained within the upper right portion of the lost muons. Data are from the Endgame dataset.

almost no difference to the fitted value of R. The default choice then is to use the previously specified cut ranges in order to increase the amount of statistics in the lost muons distribution with which to fit. A study into the exact rate of these beam contaminants is included in Reference [101].

The last background is the quadruples spectrum. Due to how the triple coincidences are constructed, real quadruples will be counted as two separate triples. While the quadruples spectrum can be used instead of the triples spectrum for a purer sample of lost muons, the statistics are much reduced, and similarly for higher order coincidences. The quadruples spectrum is constructed in the same was as the triples with the same cut ranges. The quadruple background is removed by subtracting off those triples which originated from quadruple coincidences.

Figure 5·22 shows the final  $\Delta t_{12}$  distribution with cuts as a function of time in-fill

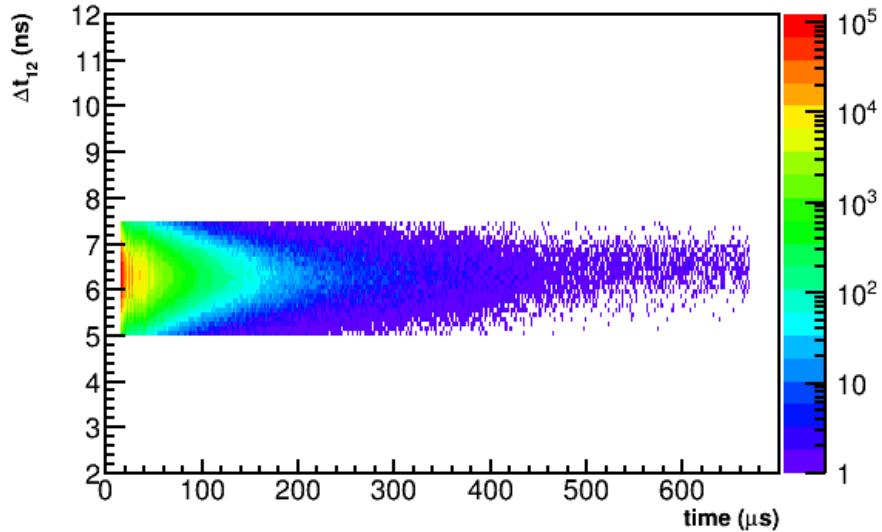


Figure 5.22: Final  $\Delta t_{12}$  distribution for selected lost muons as a function of time in-fill. Data are from the Endgame dataset.

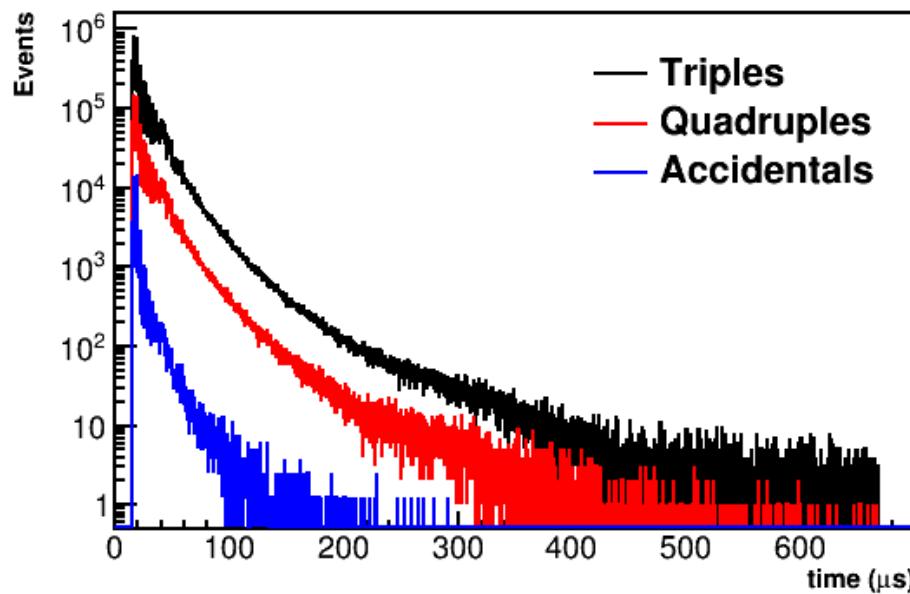


Figure 5.23: Lost muon triples spectrum  $L(t)$  as a function of time in-fill, after all cuts and background subtractions. Also shown are the quadruples and accidental background distributions, which can be seen to be much smaller than the triples. The shape of the triple and quadruple spectra can be seen to be nearly the same. The shape of the triples spectrum comes from beam dynamics (CBO) and lost muon acceptance effects. Data are from the Endgame dataset.

for selected lost muons. The final triple losses spectrum  $L(t)$  is shown in Figure 5·23. Once we have the lost muon distribution we need to include it in the fit function in order to account for the changing number of hits over the course of the fill. The true lost muon rate will be given by  $L(t)/\epsilon$ , where  $\epsilon$  is the loss detection efficiency. The change in the number of muons within the storage ring can be written as

$$dN = -\frac{N}{\tau}dt - \frac{L(t)}{\epsilon}dt, \quad (5.27)$$

where the solution can be determined by inspection as

$$N(t) = N_0 \cdot e^{-t/\tau} \cdot \left(1 - \frac{1}{\epsilon N_0} \int_{t_0}^t L(t')e^{(t'/\tau)}dt'\right). \quad (5.28)$$

The parameter  $\tau$  is the same muon lifetime as in the T-Method fit function, except here it is set as the default value of  $64.44\text{ }\mu\text{s}$ <sup>13</sup>. The value for  $t_0$  can be taken as any time at or before the start of the fitting range, as it simply changes the scale of  $N_0$ . The modification  $\Lambda(t)$  as listed in Equation 5.15 is then taken as

$$\Lambda(t) = 1 - \kappa_{loss} \int_{t_0}^t L(t')e^{(t'/\tau)}dt' \quad (5.29)$$

where  $\kappa_{loss} = 1/\epsilon N_0$  is taken as the fit parameter<sup>14</sup>.

It should be noted that the Ratio Method is largely insensitive to the slow lost muons effect which divides out. No rise appears in the FFT of the ratio fit residuals, Figure 5·16, and when letting  $\kappa_{loss}$  float, the fit does not converge. It was found however that  $\mathbf{R}$  changes on the order of tens of ppb when the lost muon term is included with  $\kappa_{loss}$  fixed to the value determined from a T-Method fit, so by default

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<sup>13</sup>The fitted  $\omega_a$  frequency is largely insensitive to this parameter, and especially so in the Ratio Method fits, so this is acceptable. If necessary, the fitting procedure can be iterated.

<sup>14</sup>By construction  $\kappa_{loss}$  is a very small parameter,  $\mathcal{O}(10^{-10})$ . This factor is absorbed into the parameter in such that the fit parameter is  $\mathcal{O}(1)$ .

it is included. See Section 5.5.4 for more discussion on this.

### 5.3.4 Fit procedures and parameters

$\chi^2$  fits are done to the pileup corrected positron time spectra and ratio data in **ROOT** using the standard **TH1F** and **TGraph** fit methods with a strategy level of two. Fits are performed in stages where groups of associated parameters are freed, fit, and then fixed before fitting the next set of parameters. In the final fit all parameters are free. Each of the four Run 1 precession frequency analysis datasets are fit with 14 parameters. 13 parameters are free in fits to the 60h, 9d, and Endgame datasets, with the  $\kappa_{loss}$  parameter determined and fixed from a T-Method fit. In the HighKick dataset  $\tau_{cbo}$  is also fixed from a T-Method fit as the lifetime is relatively short and the Ratio Method can't successfully fit it, resulting in 12 free fit parameters. In the T-Method fits there are two extra free fit parameters, those being the  $N_0$  and  $\tau$  terms which are divided out of the Ratio Method fits.

Fits to calorimeter sum spectra are done over the range 30.2–650  $\mu\text{s}$ , corresponding to 4155 bins<sup>15</sup>. The choice of fit start time was made to allow for the muon beam to stabilize in the storage region after the scraping procedure. In addition the start time was chosen to lie directly on a  $g - 2$  zero crossing, which from E821 experience was shown to reduce some systematic errors. The choice of fit end time was made to be well before the limit at which the bin variance become non-Gaussian. Individual calorimeter fits were performed out to 400  $\mu\text{s}$  considering this fact. Table 5.5 gives the various fit procedure parameters.

As a reminder  $R$  is blinded at the hardware and software levels [94, 95]. The software blinding string used for the 60h dataset was different than that used when fitting the HighKick, 9d, and Endgame datasets, each of which used the same blinding

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<sup>15</sup>The fit range is not defined on exact bin edges, and those bins where the fit begins and ends are included in the fit.

<b>Fit Procedure Parameters</b>	
Parameter	Value
Fit strategy level	2
Fit start time	30.2 $\mu$ s
Fit end time (calorimeter sum)	650 $\mu$ s
Fit end time (single calorimeters)	400 $\mu$ s
Bins in fit (calorimeter sum)	4155
Number of free fit parameters (60h, 9d, Endgame)	13
Number of free fit parameters (HighKick)	12

Table 5.5: Various parameters used in the fit procedure.

<b>SAM Dataset Names</b>	
Dataset	Name
60h	gm2pro_daq_full_run1_60h_5039A_goldList
HighKick	gm2pro_daq_full_run1_HighKick_5040A_goldList
9d	gm2pro_daq_full_run1_9d_5039A_goldList
Endgame	gm2pro_daq_full_run1_EndGame_5040A_goldList

Table 5.6: The SAM dataset names of the Run 1 precession frequency analysis datasets.

string. This was done in order to perform a software-level relative unblinding exercise between different analyzers with the 60h dataset in order to determine if there were any obvious problems in the analyses [102, 103]. Therefore in the following results,  $R$  values between different datasets are comparable between the datasets except for the 60h, barring any differences due to field conditions which are not discussed in this dissertation.

The names of the specific SAM (Sequential Access via Metadata) datasets that were analyzed for this dissertation are given in Table 5.6. SAM is a data handling system used at Fermilab to store and retrieve files along with associated metadata. The specific DQC cuts, included gain corrections, processed runs, production software versions, etc. can all be recovered by examining the associated SAM metadata.

## 5.4 Fit results

Figure 5·24 shows fits to the four Run 1 precession frequency analysis datasets for single random seeds. Table 5.7 gives all fit parameters and their errors. In each dataset case the  $\chi^2/\text{NDF}$  is acceptable as evidenced by the p-value included in the table results. Fit pulls and the FFT of the fit residuals for the 60h dataset are provided in Figure 5·25. As shown all structure has been eliminated within the fit residuals implying that all effects in the data have properly been accounted for in the fit function. The same checks were made for the HighKick, 9d, and Endgame datasets and in each case no residual structure remained. Appendix D provides the correlation matrices for the various datasets. The only fit parameter that is significantly correlated with  $R$  is the  $g - 2$  phase in all datasets. This increases the confidence in the  $\omega_a$  extraction, as effects in the data which might potentially be mismodeled will only weakly correlate with the final fitted  $R$  value. The various different CBO parameters are self-correlated to different degrees depending on the parameter and the dataset that is being fit. Typically either the phases and frequencies are correlated, or the lifetimes and amplitudes.

The  $g - 2$  phases for the different datasets showed small differences, due primarily to upstream beam adjustments before injection into the storage ring. Similarly, slightly different asymmetries can be attributed to very small acceptance differences in the stored beams. As described in Section 5.3.3, the value for  $\kappa_{loss}$  was determined and fixed from a T-Method fit to the data. The values themselves don't directly correspond to the level of losses, as each dataset has it's own loss function  $L(t)$ .

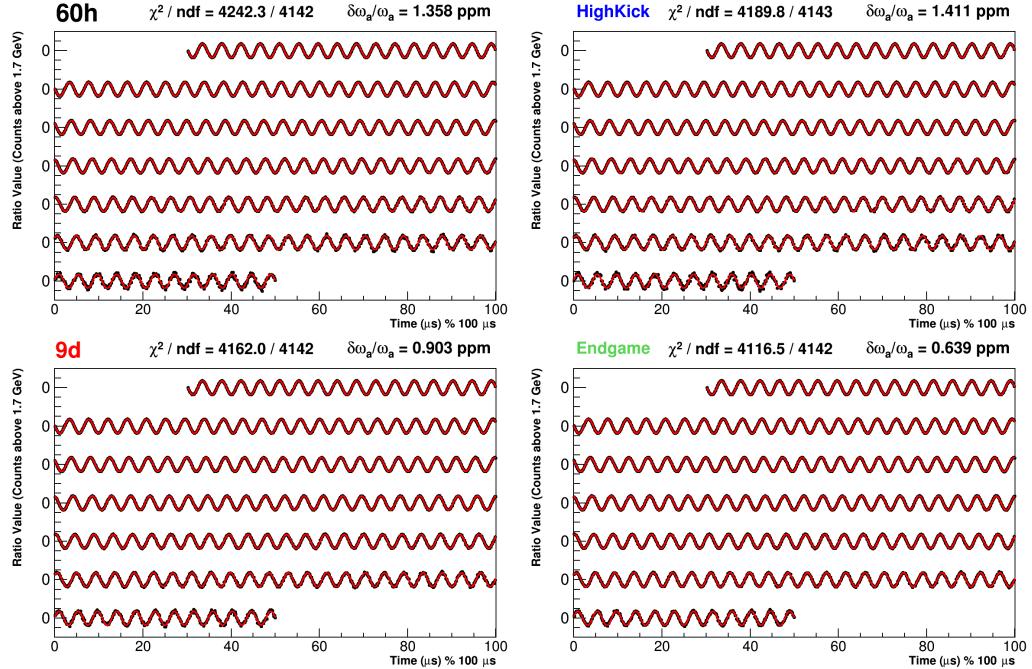


Figure 5.24: Single random seed fits to calorimeter summed data of Run 1 precession frequency analysis datasets. Data is in black and the fits are in red. The x axis is in units of  $\mu\text{s}$  modulo  $100 \mu\text{s}$ , with successive portions of the data points and fit shifted downwards on the plot. The fit ranges from  $30.2\text{--}650 \mu\text{s}$ . Dataset names are given in the upper left corners of the figures, alongside the  $\chi^2$  per degree of freedom and relative error on  $\omega_a$  or  $R$ .

Ratio Method Fit Results								
	60h		HighKick		9d		Endgame	
$\chi^2/\text{NDF}$	4242/4142		4190/4143		4162/4142		4116/4142	
p-value	0.1356		0.3018		0.4104		0.6079	
Parameter	Value	Error	Value	Error	Value	Error	Value	Error
$A$	0.3637	$4.4 \times 10^{-5}$	0.3632	$4.6 \times 10^{-5}$	0.3639	$2.9 \times 10^{-5}$	0.3686	$2.1 \times 10^{-5}$
<b><math>R</math> (ppm, blinded)</b>	<b>-20.8479</b>	<b>1.3581</b>	<b>-17.5433</b>	<b>1.4112</b>	<b>-17.8214</b>	<b>0.9033</b>	<b>-17.5674</b>	<b>0.6393</b>
$\phi$	2.091	$2.2 \times 10^{-4}$	2.081	$2.3 \times 10^{-4}$	2.080	$1.5 \times 10^{-4}$	2.076	$1.1 \times 10^{-4}$
$\omega_{cbo}$ (rad/ $\mu$ s)	2.338	$1.4 \times 10^{-3}$	2.599	$6.6 \times 10^{-3}$	2.615	$5.6 \times 10^{-3}$	2.339	$0.8 \times 10^{-3}$
$\tau_{cbo}$ ( $\mu$ s)	175.2	46.8	99.4	0	137.4	62.0	200.3	33.5
$A_{cbo-N}$ ( $\times 10^{-4}$ )	43.1	5.0	42.8	9.9	39.3	9.7	32.3	2.0
$\phi_{cbo-N}$	-2.343	0.107	3.817	0.446	3.302	0.374	-0.710	0.062
$A_{2cbo-N}$ ( $\times 10^{-4}$ )	1.9	1.3	4.9	4.5	2.2	2.7	1.2	0.5
$\phi_{2cbo-N}$	3.331	0.638	5.665	1.274	-4.936	1.127	0.322	0.448
$A_{cbo-A}$ ( $\times 10^{-4}$ )	5.5	3.9	9.5	4.1	6.4	2.5	2.7	1.9
$\phi_{cbo-A}$	-0.271	0.737	-2.073	0.600	1.750	0.561	-2.825	0.686
$A_{cbo-\phi}$ ( $\times 10^{-4}$ )	8.0	4.2	5.7	4.4	8.8	3.1	1.9	1.9
$\phi_{cbo-\phi}$	-1.183	0.533	1.227	0.920	4.313	0.415	-1.576	0.995
$\kappa_{loss}$	8.974	0	5.651	0	2.510	0	2.345	0

Table 5.7: Fit parameters for the four Run 1 precession frequency analysis datasets for a single random seed. The bold row highlights the final fitted  $R$  values and their respective errors. The 60h dataset has a different blinding to the rest. The  $\kappa_{loss}$  parameter is fixed in each dataset fit corresponding to the 0 value in the error column, and similarly for  $\tau_{cbo}$  in the fit to the HighKick dataset.

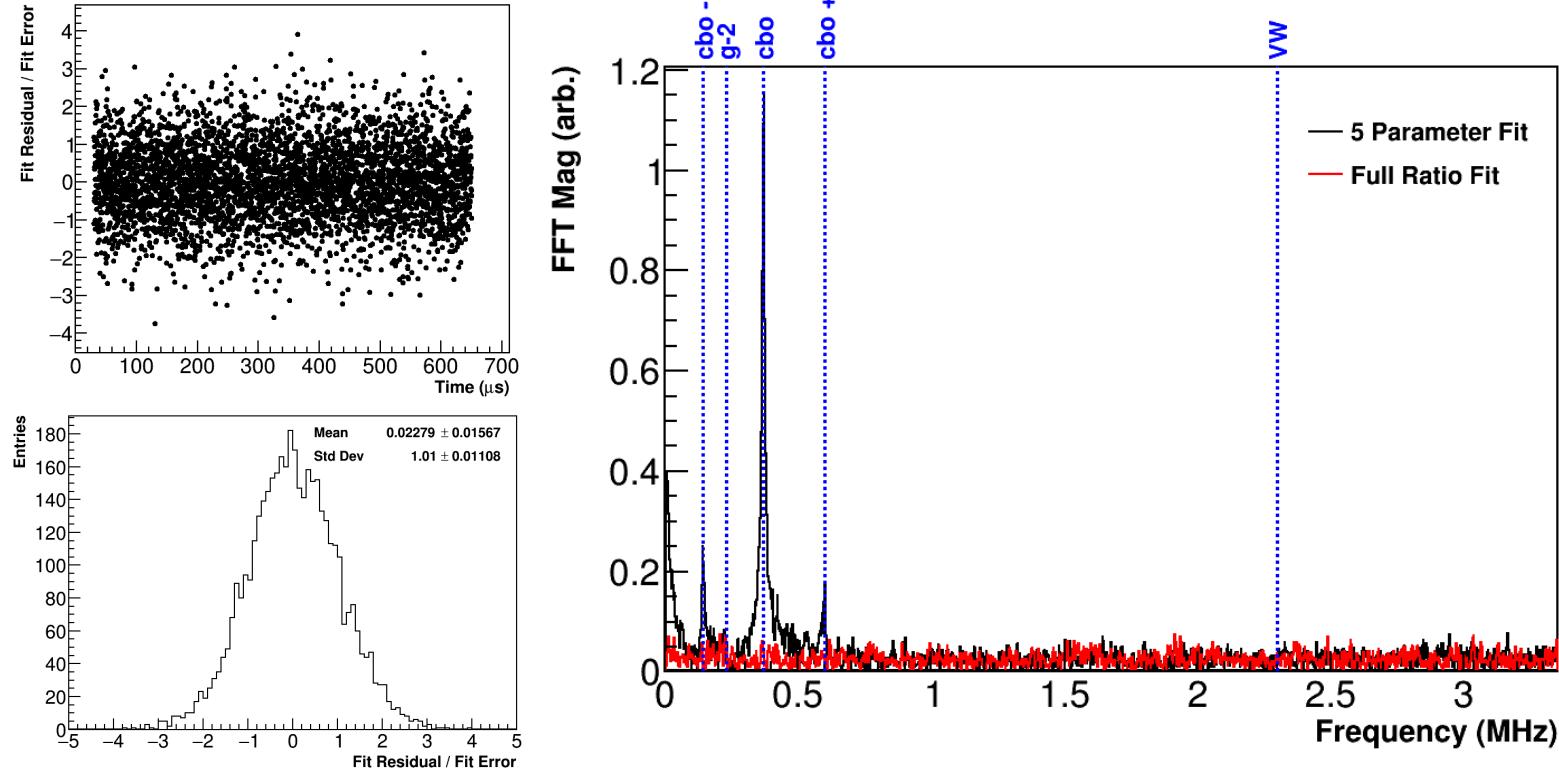


Figure 5.25: Fit pulls (top-left), their projection onto the y-axis (bottom-left), and the FFT of the fit residuals (right) for the ratio fit to the 60h dataset. Note the pull projection has a Gaussian shape centered around zero with unit width. In the FFT the results from a five parameter fit to the data are overlayed, along with blue dashed lines for the main beam dynamics peaks which appear in the data. There is no obvious structure in the pulls and no remaining peaks above the noise in the ratio fit FFT.

The CBO frequencies for the 60h and Endgame datasets with  $n$  values of 0.108 were found to be 2.338 and 2.339 rad/ $\mu$ s respectively, corresponding to approximately 0.37 MHz. For the HighKick and 9d datasets with  $n$  values of 0.120, the CBO frequencies were found to be 2.559 and 2.615 rad/ $\mu$ s respectively, corresponding to approximately 0.415 MHz. These frequencies correspond to the expected frequencies as described in Section 2.6, with some slight deviations due to statistics, bad resistors, and the reduced sensitivity in the Ratio Method<sup>16</sup>. The CBO lifetimes between the different datasets are relatively consistent, barring the HighKick dataset for which a smaller CBO lifetime was measured, attributed to the changing behavior of the bad resistors in that dataset. Ratio Method fits typically converge with lifetimes with large errors compared to T-Method fits, due to the reduction in sensitivity in the Ratio Method. In the HighKick dataset, the CBO lifetime did not like to converge nicely in the ratio fits, and was therefore fixed to that from a T-Method fit. The main CBO amplitudes  $A_{cbo-N}$  for the different datasets were on the order of 0.3–0.4%, while the higher order CBO amplitudes were in general an order of magnitude less. The strength of the various higher order CBO amplitudes fluctuated for different datasets, with one parameter being large compared to another in one dataset and vice versa in a different dataset. In some cases, the errors on the higher order CBO term amplitudes were of the order the amplitude itself. While this implies these terms can be dropped from the fit function, all terms were included for analysis uniformity among the different datasets. These relatively large errors, while making some of the fits slightly more challenging to achieve convergence, were nonetheless handled appropriately.

The final statistical errors on  $R$  for the 60h, HighKick, 9d, and Endgame datasets are 1.358, 1.411, 0.903, and 0.639 ppm respectively. The single seed  $R$  results for the

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<sup>16</sup>The VW frequencies, though time-randomized out in the analysis presented here, were found to be approximately 2.30 and 2.04 MHz for the datasets with  $n = 0.108$  and  $n = 0.120$  respectively.

HighKick, 9d, and Endgame datasets, all of which used the same blinding string, are all well within  $1\sigma$  of each other. The average  $R$  value for fits to 50 different random seeds are provided in Section 5.4.5.

Beyond looking at single fit residuals to evaluate the integrity of the fits, other checks were made to verify consistency. In general this consisted of slicing up the data in different ways and fitting the subsets. These tests and scans included fitting individual calorimeters, modifying the fit start and end times, adjusting the applied energy thresholds, and fitting individual beam bunches.

### 5.4.1 Individual calorimeter fits

Fits to all 24 individual calorimeters for each of the datasets were performed with the same number of free fit parameters as used in the calorimeter sum fits. Figure 5.26 shows the  $\chi^2/\text{NDF}$ 's for the calorimeter fits which are nicely spread around 1. Figure 5.27 shows the fitted  $R$  values as a function of calorimeter number. Straight line fits were performed to the  $R$  values with good  $\chi^2$ 's, and the fitted constant returned a value in each case that was consistent with the calorimeter sum fit  $R$  values. Examining the  $R$  values as a function of calorimeter between datasets, particular calorimeter numbers do not tend to lie above or below the fitted average. The spread in  $R$  values for each calorimeter then can be said to be driven statistically, though it should be noted that with the larger error bars on the individual calorimeter fits it's hard to tell if there are any preferences one way or another.

Figures 5.28 and 5.29 show calorimeter fit results for the other free parameters in the fit for the Endgame dataset. The  $g - 2$  phases are seen to be distinct among the different calorimeters, attributed to the different level of material upstream and therefore different acceptance. This is particularly noticeable in calorimeters 13 and 19 which sit behind the tracker stations. Any correlated effects on  $R$  are not immediately observed, and might potentially be hidden behind the large errors of the fit. Similarly, the different calorimeters have different fit asymmetries, once again due to their different acceptances. The CBO parameters are in general consistent with some spread due to acceptance, with the phases running from  $0-2\pi$  around the ring as expected. As one might notice, the amplitudes of the CBO parameters are an order of magnitude higher than in the calorimeter sum fits. Because the phases vary around the ring, when adding up all the calorimeters the CBO effect becomes reduced. In fact, while it is not always necessary to include the higher order CBO terms for good fits to the calorimeter sum data, there are many calorimeters which need the higher



Figure 5.26:  $\chi^2/\text{NDF}$  versus calorimeter for the Run 1 precession frequency analysis datasets. Red dashed lines are placed at  $\chi^2/\text{NDF} = 1$  to aid the eye. No individual calorimeter fits are preferentially low or high when comparing across datasets.



Figure 5.27:  $R$  versus calorimeter for the Run 1 precession frequency analysis datasets. The scale is the same on each of the plots. A straight line fit was performed on the fitted values, with the fit result shown in the upper right box as parameter  $p_1$  in units of ppm. The different blinding in the 60h dataset is readily observed, along with the higher precision fits in the 9d and Endgame datasets with their correspondingly smaller error bars.

order terms for good fits.

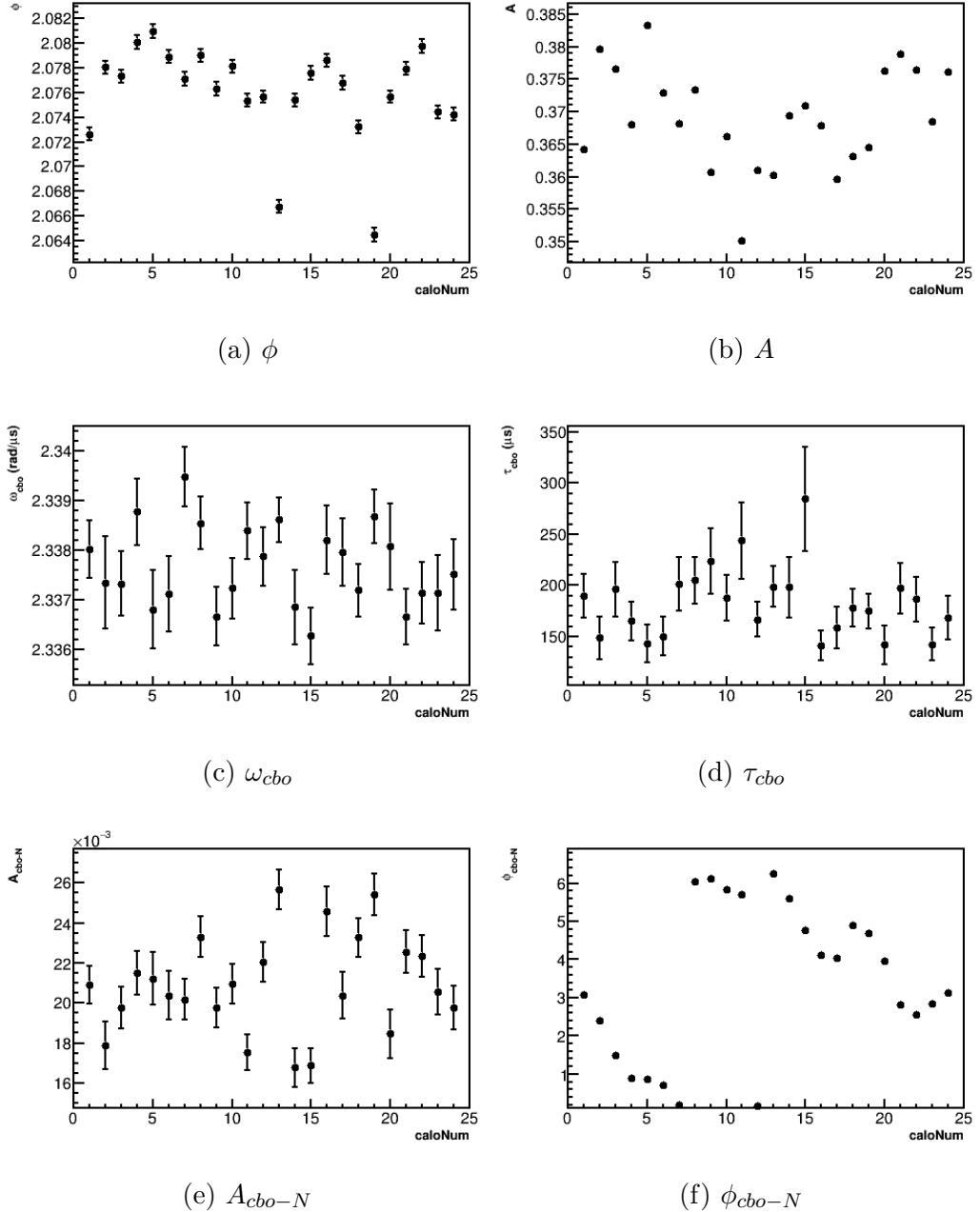


Figure 5.28: Endgame fit parameters versus calorimeter number. In the  $g - 2$  phase  $\phi$  (top-left) two points lie below the others, corresponding to those calorimeters which sit behind tracker stations. The CBO phase  $\phi_{.cbo-N}$  (bottom-right) runs from  $0-2\pi$  around the ring. These plots are typical of all datasets, with small variations in the final fitted parameters.

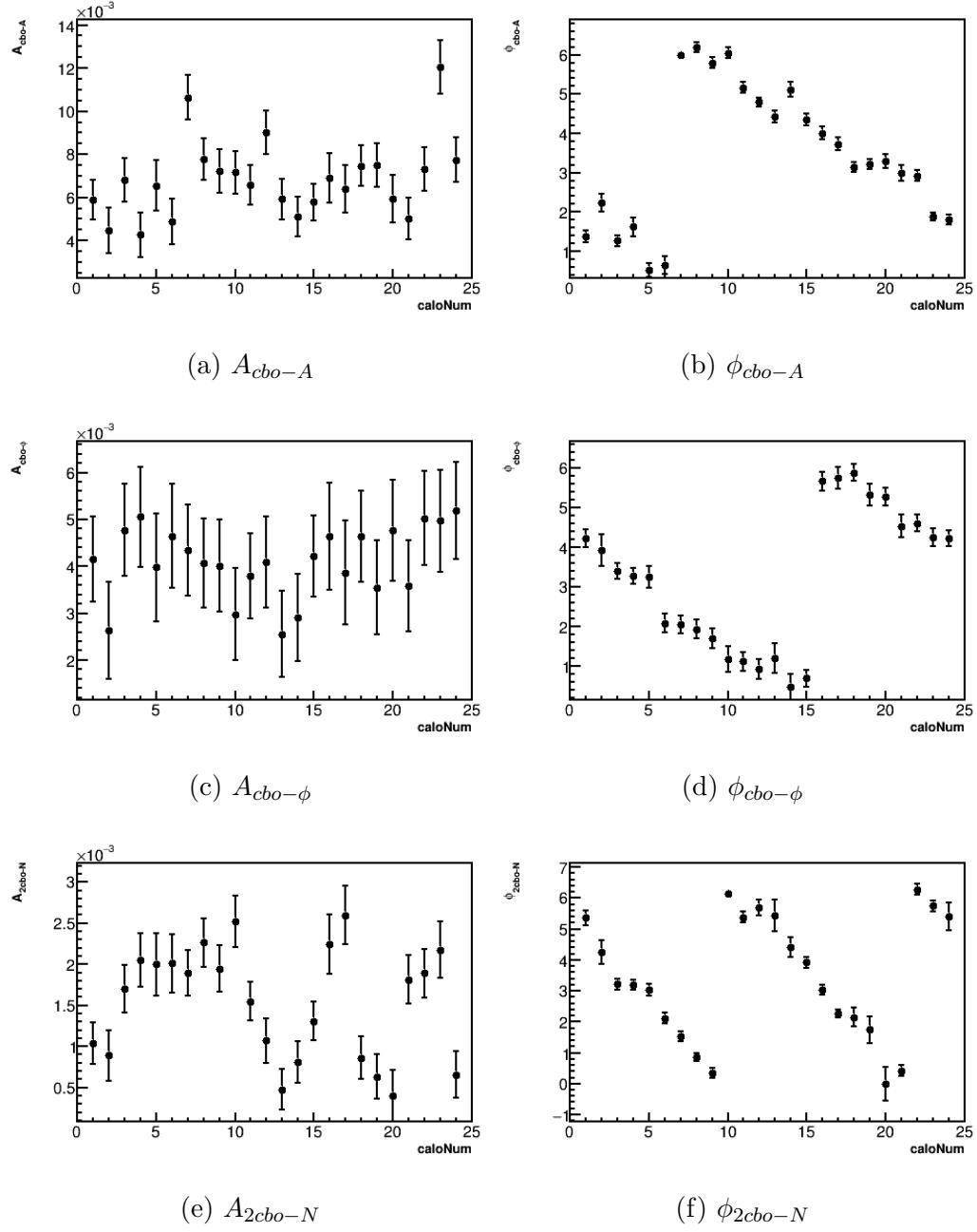


Figure 5.29: Endgame fit parameters versus calorimeter number. The CBO phases  $\phi_{cbo-A}$  and  $\phi_{cbo-\phi}$  run from  $0-2\pi$  around the ring, while  $\phi_{2cbo-N}$  runs around twice. These plots are typical of all datasets, with small variations in the final fitted parameters.

### 5.4.2 Fit start and end scans

In order to determine if there are any deficiencies as a function of time, for instance if the CBO was modeled incorrectly, fits are performed with varying fit start and end times. If a parameter is incorrectly modeled, it will wander away from the statistically allowed deviation as the mis-modeled effect grows stronger or weaker. Two examples of this are given in Figure 5.30. In the first case the CBO frequency was mis-measured when excluding the changing CBO frequency model, and in the second case the asymmetry was mis-measured when pileup was not subtracted from the data. In general, doing a fit start scan is a very useful tool in the precession frequency analysis beyond just verifying consistency of a fit parameter as a function of time, as it also provides hints as to what might be wrong with how the data is being handled.

The statistically allowed deviation between two sets of  $g - 2$  data, where one is a subset of the other, is given by [29]

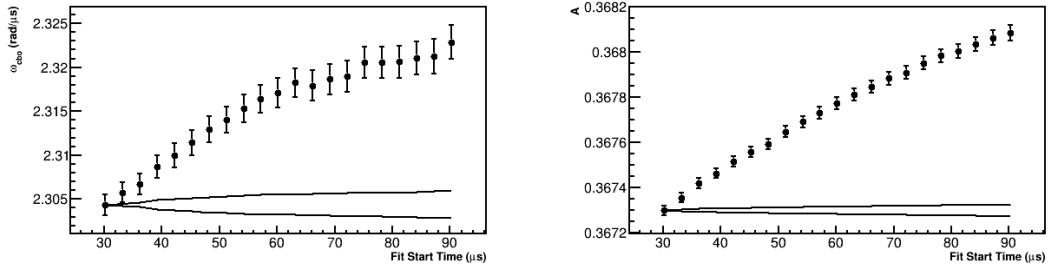
$$\sigma_{diff} = \sqrt{\sigma_2 - \sigma_1(2\frac{A_1}{A_2} \cos(\phi_1 - \phi_2) - 1)}, \quad (5.30)$$

where the subscript 2 stands for the larger dataset while the subscript 1 stands for the smaller sub-dataset. This statistically allowed deviation depends both on the size of the datasets as well as their “analyzing powers,” which come from the asymmetries and phases of the datasets. For fit start scans the analyzing powers are in general the same, such that the approximation

$$\sigma_{diff} \approx \sqrt{\sigma_2 - \sigma_1} \quad (5.31)$$

can be made.

It should be noted that for fits with late fit start times, certain CBO parameters



(a) The fit start scan for the fitted CBO frequency when a constant frequency model is included. The p-value of the first ratio fit in this case is 0.191, indicating an acceptable fit. Only by performing a fit start scan or examining the fit residuals closely is the deficiency in the fit observed.

(b) The fit start scan for the fitted asymmetry when the pileup is not subtracted out. If the pileup is not properly accounted for, then decay positrons with lower asymmetries contaminate the observed decay time spectrum leading to a lower fitted asymmetry value. As the pileup diminishes the fitted asymmetry tends to its true value.

Figure 5.30: Examples of fit start scans when effects are improperly accounted for in fitting. The parabolic bands indicate the one sigma allowed deviation in the fit parameters due to the changing statistics between the fits. Data are from the Endgame dataset.

start to become unstable as the CBO effect becomes diminished in the data<sup>17</sup>. Fits with start times at 100  $\mu$ s are half a lifetime or more along the CBO effect and convergence is difficult to achieve. For those parameters which were found to be unreliable, they were fixed to their starting fit values. These typically included the CBO lifetime and most of the higher order CBO amplitudes and phases, though in some cases they could still be included out to late fit start times depending on the dataset.

Fit start time scans were performed from the default value of 30.2  $\mu$ s up to 100.2  $\mu$ s in steps of 1  $\mu$ s corresponding to 71 separate fits. In order to assist fit convergence, the final fit parameters from one fit were passed on as the starting parameters of the next. The  $\chi^2/NDF$ 's and fitted  $R$  values as a function of fit start time for the four Run 1 precession frequency analysis datasets are shown in Figure 5.31. The error on the individual points in the  $\chi^2$  plots are given as

$$\sigma_{\chi^2/NDF} = \sqrt{2/NDF}, \quad (5.32)$$

where NDF changes as the fit start time is pushed later in time and bins are left out of the fit. The black parabolic bands indicate the  $1\sigma$  statistically allowed deviation as given by Equation 5.31. As shown the goodness of fit for the four datasets are all consistent with fit start time, only wandering in and near the bands without diverging.

$R$  is similarly consistent as a function of fit start time. The only dataset for which  $R$  goes noticeably outside the bands more than the others is that for the HighKick. Since the deviation is less than  $2\sigma$  and it appears to have a trend back towards the bands however, it does not appear particularly indicative of unaccounted effects in the data. In general  $R$  is very insensitive to changing fit start time, most likely due to the small correlations with other fit parameters.

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<sup>17</sup>The same applies to the VW effect when it is not time-randomized out of the data.

Figure 5.32 shows the fit start scan results for the other free parameters for the Endgame dataset. In all cases the fit parameters only wander in and near the bands, showing that all effects in the ratio data are properly accounted for.

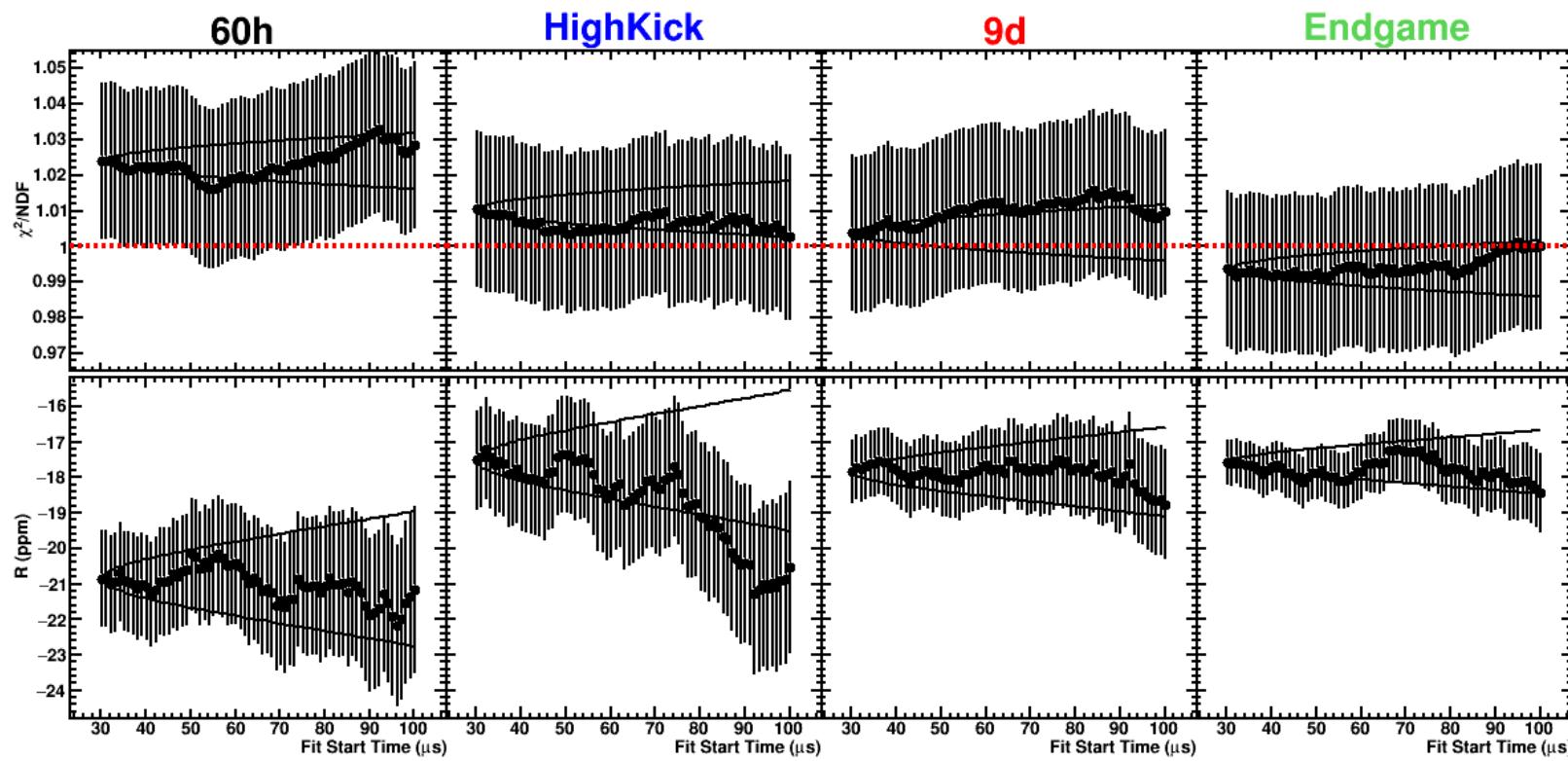


Figure 5.31:  $\chi^2/\text{NDF}$  (top) and  $R$  (bottom) versus fit start time for the Run 1 precession frequency analysis datasets. Plots in the top and bottom rows are on the same scale respectively. The black parabolic bands represent the  $1\sigma$  statistically allowed deviation. The  $\chi^2$  values are all consistent with 1, for which a dashed red line has been overlayed. In all cases the fit points lie in and around the  $1\sigma$  statistical bands.

For fit end scans, all of the same methods and conclusions apply. In general fit end scans are both less dangerous and more stable than fit start scans, as the amount of data being removed from the fit is relatively small. While these fit end scans in a T-Method fit might be able to be completely ignored in favor of the fit start scans, it's important to check that they satisfy the statistical deviations in the Ratio Method as the ratio data errors grow larger with less data [102]. Fit end time scans were performed from 650  $\mu$ s to 400  $\mu$ s in steps of 10  $\mu$ s, corresponding to 26 separate fits. As in the fit start time scan, fit results from the end of one fit were passed on as the starting parameters to the next.  $R$  values for fit end scans for the Run 1 precession frequency analysis datasets are shown in Figure 5.33. As shown the  $R$  values are comfortably within and near the bands.

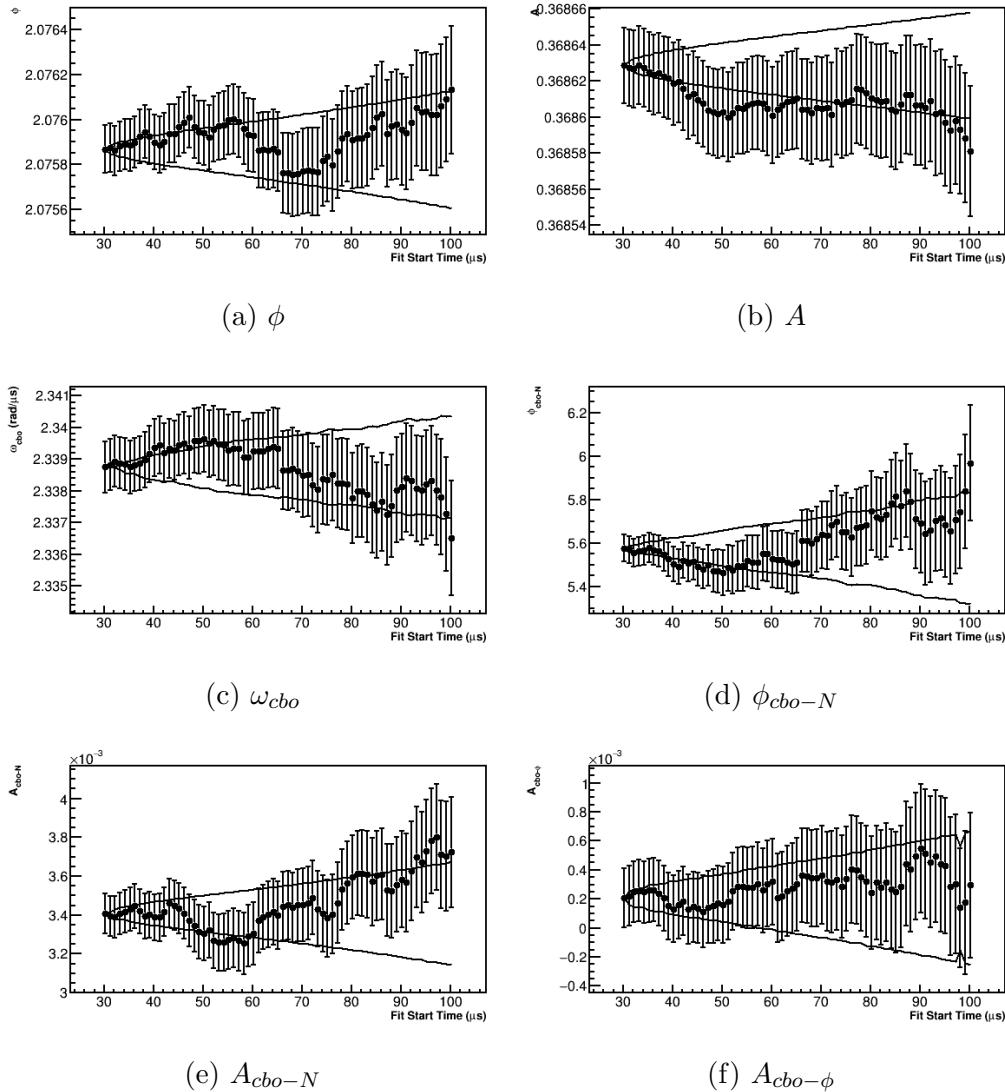


Figure 5.32: Fit start scans for free parameters in the Endgame dataset. Those parameters not shown here are fixed to their starting values over the course of the scan, as at late times they can be unstable as the CBO effects die away.

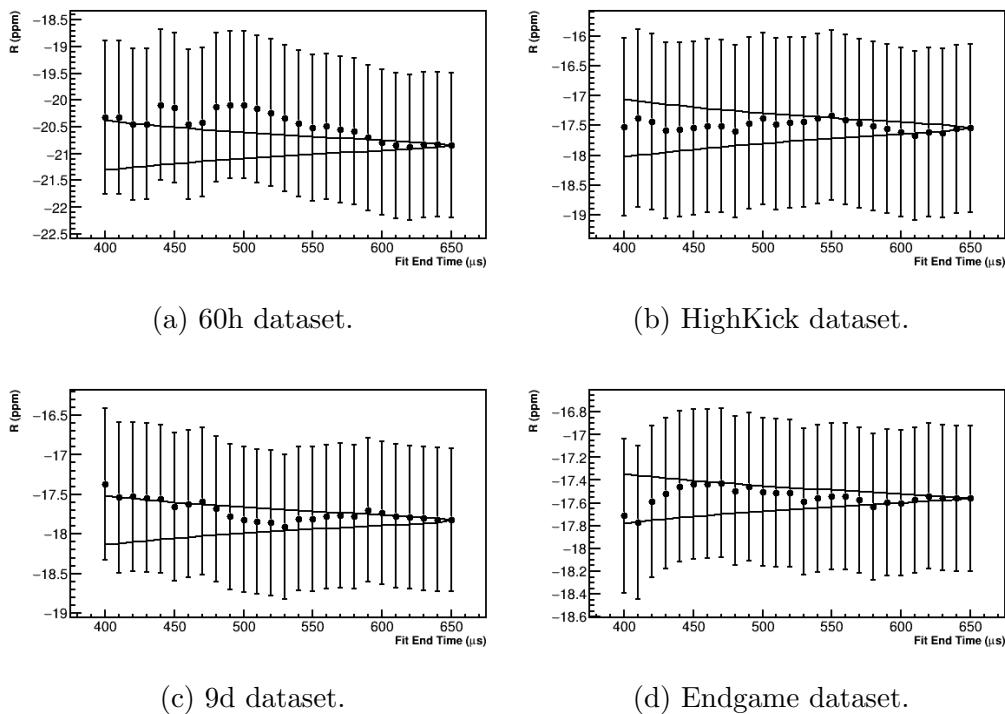


Figure 5.33:  $R$  versus fit end time for the Run 1 precession frequency analysis datasets. The fit points lie in and around the  $1\sigma$  statistical bands.

### 5.4.3 Energy threshold scans

Similarly to fit start and end time scans, it is worthwhile to verify that  $R$  is consistent regardless of the energy threshold applied to the decay positron time spectrum. The energy threshold was varied from 1.2 GeV to 2.2 GeV in steps of 50 MeV corresponding to 21 separate fits. The fitted  $R$  values for the four Run 1 precession frequency analysis datasets are shown in Figure 5.34. The statistical bands are the same as those defined in Equation 5.30, where now the analyzing power part of the equation plays a larger role as the asymmetries and phases of the different fit points are significantly different. As shown there are no major deviations in the fitted  $R$  values.

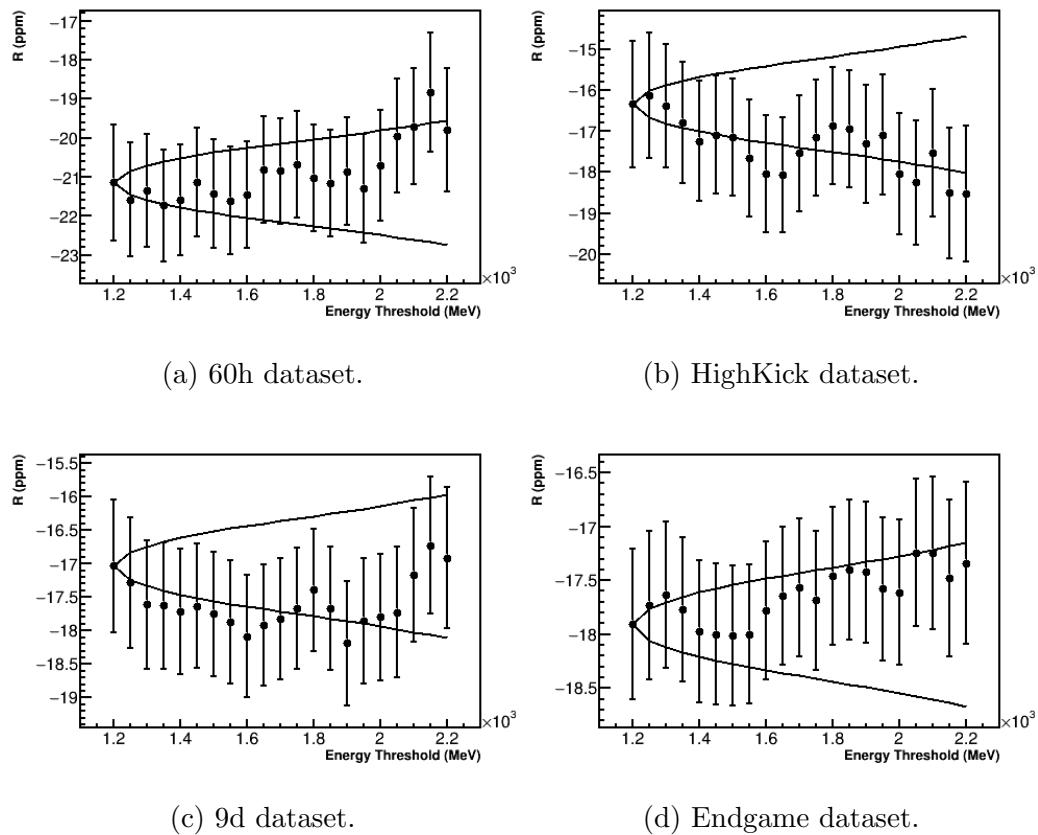


Figure 5.34:  $R$  versus energy threshold for the Run 1 precession frequency analysis datasets. The fit points lie in and around the  $1\sigma$  statistical bands.

#### 5.4.4 Fits to bunch number

As described in Section 2.3, eight distinct and separate bunches of muons are sent to the E989 experiment within the accelerator timing structure. Differences upstream and at injection mean that the stored beam is slightly different between the bunches. In order to verify that the treatment of fit parameters was correct regardless of the bunch number and improve confidence in the results, the bunches were fit individually. Figure 5.35 shows the fitted  $R$  values for the eight individual bunches alongside the bunch-sum result, for the Run 1 precession frequency analysis datasets. In all cases there appear no systematically different  $R$  values per bunch. When the eight individual bunches were fit to a straight line, they were found to be consistent with the bunch-sum result.

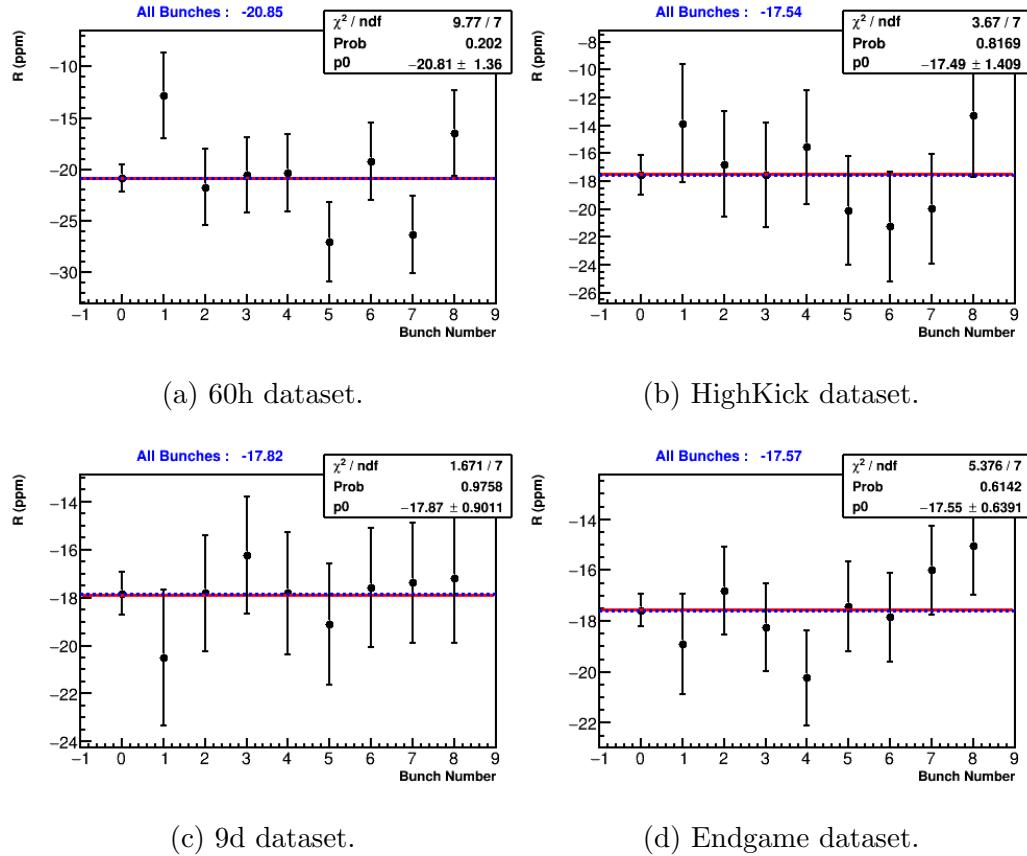


Figure 5.35:  $R$  versus bunch number for the Run 1 precession frequency analysis datasets. Bunch number 0 corresponds to the data from all bunches added together. The blue dashed line intersects the bunch number 0 point, and its value is displayed in the top left of each plot. The red line corresponds to a fit to bunches 1–8, with  $p_0$  being the fit parameter. In all cases the fitted  $R$  value to bunches 1–8 is nearly identical to the all-bunches result.

<b>Random Seed Fit Results</b>				
Dataset	$\chi^2$ Mean	$R$ Mean	$R$ RMS	$R$ Error on Mean
60h	0.999	-20.5562	0.3443	0.0487
HighKick	1.001	-17.4755	0.4226	0.0598
9d	0.999	-17.7182	0.2118	0.0300
Endgame	1.002	-17.3406	0.1249	0.0177

Table 5.8: Random seed fit results to the four Run 1 precession frequency analysis datasets. The  $\chi^2$  means are consistent with 1. As a reminder the 60h dataset used a different blinding than the other three, hence the significantly different  $R$  mean. Units are in ppm.

#### 5.4.5 Fits to many random seeds

While the single seed fit results presented earlier indicate good fits and well understood parameters, it is always a good idea to fit other random seeds in case the single seed results ended up on an outlier. Doing so not only improves the confidence of the result, but also gives a more central  $R$  value to quote as being closer to the ‘true’  $R$  of the dataset. Figures 5.36 and 5.37 give the  $\chi^2$  and  $R$  distributions for fits to 50 different random seeds for the four datasets. As shown the  $\chi^2$  distributions are centered around 1 as expected. Table 5.8 compares the random seed fit results between the different datasets. The means for the  $R$  distributions of the datasets which shared the same blinding string (HighKick, 9d, Endgam) are statistically consistent with one another, with differences on the order of several hundreds of ppb, well within the statistical errors of 600 ppb or greater<sup>18</sup>.

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<sup>18</sup>The errors on the mean are calculated as  $\sigma_\mu = \text{RMS}/\sqrt{N}$ , where  $N$  is the number of random seeds. These errors come from the randomization of cluster times before fitting, and have no bearing on statistical consistency when comparing different datasets.

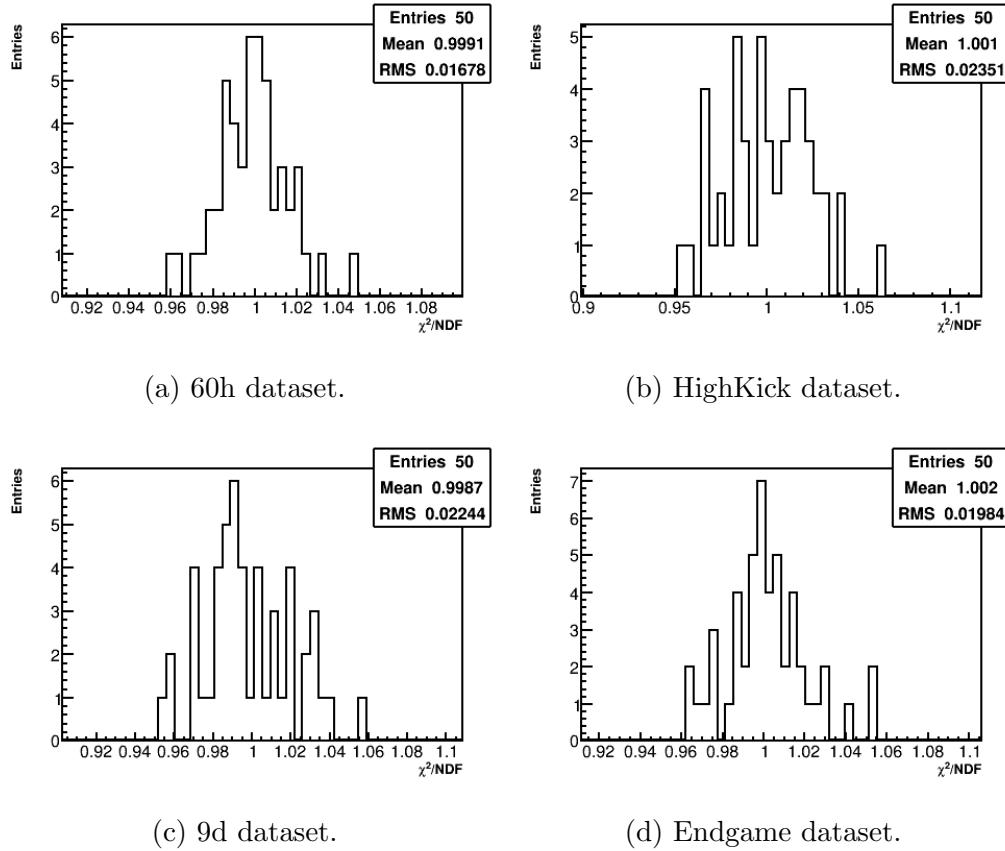


Figure 5.36:  $\chi^2$  values for fits to 50 different random seeds for the Run 1 precession frequency analysis datasets. The distributions are nicely centered around 1 which is to be expected if the randomized data is properly distributed and fit correctly.

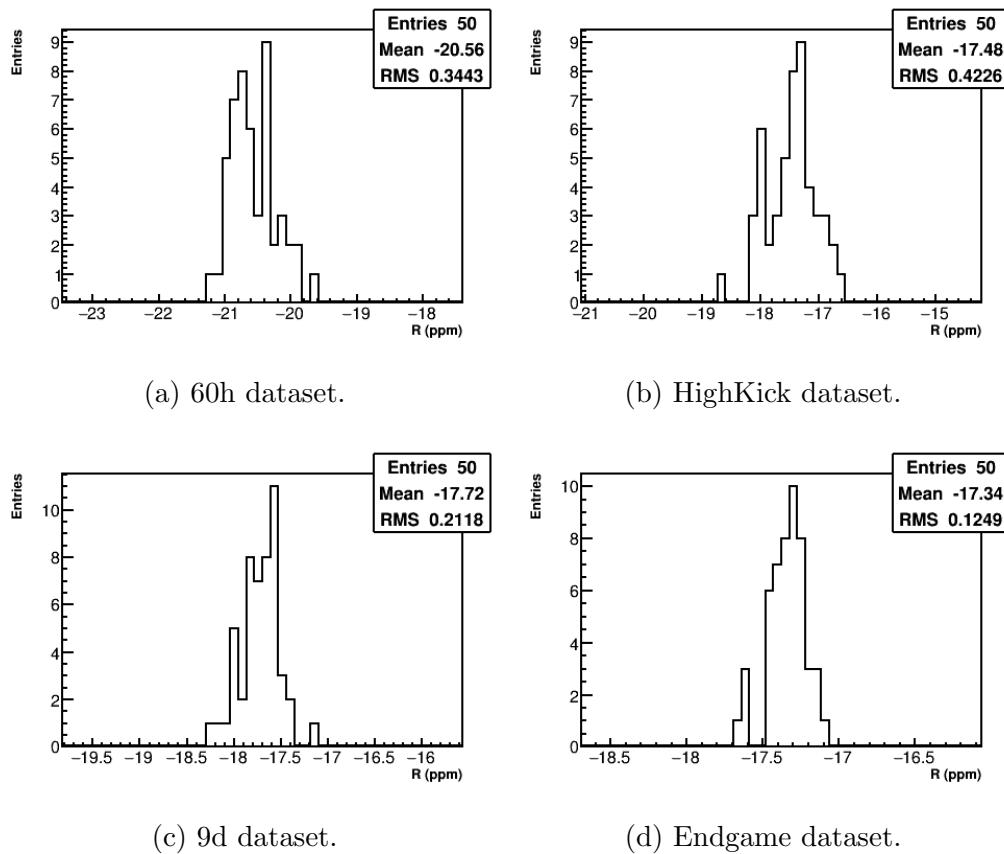


Figure 5.37:  $R$  values for fits to 50 different random seeds for the Run 1 precession frequency analysis datasets. The statistics box has units of ppm for the mean and RMS of the distributions.

Target $\omega_a$ Systematic Uncertainties	
Sources of uncertainty	E989 Goal (ppb)
Pileup	40
Gain changes	20
CBO	30
Lost muons	20
E-field and pitch corrections	30
Other	30
Quadrature sum	$\sim 70$

Table 5.9: Target systematic errors in the precession frequency measurement for the E989 experiment [38].

## 5.5 Systematic errors

Evaluating the systematic errors which bias the  $\omega_a$  frequency extraction is a major part of the precession frequency analysis. Systematic errors typically stem from sources which introduce time-dependent phases over the course of a fill,

$$\cos(\omega_a t + \phi(t)) \rightarrow \cos((\omega_a + \phi_1)t + \phi_0 + \phi_2 t^2 + \dots), \quad (5.33)$$

where  $\phi(t)$  can in general be expanded to higher orders. Sources of systematic uncertainty can be separated into several general categories. These include systematic errors in the pileup subtraction, gain corrections, CBO fit model, muon losses, and E-field and pitch corrections, among others. In each case early-to-late phase shifts are a danger in the analysis and must be evaluated appropriately. Table 5.9 gives the target systematic uncertainties for the E989 experiment, which has the goal of a 70 ppb systematic error on the precession frequency measurement. This represents a three to four-fold improvement over the E821 experiment systematic errors which were of order 210 – 310 ppb depending on running period [29].

Run 1 of E989 consisted of more than four separate datasets with distinct running

conditions, necessitating the need for independent systematic uncertainty evaluations. Such uncertainty evaluations for four of the Run 1 datasets are presented here. Systematic errors for the lost muon phase error, the E-field and pitch corrections, and beam movement are dependent on analyses from other working groups. The numbers given here for them are preliminary estimations.

### 5.5.1 Pileup systematic errors

As described in Section 5.2.1, the pileup background oscillates at  $\omega_a$  which by extension means a strong effect on the final fitted  $R$  value. If the subtracted pileup spectrum is mis-constructed in any way, there will be a systematic error on  $R$ , Section 5.2.1. In general the pileup systematic error can be separated into two parts, those being the errors on the amplitude and phase of the correction respectively. In order to estimate the two parts, the uncertainties on the pileup amplitude and phase need to be estimated along with the sensitivities of  $R$  to them. Table 5.1 gives the default values used for the pileup construction parameters {ADT, SDT, SGT, C}. How these parameters feed into the amplitude and phase systematic errors will be discussed in turn, and the overall errors calculated for the different Run 1 datasets.

As a reminder the default values used for the ADT and SDT were 5 ns each, and a default automatic pileup amplitude multiplier of  $\sim 1.03$  was applied to the pileup spectra. In order to calculate the systematic dependence on the choice of ADT or SDT, the SDT parameter was scanned from 5 ns to 10 ns in steps of 1 ns. This was done with and without the same automatic pileup amplitude scaling procedure as described in Section 5.2.1. The results of the study for the 9d dataset are shown in Figures 5.38 and 5.39. In the case where there was no automatic scaling applied there is a clear minimum in the  $\chi^2$  results and a steep slope in  $R$  corresponding to a large sensitivity of  $R$  to the choice of SDT. In the case where the automatic scaling was applied however, the minimum in the  $\chi^2$  results has disappeared, while the sensitivity

of  $R$  has become much reduced to the point of no longer being a clear trend. Applying the automatic pileup amplitude scaling produces nearly identical pileup spectra with no clear trend in  $R$  regardless of the choice of SDT (and by extension ADT). Any systematic error due to the choice of these two parameters can then be subsumed into the direct pileup amplitude error itself, discussed below. It should be noted that the choices of ADT and SDT are largely irrelevant barring statistics, as the automatic amplitude scaling procedure can always account for any differences between the two.

In order to calculate the systematic dependence on the choice of SGT (default value of 10 ns), the SGT parameter was scanned from 10 ns to 20 ns in steps of 1 ns. The results of the study with the automatic pileup amplitude scaling applied are shown in Figure 5·40. Just as in the SDT scan with the automatic pileup amplitude scaling, there is no minimum in the  $\chi^2$  results, the sensitivity of  $R$  to the value of SGT is negligible, and the pileup spectra for the various choices of SGT are nearly identical. Therefore again any systematic error due to the choice of SGT is subsumed into the pileup amplitude error.

The pileup amplitude systematic error is the error on  $R$  assuming the scale of the pileup was incorrectly constructed. In order to evaluate this error, multipliers were applied to the pileup spectra from 0.9 to 1.1 in steps of 0.01 (dropping the default automatic pileup scaling of  $\sim 1.03$  mentioned before). The data was then re-fit to find the change in  $R$ . The results of the study for the 9d dataset are shown in Figure 5·41. As shown there is a clear minimum near 1 in the  $\chi^2$  results and a non-zero sensitivity of  $R$  to the multiplier. The systematic error on  $R$  is calculated as

$$\delta R = \sigma_{P_m} \times \frac{dR}{dP_m}, \quad (5.34)$$

where  $P_m$  is the value of the pileup multiplier. The uncertainty  $\sigma_{P_m}$  is calculated as the width of the fitted parabola in the  $\chi^2$  plot, defined as the change in  $P_m$  from the

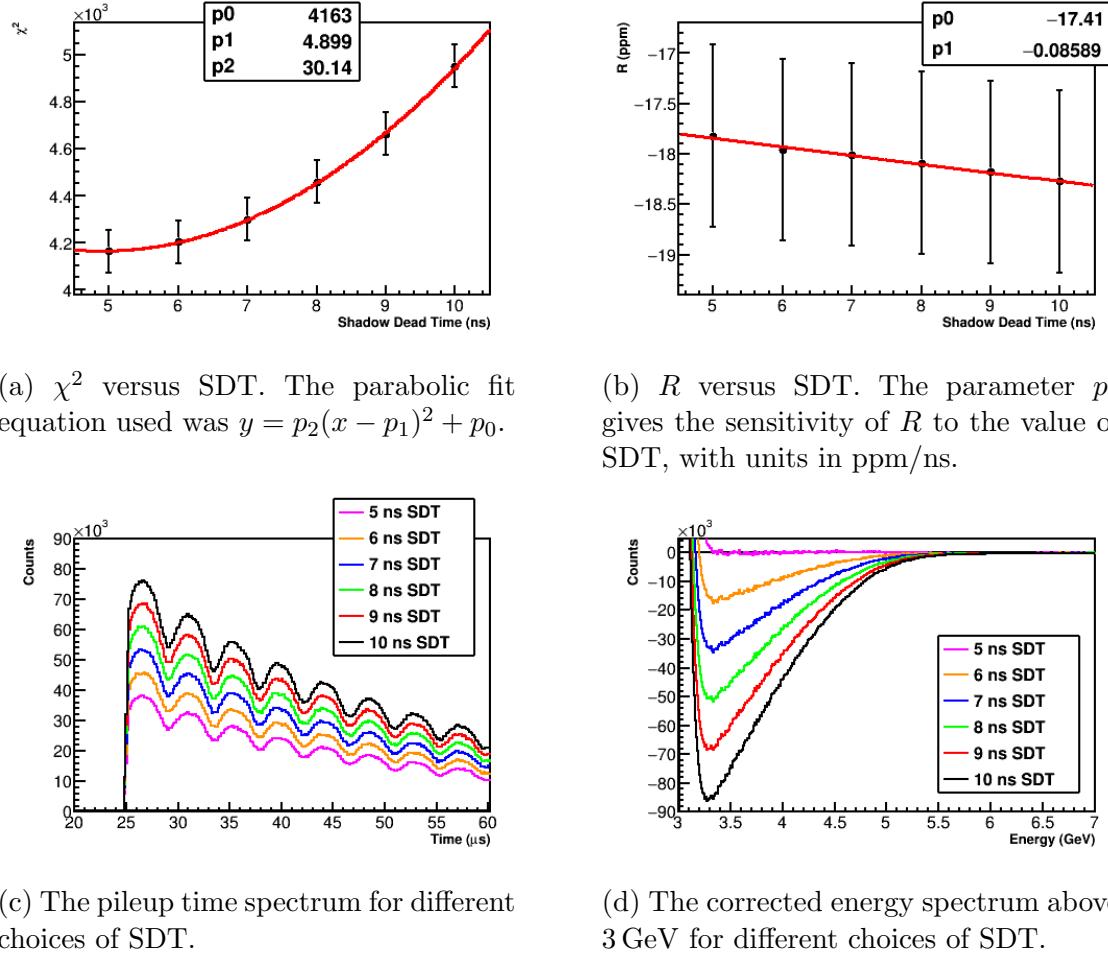


Figure 5.38: Shadow dead time scan results without automatic pileup amplitude scaling. A clear minimum in the  $\chi^2$  plot is seen near 5 ns corresponding to the choice of SDT = ADT, and a large sensitivity for  $R$  is observed. A larger choice of SDT leads to a greater estimation of the pileup, which as shown in the energy spectra plot leads to a corresponding over-subtraction at energies where pileup pulses dominate. Data are from the 9d dataset.

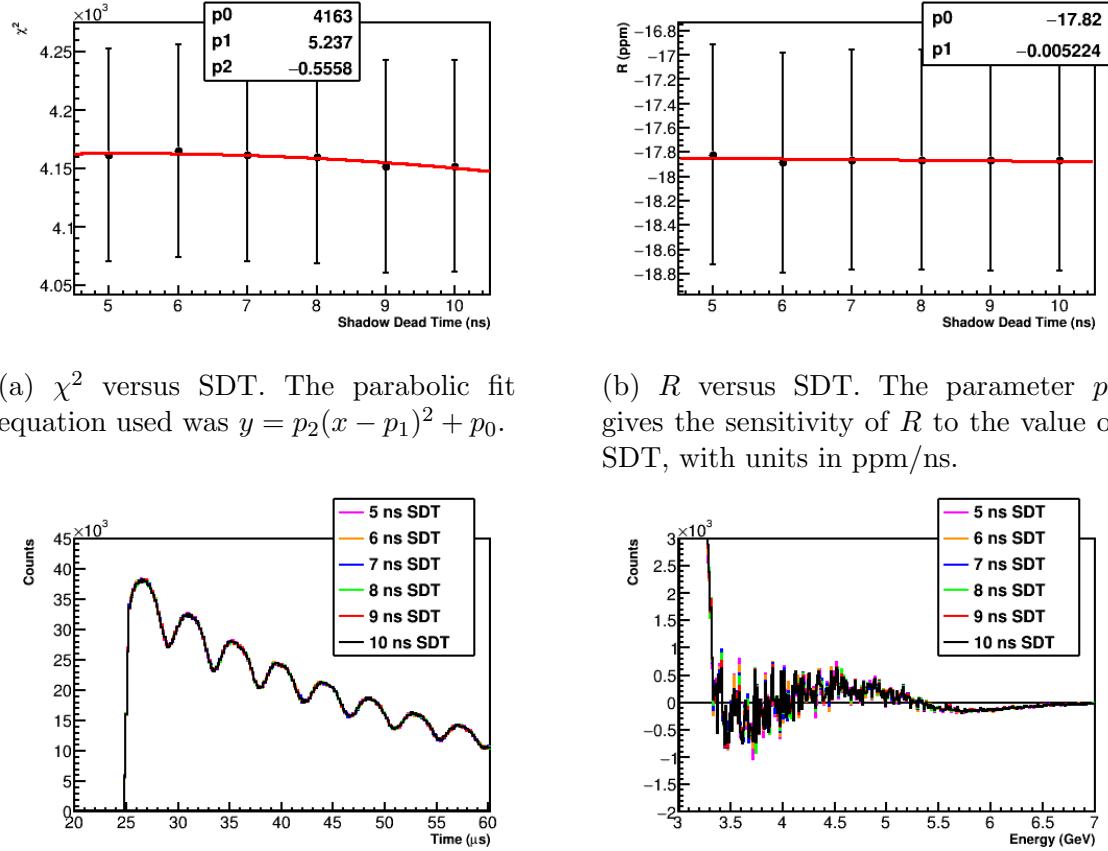


Figure 5.39: Shadow dead time scan results with automatic pileup amplitude scaling. No clear minimum is observed in the  $\chi^2$  plot, and the sensitivity for  $R$  is small. With the automatic amplitude scaling applied, the time and energy spectra are nearly identical and lie on top of each other. Combined with the lack of clear minimum in the  $\chi^2$  plot and no clear sensitivity in  $R$  this indicate that there is no real systematic error due to the choice of SDT. Data are from the 9d dataset.

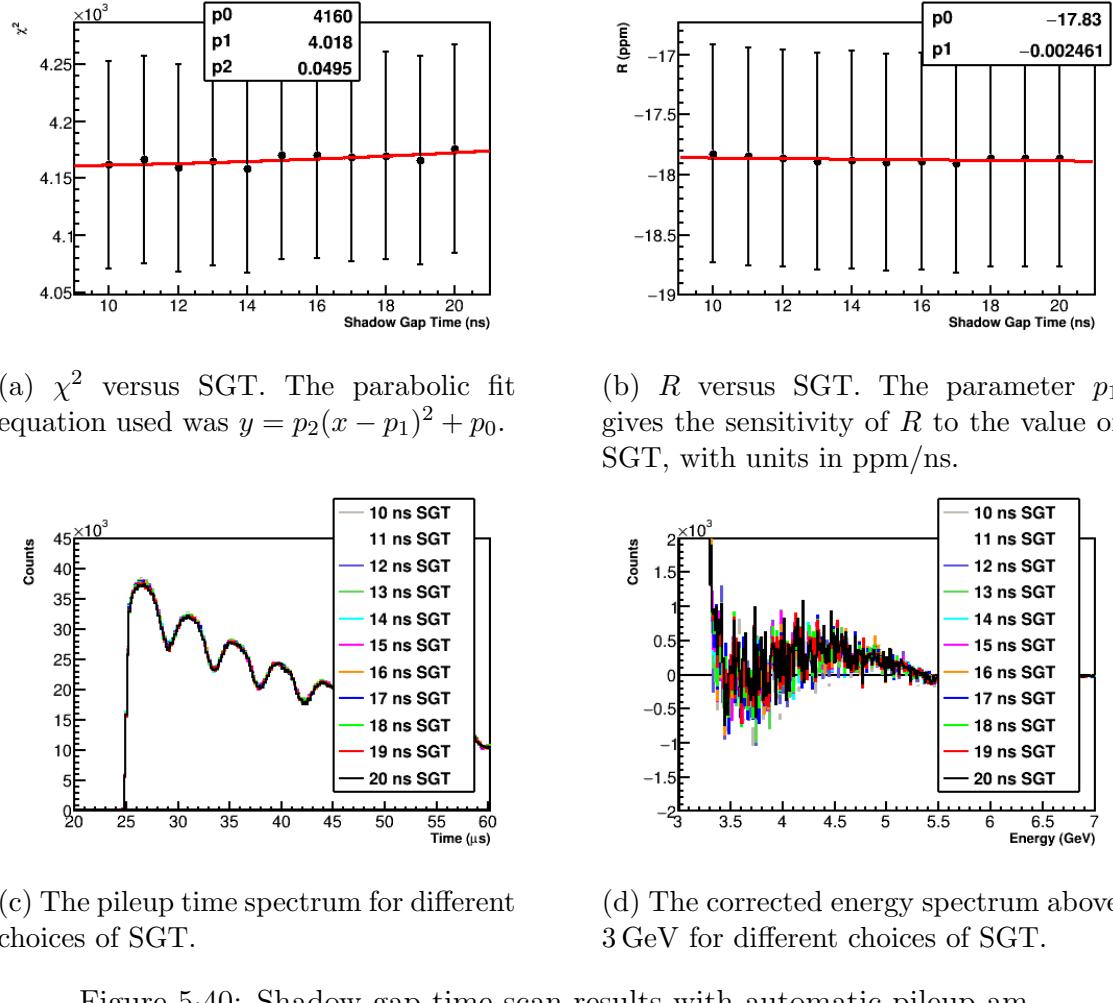
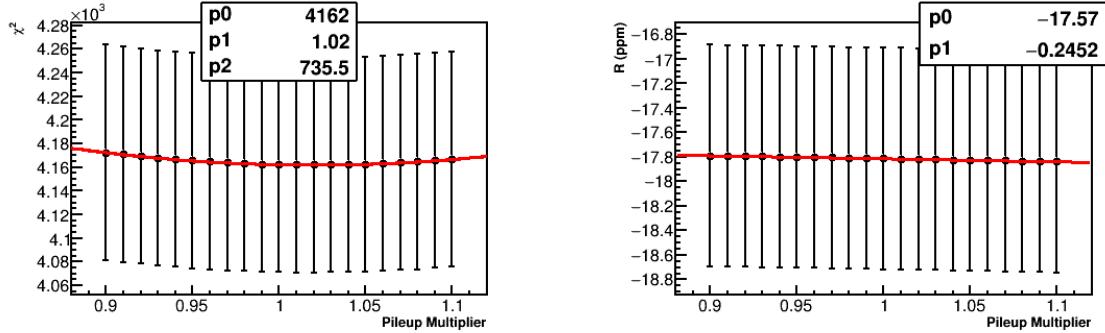


Figure 5.40: Shadow gap time scan results with automatic pileup amplitude scaling. No clear minimum is observed in the  $\chi^2$  plot, and the trend for  $R$  isn't clear, with points fluctuating above and below the fit curve. With the automatic amplitude scaling applied, the time and energy spectra lie on top of each other. Combined with the lack of clear minimum in the  $\chi^2$  plot and small sensitivity in  $R$ , this indicate that there is no real systematic error due to the choice of SGT. Data are from the 9d dataset.



(a)  $\chi^2$  versus pileup multiplier. The parabolic fit equation used was  $y = p_2(x - p_1)^2 + p_0$ . While the trend may initially appear somewhat flat, there is indeed a very clear minimum in the results when error bars are removed.

(b)  $R$  versus pileup multiplier. The parameter  $p_1$  gives the sensitivity of  $R$  to the value of the pileup multiplier, with units in ppm.

Figure 5.41: Pileup multiplier scan. Data are from the 9d dataset.

minimum for the  $\chi^2$  to increase by 1. This is calculated as

$$\sigma_{P_m} = \sqrt{\frac{2}{f''(\chi^2)}} = \frac{1}{\sqrt{p_2}}, \quad (5.35)$$

where  $p_2$  is the fit parameter as given in the top right of the  $\chi^2$  plot. The sensitivities of  $R$  to the pileup multiplier, uncertainties in the pileup amplitude, and final corresponding systematic errors for the Run 1 precession frequency analysis datasets are given in Table 5.10. As shown in the table, the uncertainties on the pileup amplitude are of order 2 to 5%, while the systematic errors on  $R$  are on the order of 10 to 20 ppb depending on dataset. It should be noted that the default automatic pileup multiplier of  $\sim 1.03$  does not necessarily correspond to the minimum in the  $\chi^2$  plot, but is within  $1\sigma$  of 1 or the minimum (except the Endgame which is closer to  $2\sigma$ )<sup>19</sup>.

The pileup phase error is the error on  $R$  assuming the phase of the pileup was

<sup>19</sup>Monte-Carlo tests with various random seeds showed this minimum fluctuating above and below 1. The distance from 1 therefore is not a good measure for the uncertainty in the pileup amplitude compared to the width of the  $\chi^2$  parabola fit.

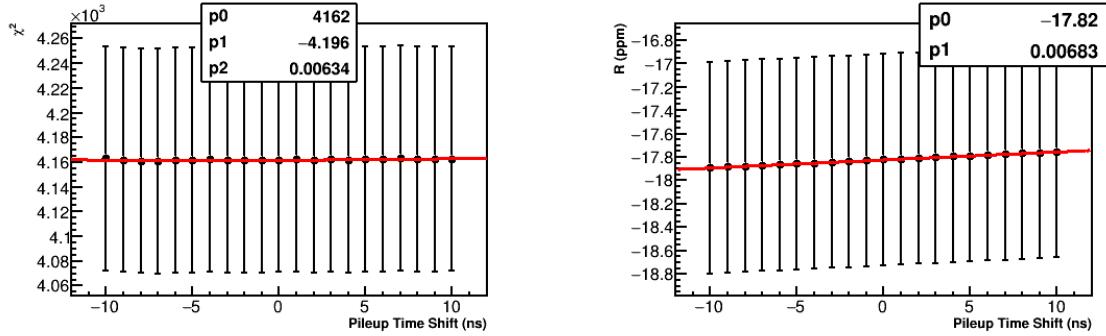
<b>Systematic Error due to Pileup Amplitude</b>				
Dataset	$dR/dP_m$	$\sigma_{P_m}$	$P_{m\min}$	$\delta R$
60h	-419.3	0.053	0.993	<b>22.2</b>
HighKick	-372.8	0.051	0.997	<b>19.0</b>
9d	-245.2	0.037	1.020	<b>9.0</b>
Endgame	-335.3	0.028	0.985	<b>9.4</b>

Table 5.10: Systematic error due to the pileup amplitude in the Ratio Method fits for the Run 1 precession frequency analysis datasets. The bold column gives the systematic error on  $R$ . Units for  $dR/dP_m$  and  $\delta R$  are in ppb.

incorrectly constructed. This can result from improper reconstruction of the pileup pulse times or energies. The error due to the former is calculated by applying time-shifts to  $t_{\text{doublet}}$  as given in Equation 5.3. Doing this artificially applies a phase shift to the pileup time spectrum. The data is then re-fit with the different pileup spectra and the change in  $R$  is calculated. Figure 5.42 shows the study results for the 9d dataset with time-shifts applied between  $-10$  ns and  $10$  ns in steps of  $1$  ns. A clear sensitivity of  $R$  to the value of time-shift is observed, however there is no clear minimum in the  $\chi^2$  results, implying the data is fit equally well with all pileup phases. Because the width of the  $\chi^2$  parabolic fit cannot be taken as the uncertainty in the pileup time-shift parameter, instead the uncertainty is taken conservatively at half the ADT at  $2.5$  ns. The systematic error is calculated in the same way as for the pileup amplitude uncertainty,

$$\delta R = \sigma_{P_t} \times \frac{dR}{dP_t}, \quad (5.36)$$

where  $P_t$  is the value of the pileup time-shift. The sensitivities of  $R$  to the pileup time-shift and corresponding systematic errors for the Run 1 precession frequency analysis datasets are given in Table 5.11. The sensitivities are on the order of  $5$ – $8$  ns, with systematic errors ranging from  $14$ – $19$  ppb.



(a)  $\chi^2$  versus pileup time-shift. There is no clear minimum in the plot.

(b)  $R$  versus pileup time-shift. The parameter  $p_1$  gives the sensitivity of  $R$  to the value of the pile time-shift, with units in ppm/ns.

Figure 5.42: Scan over pileup time-shift. Data are from the 9d dataset.

<b>Systematic Error due to Pileup Time Shift</b>		
Dataset	$dR/dP_t$	$\delta R$
60h	7.0	<b>17.6</b>
HighKick	7.6	<b>19.0</b>
9d	6.8	<b>17.1</b>
Endgame	5.7	<b>14.3</b>

Table 5.11: Systematic error due to the pileup time-shift parameter  $P_t$  in the Ratio Method fits for the Run 1 precession frequency analysis datasets. The bold column gives the systematic error on  $R$ . Units for  $dR/dP_t$  and  $\delta R$  are in ppb/ns and ppb respectively. The error on the  $P_t$  is by default taken to be 2.5 ns as described in the text.

Mis-reconstruction of pileup pulse energies means that pileup shadow doublets will be added or lost near the applied energy threshold, leading to an error in the energy-dependent pileup phase and by extension a systematic error on  $R$ . This would happen if the energy sum of two clusters on top of one another is different than two separated by a small time-gap as a result of detector performance. The energy of the shadow doublets is dependent on the constant  $C$  in the calculation of  $E_{\text{doublet}}$  as given in Equation 5.2. In order to calculate the systematic error from the energy construction, the parameter  $C$  was scanned from 0.9 to 1.1, in steps of 0.01. The results of the study for the 9d dataset are shown in Figure 5.43. The systematic error on  $R$  is calculated as

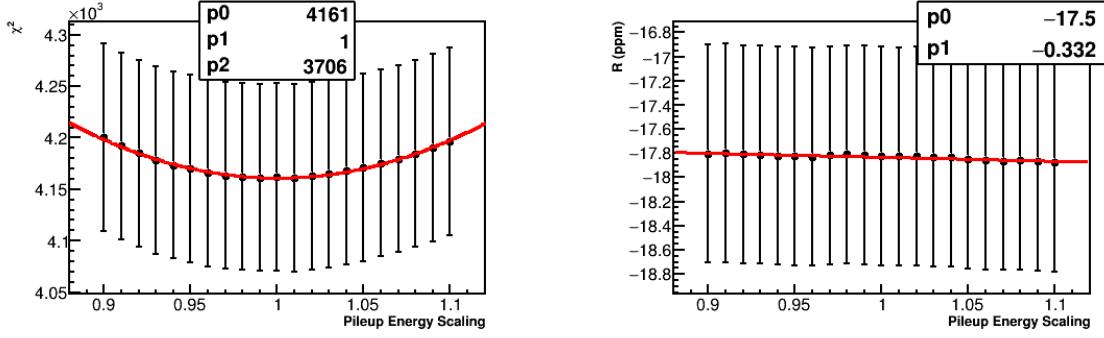
$$\delta R = \sigma_C \times \frac{dR}{dC}. \quad (5.37)$$

Similarly to the pileup amplitude error, there is a clear minimum in the  $\chi^2$  results which can be used to estimate the uncertainty in the pileup energy scale. Table 5.12 gives the sensitivities of  $R$  to the pileup energy scale, uncertainties in the pileup energy scale, and the corresponding final systematic errors for the Run 1 precession frequency analysis datasets. As shown the uncertainties on the pileup energy scale are of order 1 to 2%, and the value for  $C$  which produces the minimum in the  $\chi^2$  results is consistent with 1<sup>20</sup>. Interestingly, the sensitivity of  $R$  to  $C$  in the 60h dataset is noticeably larger than in the rest of the datasets; the origin of this is currently under investigation. Regardless, the systematic error is still small at 19.4 ppb.

Table 5.13 gives the quadrature sum for the total pileup systematic errors for the Run 1 precession frequency analysis datasets. As shown for each dataset the total error is below the target final error of 40 ppb in spite of the contamination in the pileup shadow method. For future runs of the experiment with increased muon

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<sup>20</sup>Because the spatial separation is turned off in the clustering portion of the reconstruction, a value of  $C = 1$  is to be expected.



(a)  $\chi^2$  versus pileup energy scale. The parabolic fit equation used was  $y = p_2(x - p_1)^2 + p_0$ .

(b)  $R$  versus  $C$ . The parameter  $p_1$  gives the sensitivity of  $R$  to the value of  $C$ , with units in ppm.

Figure 5.43: Scan over pileup energy scale. Data are from the 9d dataset.

Systematic Error due to Pileup Energy Scale				
Dataset	$dR/dC$	$\sigma_C$	$C_{\min}$	$\delta R$
60h	-835.1	0.023	0.997	<b>19.4</b>
HighKick	-167.7	0.022	0.995	<b>3.7</b>
9d	-332.0	0.016	1.000	<b>5.5</b>
Endgame	-431.4	0.012	0.982	<b>5.3</b>

Table 5.12: Systematic error due to the fixed pileup energy scale parameter  $C$  in the Ratio Method fits for the Run 1 precession frequency analysis. The bold column gives the systematic error on  $R$ . Units for  $dR/dC$  and  $\delta R$  are in ppb.

Total Pileup Systematic Errors					
Type of Error	Parameter	60h	HighKick	9d	Endgame
Amplitude	$P_m$	22.2	19.0	9.0	9.4
Phase (time)	$P_t$	17.6	19.0	17.1	14.3
Phase (energy)	$C$	19.4	3.7	5.5	5.3
Quadrature sum		34.3	27.1	20.1	17.9

Table 5.13: Total pileup systematic errors for the Run 1 precession frequency analysis datasets.

storage and therefore increased pileup, these errors may grow. In that case either the pileup shadow method might need to be improved to account for the contamination and pileup triplets, or discarded in favor of a different method.

### 5.5.2 Gain systematic errors

As described in Section 5.1, the energies of the positron hits are gain-corrected for in-fill, short-term double pulse, and out-of-fill effects. The latter is largely temperature dependent and occurs over time scales much longer than a fill. This does not bias the precession frequency measurement as the phase is not time-dependent over the course of a fill. For the cases of in-fill or STDP gain variations, any uncorrected fluctuations in the gain causes an effective change in the energy threshold over the course of a fill which then modifies the average measured phase of the detected positrons. This causes a systematic shift in the extracted  $R$  value.

The IFG function as measured by the laser calibration system, Section 3.2.3, describes the measured energies of the individual crystal hits as a function of time in-fill. It is given by [58]

$$E = E_0(1 - C_A e^{-t/\tau_g}), \quad (5.38)$$

where  $E_0$  is the ‘true’ energy of the detected positron and  $E$  is the measured energy.

Constants  $C_A$  and  $\tau_g$  are measured with the laser calibration system, and then the function given in Equation 5.38 is used to convert the measured energies back to their real energies. In order to determine a systematic error from the applied IFG function, the IFG function was re-applied to the crystal hits with modified parameters. Specifically, the amplitude of the IFG function  $C_A$  and the lifetime  $\tau_g$  were scanned over in multiplicative steps, separately, such that all crystal energies included in a cluster hit were adjusted by a multiplicative factor before re-summing. The multipliers applied to the IFG parameters were scanned over from 0 to 2 in steps of 0.25. The results of the scans for the 60h dataset are shown in Figure 5.44, with a comparison to the results from fits with the T-Method. It is immediately apparent that the sensitivities of the T-Method to the IFG parameters are significantly greater than that of the Ratio Method. The Ratio Method's insensitivity to slowly varying effects which it divides out is one of its primary strengths. In fact, while there is an observable minimum in the T-Method  $\chi^2$  results for the amplitude multiplier scan, no such minimum exists for the Ratio Method results.

Tables 5.14 and 5.15 give the results for the IFG amplitude multiplier and lifetime multiplier scans respectively for the different datasets. Sensitivities for both the T-Method and Ratio Method fits are included. The uncertainties on the IFG parameters from fits to the laser data were in general of order 25% [104]. This uncertainty multiplied against the as-measured Ratio Method sensitivities results in systematic errors of <6 ppb on the amplitude multiplier and <17 ppb for the lifetime multiplier for the various datasets. The individual numbers are given in column 5 of Table 5.14 and column 3 of Table 5.15 respectively.

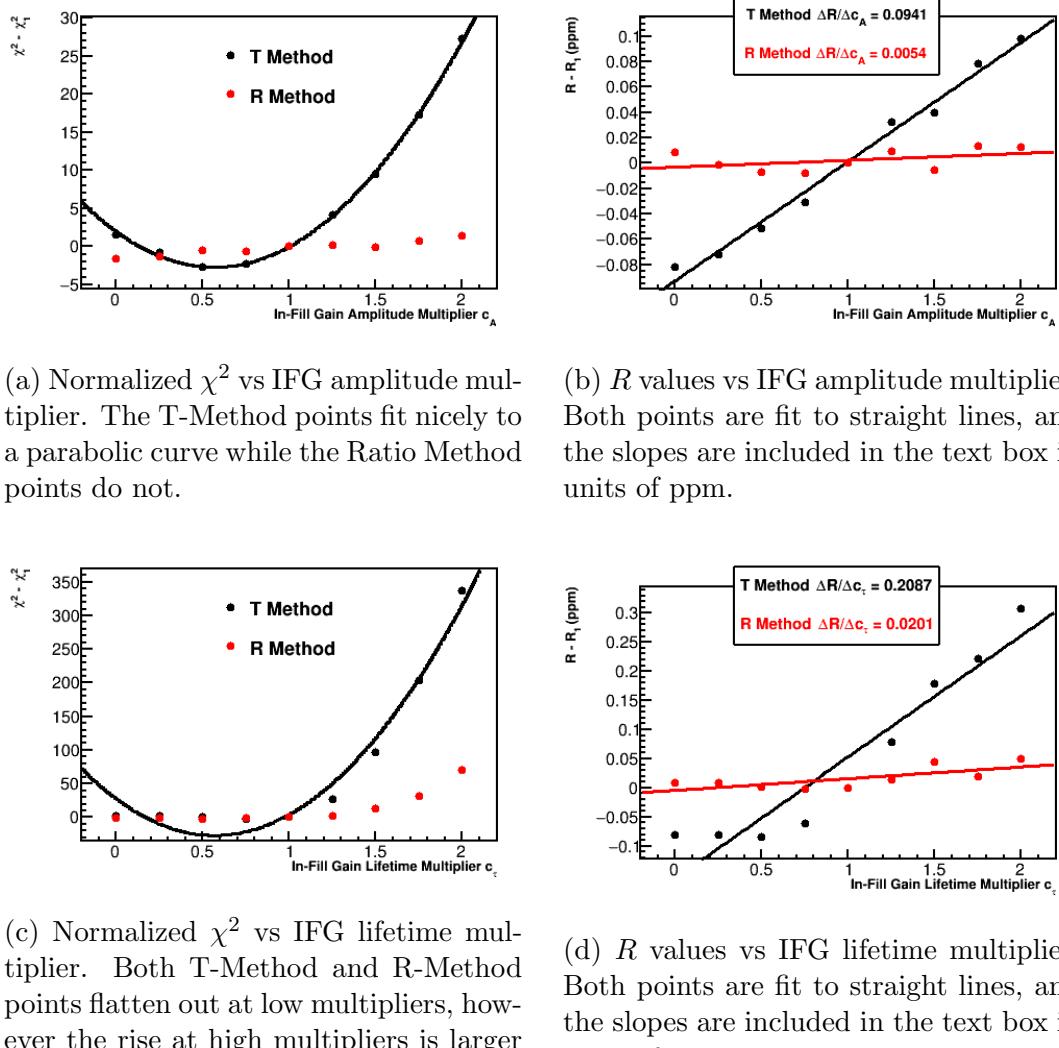


Figure 5.44: Fitted  $\chi^2$ 's and  $R$  values for the 60h dataset as a function of IFG amplitude multiplier (top) and IFG lifetime multiplier (bottom). Results with ratio fits (red) are compared to T-Method fits (black). Values are normalized to their  $C_A = 1$  and  $C_\tau = 1$  results respectively in order to put the curves on the same scale. As shown the sensitivities in the T-Method to the IFG parameters are significantly larger than that in the Ratio Method. This is one of the primary strengths of the Ratio Method.

Systematic Error due to IFG Amplitude							
Dataset	T-Method $dR/dC_A$	T-Method $C_{A_{\min}}$	R-Method $dR/dC_A$	R-Method $\delta R_{\sigma=0.25}$	R-Method $\Delta R_{0x}$	R-Method $\Delta R_{2x}$	R-Method $\delta R_{\max - 0x/2x}$
60h	94.1	0.57	5.4	<b>1.4</b>	8.2	12.4	—
HighKick	69.5	0.70	-23.6	<b>5.9</b>	28.6	-16.0	<b>28.6</b>
9d	64.9	0.27	1.4	<b>0.4</b>	-2.5	-3.9	—
Endgame	96.2	0.19	22.7	<b>5.7</b>	-6.9	43.6	<b>43.6</b>

Table 5.14: Sensitivities and systematic errors for the IFG amplitude. T-Method sensitivities are included for comparison, along with the  $\chi^2$  minima. Also included are changes in  $R$  for fits with IFG amplitude multipliers of 0x and 2x. Systematic error columns are in bold, where the one on the left corresponds to the Ratio Method sensitivity multiplied by a 25% uncertainty in the amplitude, and the one on the right corresponds to the absolute value of the maximum change in  $R$  with the 0x and 2x multipliers applied. Only the HighKick and Endgame values are used from the column on the far right. Units for errors and sensitivities are in ppb.

Systematic Error due to IFG Lifetime				
Dataset	T-Method $dR/dC_\tau$	R-Method $dR/dC_\tau$	R-Method $\delta R_{\sigma=0.25}$	R-Method $\Delta R_{0x}$
60h	208.7	20.1	<b>5.0</b>	8.2
HighKick	114.7	-44.8	<b>11.2</b>	28.6
9d	81.7	-46.4	<b>11.6</b>	-2.5
Endgame	207.3	66.1	<b>16.5</b>	-6.9

Table 5.15: Sensitivities and systematic errors for the IFG lifetime. T-Method sensitivities are included for comparison. The systematic error columns is in bold, where the error corresponds to the Ratio Method sensitivity multiplied by a 25% uncertainty in the lifetime. Units for errors and sensitivities are in ppb.

Examining Table 5.14 in more detail however, some interesting features can be seen. The T-Method results, while not the primary subject of this analysis, nevertheless vary between the datasets, in that the scan minima are not consistent with one another. For the Ratio Method results, the HighKick and Endgame datasets exhibit greater sensitivities to the IFG amplitude than the 60h and 9d datasets. The HighKick dataset can be seen to have a negative sensitivity compared to the positive sensitivities of the other datasets, and the T-Method sensitivities for all datasets. The Endgame dataset has a greater sensitivity stemming from the fact that fits to the multiplicative factors of  $C_A = 1.75$  and  $C_A = 2$  produced noticeably better  $\chi^2$ 's and higher  $R$  values,  $\mathcal{O}(40 \text{ ppb})$ . These fits pulled the slope up from what is otherwise a very flat fit to the rest of the points, and imply that the default IFG multiplier should be greater than 1.

These various peculiarities imply there are imperfections in the gain corrections which have not been fully accounted for. Indeed recent investigations into the applied gain corrections show that for the HighKick and Endgame datasets the out-of-fill corrections were improperly applied during data production [105]. While this does not introduce a systematic uncertainty as a function of time in-fill as described before,

it does shift the scale of the correction and by extension the extracted  $R$  value. In order to determine more conservative bounds on the systematic errors to account for these deficiencies, the absolute changes in  $R$  for the 0x and 2x multiplier fits were examined. These values are included in Table 5.14. The maximum of the two values was then taken as the systematic error on  $R$  due to the IFG amplitude. In general the systematic errors as calculated from the maximum of the 0x or 2x fits are an order of magnitude larger than those errors calculated directly from the sensitivities. Because the scale of these numbers is still small compared to the Run 1 statistical errors however, they are deemed acceptable for this analysis.

As for the lifetime multiplier scan results, it was observed that the  $R$  and  $\chi^2$  plots flattened out for small lifetimes as compared to large lifetimes, with no identifiable minima in the scan results.  $R$  was seen to change by very little when the lifetime was small, as shown in the far right column of Table 5.15. Because of the larger changes in  $R$  with the larger lifetimes, the sensitivities in the table were pulled to greater values than they otherwise might be. The choice was made to keep these large sensitivities in and use them in the calculation of the systematic error for a conservative approach, again justified by the gain issues mentioned previously.

In the calculation of the combined systematic error due to the IFG function parameters, the amplitude and lifetime parameter errors are added in quadrature, conservatively ignoring correlations between the two parameters in the fits to the laser data. Because the two parameters are correlated to a large degree, the 0x and 2x lifetime numbers were not used in the calculation of the HighKick and Endgame lifetime systematic errors as they have already been conservatively treated in the amplitude part of the error<sup>21</sup>. The confidence in the scale of the Ratio Method results is ultimately preserved from the fact that the Ratio Method fits are not as sensitive to the gain variations as the T-Method fits, as the gain effects divide out. Going forward

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<sup>21</sup>If the amplitude is zero then the lifetime is meaningless, and vice versa.

towards the end of the Run 1 analysis, once the outstanding gain issues have been resolved the systematic errors will be calculated directly from the sensitivities in all cases.

The STDP correction is similar in many to the IFG correction [57]. This correction is applied to pulses close in time,  $\mathcal{O}(\text{ns})$ , and is applied to the energies of the pulses before the in-fill gain correction. For the systematic error on  $R$  due to the application of the STDP gain correction, a new production of the 60h dataset was processed without the STDP correction applied. Because of how the time-randomization is applied to the clusters in this analysis, a new version of the same dataset by default uses a different default randomization to the cluster times and the ratio histogram filling. As shown in Section 5.4.5, the width in the fitted  $R$  values for many random seeds is of order  $\mathcal{O}(100 \text{ ppb})$  making direct ratio fit comparisons between the two datasets un-informative. In order to avoid this difficulty, fits were done with the T-Method, and cluster times were randomized per-fill rather than per-cluster. Since the per-fill randomization uses fill ID's as the seeds for the randomization, and because the T-Method does not split the data into sub-datasets as the Ratio Method does, the randomization between the 60h dataset with and without the STDP correction is identical. The change in  $R$  for T-Method fits with and without this correction is then taken as the upper bound on the systematic error for the inclusion of the STDP in the Ratio Method fits. This is reasonable as the previous IFG systematic studies showed a reduction in sensitivity to gain effects which is reasonably extended to the STDP correction error. Table 5.16 gives the T-Method fit results with and without the STDP, along with the change in  $R$ . This difference was found to be 11.0 ppb, which is then taken as the systematic error for all datasets.

Table 5.17 gives the quadrature sum for the total gain systematic errors for the Run 1 precession frequency analysis datasets. As shown for the 60h and 9d datasets

<b>Systematic Error due to STDP</b>			
Fit Type	$R$ with STDP (ppm)	$R$ without STDP (ppm)	$\delta R$ (ppb)
T-Method	−20.1619	−20.1729	<b>11.0</b>

Table 5.16: T-Method fit results with and without the STDP gain correction on the 60h dataset. T-Method fits were done instead of Ratio Method fits in order to force the cluster-time randomization to be consistent between the two dataset productions. The change in  $R$  in the bold column is taken as the upper bound on the systematic error in the Ratio Method due to the STDP gain correction.

<b>Total Gain Systematic Errors</b>				
Type of Error	60h	HighKick	9d	Endgame
IFG amplitude	1.4	28.6	0.4	43.6
IFG lifetime	5.0	11.2	11.6	16.5
STDP	11.0	11.0	11.0	11.0
Quadrature sum	12.2	32.6	16.0	47.9

Table 5.17: Total gain-related systematic errors for the Run 1 precession frequency analysis datasets. Units are in ppb.

the total error is below the target final error of 20 ppb. For the HighKick and Endgame datasets, the errors are slightly larger than 20 ppb, and should reduce once their known issues have been resolved. All gain errors will be re-evaluated once the final datasets are available. It is expected that the errors given here are reasonable approximations of the final errors, especially due to the Ratio Method’s insensitivity to the gain relative to the T-Method. As a reminder, the errors given here, even with the crude approximations, are small compared to the statistical errors of the various datasets.

### 5.5.3 CBO systematic errors

If the CBO is mis-modeled then there will be a systematic error on  $R$  since there is a an early-to-late change in both the frequency and the scale of the CBO. The

<b>Change in <math>R</math> with station 18 CBO parameters</b>	
Dataset	$\delta R$
60h	7.5
HighKick	0.4
9d	-2.0
Endgame	8.0

Table 5.18: Changes in the fitted  $R$  values with tracker station 18 CBO frequency model parameters instead of tracker station 12. The systematic errors are conservatively taken as the absolute value in the changes in  $R$ . Units are in ppb.

CBO model is well constrained by tracker measurements, but systematic errors can be evaluated by modifying the fixed frequency function given in Equation 5.22, and the decoherence envelope of the CBO.

Table 5.2 gives the CBO frequency model parameters for both tracker stations. The station 12 values are by default used in all fits to the data. Fits were performed with the station 18 values, and the changes in  $R$  are given in Table 5.18. While the CBO parameters in the tracking analysis fits do have errors on the parameters, they are tiny compared to the systematic errors between the two tracker stations [96]. Some fits were made by varying the fixed frequency parameters by  $1\sigma$  in their individual respective errors, and the changes in  $R$  were found to be negligible. For this reason the absolute values of the changes from station 12 to station 18 for the different datasets are conservatively taken as the systematic error on  $R$  due to the choice of fixed CBO frequency model parameters.

The shape of the CBO, or the decoherence envelope, is also similarly constrained by the tracking analysis. The envelope by default is taken to be an exponential as given in Equation 5.19 and shown in Figure 4.29. The other envelope which could reasonably exist in the data upon inspection in the tracking analysis is an exponential

plus a constant

$$e^{-t/\tau_{cbo}} \rightarrow e^{-t/\tau_{cbo}} + C, \quad (5.39)$$

where  $C$  is some constant CBO amplitude which persists over the course of the fill. In order to assess this systematic error, this new envelope was introduced into the  $N_{cbo}(t)$  fit term. Fits were done with  $C$  floating, where the starting values for  $C$  in the Ratio Method fits were taken from T-Method fits to the data<sup>22</sup>. While in T-Method fits the  $C$  parameter converged to values with errors about half the value, in general in the Ratio Method fits the  $C$  parameters had relatively large errors. In spite of the large errors however, the fits converged properly with the floating  $C$  parameter. In general the final fit parameters are largely the same, with the exception being the fitted CBO lifetime and amplitude on the  $N_{cbo}(t)$  term which both about halved. This is unsurprising as the lifetime is highly correlated to the amplitude. Only in the 9d dataset did some small complications arise, where the fit didn't particularly handle the floating  $C$  parameter well. Fixing the lifetime to a value of 99  $\mu$ s, chosen such that the lifetime reduction was somewhat comparable to that seen in the HighKick dataset which had the same  $n$  value, the  $C$  parameter converged to a value close to and consistent with zero.

The fitted values for the  $C$  constants and the changes in the final fitted  $R$  values are given in Table 5.19. As shown the changes in  $R$  are of order 10s of ppb for some of the datasets, with  $R$  varying both positively and negatively. These changes in  $R$  are conservatively taken as the systematic errors on  $R$  for the different datasets.

Table 5.20 gives the quadrature sum for the total CBO systematic errors for the Run 1 precession frequency analysis datasets. As shown for each dataset the total error is below the target final error of 30 ppb. The changing frequency of the CBO

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<sup>22</sup>T-Method starting values for  $C$  were taken to be 0 which resulted in well-converging fits.

<b>Systematic Error due to CBO Envelope</b>			
Dataset	$C \times 10^{-4}$	$\sigma_C \times 10^{-4}$	$\delta R$
60h	10.7	8.3	17.6
HighKick	11.6	10.4	-18.0
9d	-2.1	11.1	9.3
Endgame	10.7	4.0	-4.3

Table 5.19: Systematic error on  $R$  due to the choice of CBO envelope. The fitted floating parameter  $C$  and its error are given along with the change in  $R$  compared to the standard exponential envelope in units of ppb.

<b>Total CBO Systematic Errors</b>				
Type of Error	60h	HighKick	9d	Endgame
CBO Frequency Model	7.5	0.4	2.0	8.0
CBO Decoherence Envelope	17.6	18.0	9.3	4.3
Quadrature sum	19.1	18.0	9.5	9.1

Table 5.20: Total CBO-related systematic errors for the Run 1 precession frequency analysis datasets.

in Run 1 was eliminated in Run 2, implying a reduction in any associated systematic errors. The increased statistics combined with the tracking analysis should constrain measurements of the CBO decoherence envelope even further, similarly reducing any associated systematic errors. On the other side of the coin, increased statistics might bring out higher order CBO effects which require improved modeling.

#### 5.5.4 Lost muon systematic errors

There are two types of systematic errors arising from lost muons. The first is due to mis-construction of the shape of the lost muon spectrum  $L(t)$  included in the fits and the second is due to any  $g - 2$  phase difference between the stored and lost muon populations.

As mentioned in Section 5.3.3, the triples spectrum is made with cuts as defined

<b><math>\Delta R</math> with Various Lost Muon Cuts</b>				
Type of fit or cut	9d Dataset		Endgame Dataset	
	$\Delta R$ (ppb)	$\kappa_{loss}$	$\Delta R$ (ppb)	$\kappa_{loss}$
Default cuts	0	2.510	0	2.345
No quadruple subtraction	0.2	1.811	-0.1	1.717
No accidental subtraction	0.1	2.503	< 0.1	2.339
$\Delta t_{13} \leq 12.5$ ns	0.1	4.469	-0.3	4.248
$5$ ns $\leq \Delta t_{12,23} \leq 8.5$ ns	0.1	2.507	< 0.1	1.709
$100$ MeV $\leq E_{1,2,3} \leq 500$ MeV	< 0.1	2.253	< 0.1	2.090

Table 5.21: Effect on the fitted  $R$  value in the 9d and Endgame datasets with various cuts used or backgrounds subtracted. Also included are the various corresponding  $\kappa_{loss}$  values which compensate for the level of statistics contained within  $L(t)$  due to the differing cuts. Ultimately how the muon loss spectrum  $L(t)$  is created has little bearing on the final fitted  $R$  value.

in Table 5.4. Various backgrounds are subtracted off the triples spectrum in order to generate a clean sample of lost muons. Table 5.21 gives the change in  $R$  for various sets of cuts and background subtractions for the 9d and Endgame datasets. As shown the various different backgrounds and cuts ultimately make very little difference in the final fitted  $R$  value. Similarly, stable beam contaminants in the form of deuterons and protons contaminate the lost muon spectrum. The former are largely removed by straightforward  $\Delta t_{12}$  and  $\Delta t_{13}$  cuts. The latter can be mostly removed by cutting on the negative side of the  $\Delta t_{13}$  distribution with  $\Delta t_{13} \leq 12.5$  ns which separates the populations more readily, Figure 5.45. While this does largely remove the protons at the cost of statistics, the fitted  $\kappa_{loss}$  parameter simply grows larger to compensate. Ultimately the effect on  $R$  is negligible, with  $\Delta R = -0.3$  ppb. The sum of these separate types of errors is conservatively taken at 0.5 ppb for all datasets<sup>23</sup>.

Because  $\kappa_{loss}$  is fixed in the Ratio Method fits from corresponding T-Method fits,

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<sup>23</sup>For the T-Method fits, though the changes in  $R$  are noticeably larger with the various cuts, they are still the same order of magnitude and the error is conservatively below 1 ppb.

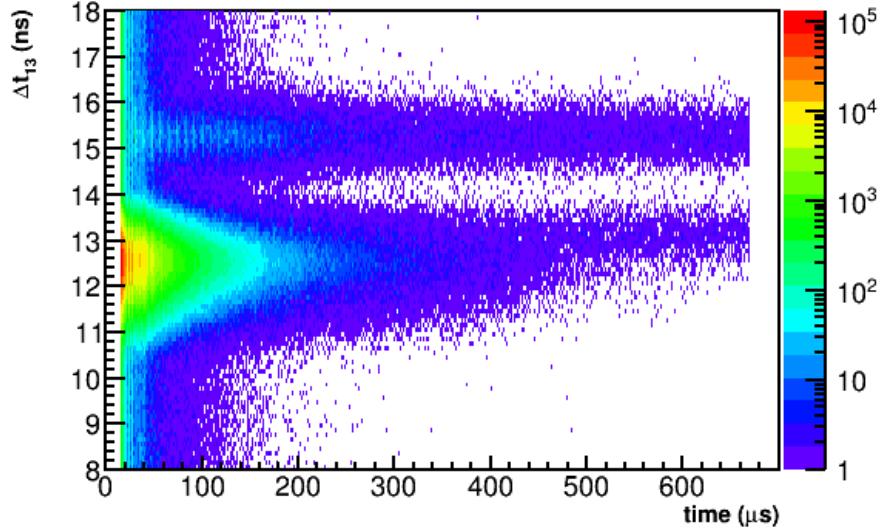


Figure 5.45: Lost muon  $\Delta t_{13}$  distribution as a function of time in-fill. The band of hits centered above 15 ns corresponds to the deuteron contamination, while the band of hits centered above 13 ns corresponds to the proton contamination.

the statistical correlation to  $R$  is neglected and a systematic error must be estimated. The systematic error can be determined by scanning over the value of  $\kappa_{loss}$  as in Figure 5.46. The error on  $\kappa_{loss}$  is taken as the statistical error determined in the T-Method fit to the same data. Table 5.22 gives the systematic errors on  $R$  for the Run 1 precession frequency analysis datasets. As shown the systematic errors as calculated from the sensitivities are all small,  $\delta R < 5$  ppb.

The most dangerous systematic error from lost muons is due to the  $g - 2$  phase of the lost muon population. If muon losses originate from a population of muons with a different average phase than the stored muons, then there will be a phase shift over the course of a fill. This implies the extracted  $\omega_a$  frequency will be systematically pulled. Simulation and data have shown a correlation between  $g - 2$  phase and muon momentum at injection [106, 107]. If the losses are momentum dependent and the phase-momentum correlation is preserved after injection, then there will be a systematic error. Losses are expected to be momentum dependent [108]; the

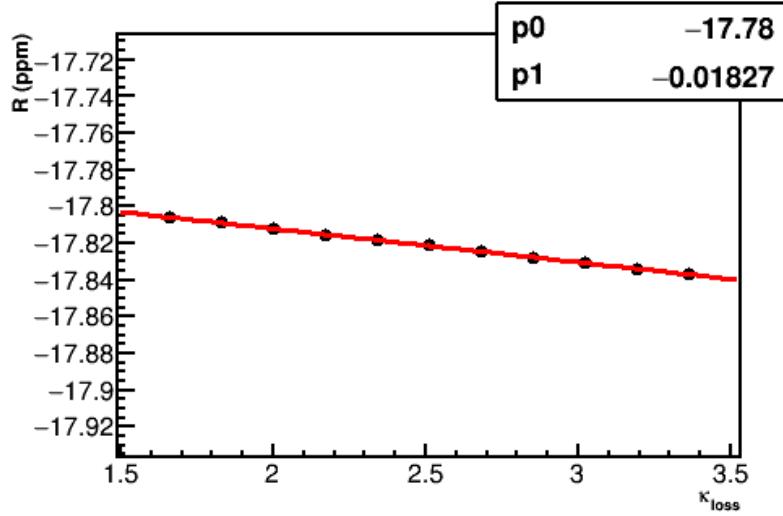


Figure 5.46: The sensitivity of  $R$  to the fixed  $\kappa_{loss}$  parameter. Error bars have been removed from the plot in order to show the trend more clearly. Units are in ppm, data are from the 9d dataset.

Systematic Error due to Fixed $\kappa_{loss}$				
Dataset	$dR/d\kappa_{loss}$	$\sigma_{\kappa_{loss}}$	$\delta R$	$\Delta R_{(w/ - w/o)}$
60h	-3.5	0.338	<b>1.2</b>	-31.4
HighKick	-7.1	0.697	<b>4.9</b>	-40.1
9d	-18.3	0.170	<b>3.1</b>	-45.7
Endgame	-2.6	0.038	<b>0.1</b>	-6.1

Table 5.22: Systematic error due to the fixed  $\kappa_{loss}$  parameter in the Ratio Method fits for the Run 1 precession frequency analysis datasets. All units are in ppb except for the  $\sigma_{\kappa_{loss}}$  parameter which is unit-less.  $\sigma_{\kappa_{loss}}$  comes from the T-Method fit results and scales with the number of statistics. The bold column gives the systematic errors on  $R$ . The far right column gives the change in  $R$  with the  $\kappa_{loss}$  parameter in the fits versus without it entirely.

low kick applied in Run 1 resulted in an off-center stored muon beam so there are more high momentum muons at greater radii near the edge of the storage aperture, compared to those muons on the inside of the storage region. The phase-momentum correlation originates from the experience of muons within the accelerator beam-line before injection. One simple effect is that higher momentum muons take longer path lengths inside the accelerator's steering magnetic fields, during which their spins will precess and their  $g - 2$  phases change. Simulation efforts are underway, along with data-driven studies, to understand and quantify the phase-momentum correlation and momentum-dependent loss probabilities [108, 109]. Here is given an estimation of the muon loss phase bias systematic error using the measured losses for the different Run 1 datasets.

From Equation 5.33, the systematic shift in  $\omega_a$  can be written as

$$\frac{\Delta\omega_a}{\omega_a} = \frac{1}{\omega_a} \frac{d\langle\phi\rangle}{dt}, \quad (5.40)$$

where  $d\langle\phi\rangle/dt$  is the change in average phase over time. With some assumptions made with regards to ‘core’ and ‘edge’ muon populations, where lost muons can only originate from the latter, this can be written as [108]

$$\frac{d\langle\phi\rangle}{dt} = f_c \cdot f_l \cdot \Delta\phi_{c-e}, \quad (5.41)$$

where  $f_c$  is the fraction of muons in the core population to all muons,  $f_l$  is the fractional loss rate, and  $\Delta\phi_{c-e}$  is the phase difference between the core and edge muon populations. The simulation studies in the afore-referenced document gives estimates of  $f_c \approx 0.9$  and  $\Delta\phi_{c-e} \approx 1$  mrad. The fractional losses (the integral in Equation 5.29 times the final fitted  $\kappa_{loss}$  parameter, assuming no losses are included in  $L(t)$  before the fit start time) for the Run 1 precession frequency analysis datasets as determined in this analysis are shown in Figure 5.47. As shown the losses rise

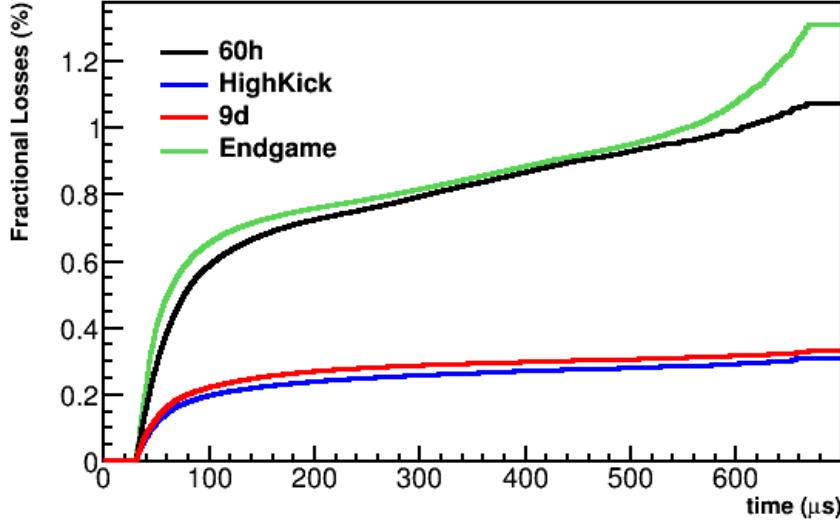


Figure 5.47: Fractional losses for the Run 1 precession frequency analysis datasets. The curves begin at  $30.2\text{ }\mu\text{s}$  which is where the fit begins. A value of 1% at a specific time  $t$  indicates that there are 1% fewer stored muons at that time than there would be if there were no losses at all. The Endgame and 60h datasets can be seen to have the most losses, while the 9d and HighKick have less. This is due to the higher kicks in the latter datasets which put the stored muon beam on a more central orbit. The upward tail at the end of the Endgame dataset corresponds to the remnant proton contamination.

sharply at early times before leveling off to a constant rate around  $100\text{ }\mu\text{s}$  into the fill. Taking the first  $70\text{ }\mu\text{s}$  of the fit and assuming that this loss rate is constant over the course of the fill, the approximate loss rates for the different datasets can be determined<sup>24</sup>. These are given in Table 5.23. Applying the various quantities into Equation 5.41, along with  $\omega_a \approx 1.44\text{ rad}/\mu\text{s}$ , the systematic shifts in the precession frequency are found. They are included in Table 5.23.

Taking the shifts in  $\omega_a$  as the systematic errors, the approximations here are on the order of 50 – 60 ppb for the 60h and Endgame datasets, and on the order

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<sup>24</sup>While the loss rate reduces after  $100\text{ }\mu\text{s}$ , losses occur mostly at early times, so this is a fine approximation.

<b>Systematic Error due to Lost Muon Phase Bias</b>		
Dataset	$f_l$ (%/70 $\mu$ s)	$\Delta\omega_a/\omega_a$ (ppb)
60h	0.60	-53.6
HighKick	0.20	-17.9
9d	0.25	-22.3
Endgame	0.65	-58.0

Table 5.23: Loss rates and associated shift in  $\omega_a$  for the Run 1 precession frequency analysis datasets. Loss rates are determined by inspection of the respective curves for the different datasets, and are approximate. The systematic shifts here are negative, due to the fact that the average phase of the stored muons is greater than that of lost muons, as determined in simulation.

of 20 ppb for the HighKick and 9d datasets. These estimates are the same order of magnitude as determined in other evaluations with slightly different parameters [107]. As simulations improve and more studies are undertaken, the uncertainties here may be replaced in favor of actual corrections, in which case the systematic errors would then be the uncertainties in the evaluation of those corrections. While greater than the target goal of 20 ppb, the systematic errors listed here are nevertheless small compared to the Run 1 statistical errors, and expected to decrease in future runs.

### 5.5.5 Ratio construction systematic errors

In the construction of the ratio data, when filling the four sub-datasets as in Equation 5.11, the parameters  $T_a$  and  $\tau_\mu$  for the  $g - 2$  period and muon lifetime need to be known a priori. If these parameters are incorrectly chosen, then it is possible there will be a systematic shift on  $R$ . This is especially important when considering  $T_a$ , because the quantity which the E989 experiment is measuring must be used in the analysis, creating a sort of self-dependence. The question then naturally arises as to how well the  $T_a$  parameter needs to be known. As described in Section 5.2.2, the input value for  $T_a$  is nominally taken as the result from the E821 experiment.

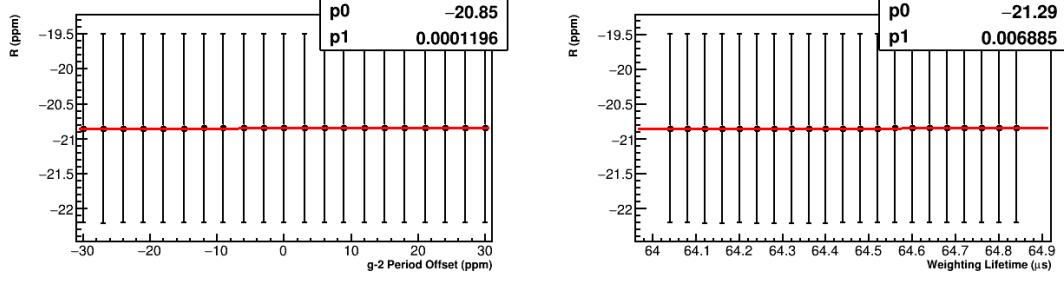
In order to determine systematic errors from the choices of  $T_a$  and  $\tau_\mu$ , they were scanned over when forming the ratio data. The input value for  $T_a$  was varied from  $-30$  ppm to  $30$  ppm around the default value in steps of  $3$  ppm. The input value for  $\tau_\mu$  was varied from  $64.04\ \mu\text{s}$  to  $64.84\ \mu\text{s}$  in steps of  $0.04\ \mu\text{s}$ . See Figure 5.48 for scan results for the  $60\text{h}$  dataset. See Table 5.24 for the sensitivities determined from the scans for all datasets. As shown the sensitivities vary both positively and negatively for the different datasets, and are extremely small. The positive and negative variations imply there is no real systematic effect at play, and that as long as a reasonable choice for these two parameters is made, then the ratio data is very insensitive to the exact values chosen.

In order to be conservative however, a scale for the changes in  $R$  is given. Since the measured  $g - 2$  period in the data is already modified by the hardware blinding, it is technically the hardware shifted  $g - 2$  period that we want to use. The calorimeter digitizers use a  $40 + \delta$  MHz clock which has been blinded to a value in the range of  $39.997$  to  $39.999$  MHz[94]. This corresponds to a  $75$  ppm range in the frequency, in a uniform distribution. Calculating the uncertainty from the uniform distribution and adding it in quadrature with a conservative  $1$  ppm uncertainty in the guess on the true  $g - 2$  period from the E821 result,

$$\delta T_a = \sqrt{(75)^2/12 + 1^2} = 21.7 \text{ ppm.} \quad (5.42)$$

This results in a change in  $R$  on the order of  $2.2$  ppb for the  $60\text{h}$  and HighKick datasets, and less for the  $9\text{d}$  and Endgame datasets.

Similarly, the sensitivities of  $R$  to the chosen muon lifetime are very small, order ppb/ $\mu\text{s}$ . Since the uncertainties in the muon lifetime are of order ns from fits to the data, any systematic errors from this parameter would be completely negligible even if the sensitivities all had the same sign.



(a)  $R$  versus input value for  $T_a$ , where the x-axis is given in units of a ppm level shift of the default choice for  $T_a$ .

(b)  $R$  versus the input value for  $\tau_\mu$ . Parameter  $p_1$  gives the sensitivity in ppm/ $\mu\text{s}$ .

Figure 5.48: Scans over ratio construction parameters for the 60h dataset. In general the points are randomly spread around, and sensitivities are practically negligible.

Sensitivity to Ratio Construction Parameters		
Dataset	$dR/dT_a$	$dR/d\tau_\mu$
60h	0.1	6.9
HighKick	-0.1	-4.1
9d	< 0.1	-1.1
Endgame	< 0.1	0.6

Table 5.24: Sensitivities of  $R$  to ratio construction parameters.  $dR/dT_a$  is in units of ppb/ppm, while  $dR/d\tau_\mu$  is in units of ppb/ $\mu\text{s}$ . In both cases the sensitivities are both extremely small, and vary negatively and positively for the different datasets.

<b>Sensitivity to Binning Parameters</b>		
Dataset	$dR/d_{\text{bin width}}$	$dR/d_{\text{bin edge}}$
60h	24.5	-0.1
HighKick	6.0	-0.7
9d	-22.8	-0.3
Endgame	-41.7	-0.6

Table 5.25: Sensitivities of  $R$  to binning parameters. Units are in ppb/ns. While some of these values may appear significant, inspection of the actual plots reveals that the actual trends are not quite so convincing.

### 5.5.6 Binning systematic errors

When constructing the time spectra, bin widths and the starting edge of the bin are by default chosen to be 149.2 ns and 0 ns respectively. In order to calculate systematic errors from the choices of these parameters, the values were scanned over. The bin width was scanned from 148.7 ns to 149.7 ns in steps of 0.1 ns, while the bin edge was scanned from 0 ns to 149.2 ns in steps of 14.92 ns. Figure 5.49 shows scan results for the 9d dataset. While trends appeared largely random, a straight line was fit to the scan results in order determine the sensitivities of  $R$  to both parameters. See Table 5.25. For the bin edge scan, it was verified that a shift of one bin width returned the same fit results as the default shift of 0 ns. This combined with the negligible sensitivities and varying points implies negligible systematic effects on  $R$  from the choice of bin edge. For the systematic error due to the choice of bin width, it should be noted that in general the choice of bin width should be optimized to be equal to the peak of the cyclotron period distribution of the stored muons, which from the fast rotation analysis informed the choice of 149.2 ns. This uncertainty from the fast rotation analysis would certainly be less than 0.1 ns. Multiplied by this uncertainty by the dataset sensitivities gives uncertainties of 2.5, 0.6, 2.3, and 4.2 ppb for the 60h, HighKick, 9d, and Endgame datasets respectively.

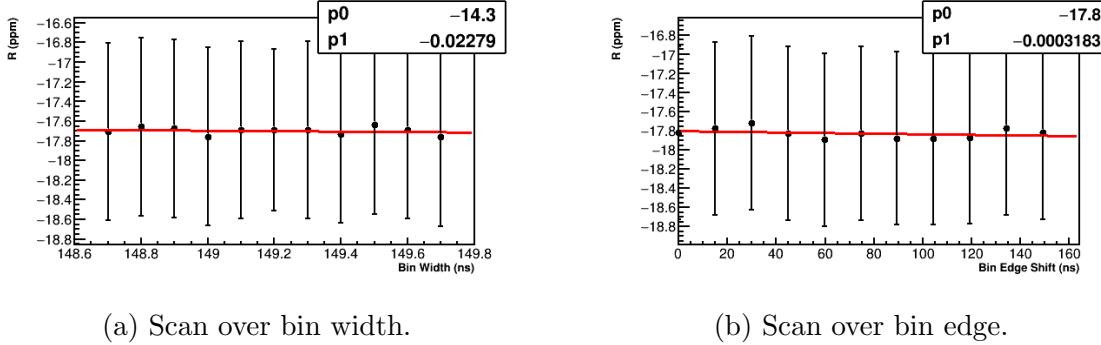


Figure 5.49: Scans over binning parameters for the 9d dataset. In general the points are randomly spread around, indicating no real systematic effect.

### 5.5.7 Systematic error from differential decay

Muons are injected into the storage ring with a range of momenta. Muons with larger momenta will in general live longer than muons with smaller momenta due to their time-dilated lifetimes. Muons at smaller radii therefore decay more often, leading to an increasing beam radial position and average momentum over the course of a fill. Decay positrons at greater radii with larger momenta in general take longer to reach calorimeters than those at small radii with smaller momenta. In the precession frequency measurement, the hit times of the detected positrons naturally correspond to offset times from when the muons decayed equal to the drift times. Because the average drift time changes over the course of a fill due to the average momentum distribution increase, there is thus a changing phase over the course of a fill. To add to that, as discussed in Section 5.5.4, there is a phase-momentum correlation for muons at injection. This only compounds upon the changing phase even further. The systematic error due to this effect is called “differential decay.”

Underway beam-line simulations are needed to fully measure the phase-momentum correlation and fully estimate this systematic error. A calculation for the previous E821 experiment resulted in a shift on  $\omega_a$  of  $\langle \omega_a \rangle / \omega_a \approx -36.8$  ppb [110], which should

be comparable to any calculation for E989. The level of the fully evaluated systematic shift may not be small compared to the final target systematic error goal, but for Run 1 the scale of this estimate is deemed to be acceptable. For the purposes of this analysis, an upper bound of 40 ppb is taken as the systematic error. Because the differential decay effect only increases the average stored momentum of the beam and can't decrease it, a correction can be applied if any final uncertainty is too large.

### 5.5.8 Systematic errors in the E-field and pitch corrections

The electric field and pitch corrections modify the final value extracted for  $\omega_a$  as described in Section 2.7. Any errors in the estimation of these corrections is by extension an error in the precession frequency measurement. The evaluation of these errors is performed by separate working groups. Given here are preliminary estimates of the errors.

The pitch correction is dependent on the vertical width of the muon beam and is evaluated by the tracking analysis. A preliminary analysis of the 60h dataset calculated an estimated correction and systematic error of [111, 112]

$$C_P \sim -160 \pm 15 \text{ ppb.} \quad (5.43)$$

This error is split into track reconstruction errors of which there are many ( $\mathcal{O}(10 \text{ ppb})$ ), a tracker-calorimeter acceptance correction with a small expected error, and uncertainty in simulation models used to extract the pitch correction from the vertical width measurements. The variation in the correction between datasets is expected to be small and the error in the correction is expected to be relatively consistent. The pitch correction error given in Equation 5.43 is taken as the systematic error for all datasets.

The E-field correction is evaluated with the fast rotation analysis and is dependent

<b>Electric field corrections</b>			
Dataset	Correction	Statistical error	Systematic error
60h	-519	1.0	27.0
HighKick	N/A	N/A	N/A
9d	-463	1.0	36.0
Endgame	-467	1.0	20.0

Table 5.26: Electric field corrections estimated by the Cornell fast rotation analysis group [113, 114, 115]. Units are in ppb.

on the momentum distribution of the stored muon beam. Analyses of the 60h, 9d, and Endgame datasets by an independent analysis group yielded the results shown in Table 5.26. While not the final numbers for Run 1, the estimates and errors given will change only slightly from further DQC cuts. For the HighKick dataset which has not been evaluated at the time of writing, a systematic error on the E-field correction is taken as the largest error from the other datasets, that being 36 ppb for the 9d dataset.

The above corrections make certain assumptions with regards to the conditions which influence the stored beam dynamics. Further errors can be attributable to non-linearities in the electrostatic quadrupole field, misalignment of the quadrupole plates, the changing voltage on the quadrupole plates due to the bad resistors, and any residual radial magnetic field. Preliminary analyses using simulation have estimated additional systematic errors on the pitch and E-field corrections of  $\sim 5$  ppb and  $\sim 20$  ppb respectively [116, 117]. The error for the pitch correction has already been included in the number given in Equation 5.43, and the error for the E-field correction is conservatively added in quadrature with the errors given in Table 5.26. Simulation efforts are continuing in order to assess these errors more accurately. While the errors listed here are greater than the target quadrature sum of 30 ppb, they are nonetheless small compared to the statistical error on  $R$  and are deemed acceptable for Run 1.

### 5.5.9 Systematic error due to vertical beam motion

As described in Section 4.4, due to the damaged resistors present in Run 1 the vertical distribution of the stored muon beam changed over the course of a fill. Because the  $g - 2$  phase is different for each vertical position due to drift length, this leads to a changing average phase and thus a systematic error. Preliminary studies fitting the average  $g - 2$  phase with the tracker data have calculated systematic shifts in  $\omega_a$  for the 60h and Endgame datasets of 115 and 180 ppb respectively [118]. Since these are one-directional shifts in  $\omega_a$ , similar to the differential decay effect, they can be applied as corrections to the final extracted precession frequency. As systematic studies mature and the shifts are better quantified, this is the intended action. For now these numbers are taken as limits on the systematic errors for their respective datasets, with the smaller number from the 60h dataset being applied to the HighKick and 9d datasets as well. The 60h result was chosen instead of the Endgame result because the Endgame dataset saw noticeably larger drift due to further degradation of the quadrupole resistors.

### 5.5.10 Systematic error summary

Table 5.27 gives all evaluated systematic errors in this analysis as described in the preceding sections. The final total quadrature sums of the systematic errors including the various preliminary and conservative estimates for some errors lie within the range 140 – 200 ppb. These are in comparison to the final target goal of 70 ppb for the precession frequency measurement. Once the gain issues are resolved and the other working groups improve their systematic uncertainty evaluations, the systematic errors for Run 1 and future runs of E989 will almost certainly reach the target goal. Indeed just removing the stored muon beam systematic error results in systematic errors in the range of 70 – 100 ppb, much closer to the target of 70 ppb. As the

systematic errors stand currently in this analysis, they are all small compared to the Run 1 statistical errors of each respective dataset, making the presented analysis statistics-limited.

<b>Run 1 Precession Frequency Systematic Errors</b>				
Error	60h	HighKick	9d	Endgame
Pileup amplitude	22.2	19.0	9.0	9.4
Pileup phase - time-shift	17.6	19.0	17.1	14.3
Pileup phase - energy-scale	19.4	3.7	5.5	5.3
In-fill gain amplitude	1.4	28.6	0.4	43.6
In-fill gain lifetime	5.0	11.2	11.6	16.5
STDP On/Off	~ 11.0	~ 11.0	~ 11.0	~ 11.0
CBO frequency model	7.5	0.4	2.0	8.0
CBO decoherence envelope	17.6	18.0	9.3	4.3
Lost muon cuts	< 0.5	< 0.5	< 0.5	< 0.5
Fixed $\kappa_{loss}$	1.2	4.9	3.1	0.1
Ratio construction $T_a$	2.2	2.2	< 2.2	< 2.2
Ratio construction $T_\mu$	< 0.1	< 0.1	< 0.1	< 0.1
Bin width	2.5	0.6	2.3	4.2
Quadrature Sum	41.3	46.4	27.8	52.2
Lost muon phase bias	53.6	17.9	22.3	58.0
Differential decay	< 40	< 40	< 40	< 40
Pitch correction	15	~ 15	~ 15	~ 15
E-field correction	33.6	~ 41.2	41.2	28.3
Stored beam motion	~ 115	~ 115	~ 115	~ 180
Quadrature Sum	138.0	130.6	131.3	195.9
Quadrature Sum *	76.3	62.0	63.4	77.4
Total Quadrature Sum	144.1	138.6	134.2	202.8
Total Quadrature Sum *	86.8	77.4	69.2	93.3

Table 5.27: Final systematic errors evaluated in the Run 1 precession frequency analysis to the 60h, HighKick, 9d, and Endgame datasets. All units are in ppb. The table is split into two sections. The upper section consists of systematic errors directly evaluated by the author while the lower section consists of preliminary systematic estimates by other working groups. The E-field and pitch correction errors have been added in quadrature with the quadrupole errors. The publication errors for the Run 1 datasets will change from these as the final DQC cuts are made and analyses improved, however the scale of these errors will remain consistent. \* Quadrature sum errors calculated excluding the stored beam motion error for comparison.

## 5.6 Final precession frequency results

Four datasets from Run 1 of E989 have been analyzed for this dissertation, those being the 60h, HighKick, 9d, and Endgame. In each case the datasets are the near-final datasets for Run 1. The precession frequency analysis was done using the Ratio Method, an analysis technique for fitting the decay positron time spectra which divides out the exponential decay and slow terms in the data. The final results for the blinded frequency  $R$  values for the different datasets along with their total statistical and systematic errors are given in Table 5.28. Figure 5.50 compares the means. The total error for the four datasets is 491.2 ppb conservatively assuming completely correlated systematic errors. The analysis is statistics limited, even with the conservative preliminary estimates for certain systematic errors as given in Table 5.27. Before combining the results, the  $R$  values given here will have their individual E-field and pitch corrections applied, and their blinding normalized. Since the magnetic field changes by a fair amount between datasets, the  $R$  values will be converted back into the precession frequency  $\omega_a$  using Equation 5.14 before combination with the magnetic field measurements to give per-dataset values of  $a_\mu$ . The dataset  $a_\mu$  values will then be combined in a weighted average.

Run 1 Precession Frequency Results				
Dataset	$R$	$\sigma_{\text{stat.}}$	$\sigma_{\text{sys.}}$	$\sigma_{\text{tot.}}$
60h	-20.5562	1.3581	0.1440	1.3657
HighKick	-17.4755	1.4112	0.1382	1.4180
9d	-17.7182	0.9033	0.1337	0.9132
Endgame	-17.3406	0.6393	0.2021	0.6707
Total (systematics 0% correlated)	0.4605			0.4737
Total (systematics 100% correlated)	0.4605			0.4912

Table 5.28: Run 1 final results for the precession frequency analysis datasets. The  $R$  values given here are mean values of fits to 50 different random seeds. The 60h dataset has a different software blinding than the rest, shown by the different mean  $R$  value. Statistical and systematic errors are included alongside the total error for each dataset. In each dataset case the error is statistics dominated. Total combined errors for the different datasets are shown for cases where the systematic errors are assumed to be 0% and 100% correlated. Units are in ppm.

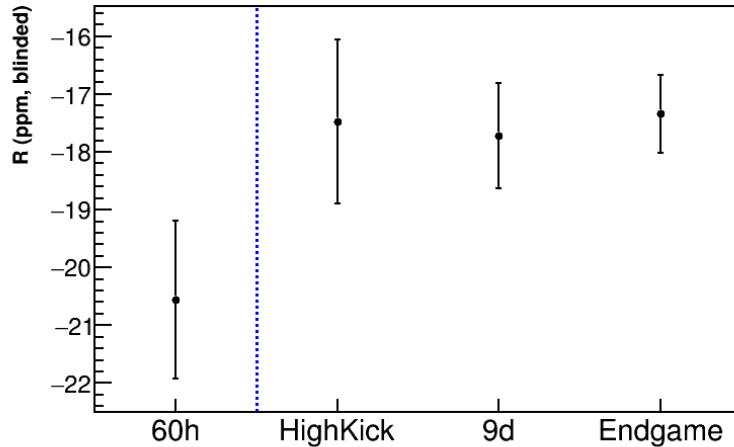


Figure 5.50: Comparison between dataset mean  $R$  values from fits to 50 different random seeds. The 60h dataset has a different software blinding than the rest.

## Chapter 6

### Conclusion

Experiment E821 measured the anomalous magnetic moment of the muon to a relative uncertainty of 540 parts per billion, corresponding to a three to four standard deviation from the Standard Model theoretical value. In order to confirm or deny this discrepancy, a new experiment has been undertaken to measure the same quantity.

E989 measures  $a_\mu$  by measuring the spin precession frequency of muons within a magnetic storage ring and the strength of the magnetic field within the storage region. The former is done by counting the number of decay positrons observed in electromagnetic calorimeters above an energy threshold, while the latter is done using NMR probes in and around the muon storage region. Straw trackers assist both measurements by measuring muon beam dynamics which directly impact both the precession frequency measurement and the distribution of muons within the measured magnetic field. This dissertation has presented the track fitting algorithm used in the reconstruction of decay positron tracks necessary for reconstructing the muon beam decay vertices. The track fitting method, GEANE, propagates positrons in the full E989 Geant4 simulation with the advantage of direct access to the geometry, material, and non-uniform magnetic field present within the tracker region. Transport matrices, error matrices, and predicted parameter vectors are generated which are then used in a global  $\chi^2$  minimization algorithm which produces an optimal state vector near the start of the tracker. This procedure was used successfully as described in Chapter 4, which revealed a wealth of beam dynamics information vital to the experiment.

The precession frequency was extracted using an analysis technique called the Ratio Method. The Ratio Method takes the ratio of time-shifted decay positron spectra in order to remove the decay exponential along with slowly varying effects in the data. Precession frequency extraction analyses for four near-final Run 1 datasets were presented with full systematic error evaluations in Chapter 5. The systematic errors are well understood, with independent working groups currently working on improving the largest systematic errors. The total Run 1 precession frequency error determined in this analysis was 491.2 ppb, where the error is statistics dominated.

The expected error in the magnetic field measurement for Run 1 is  $\mathcal{O}(140 \text{ ppb})$ . That combined with the total precession frequency error determined in this analysis would result in an error on  $a_\mu$  of  $\mathcal{O}(510 \text{ ppb})$ , comparable to the uncertainty in the E821 measurement of 540 ppb. For the final Run 1 production datasets with the last data quality cuts and gain improvements, the final errors are expected to improve slightly. An independent measurement of  $a_\mu$  that is statistically consistent with the E821 result would go a long way towards increasing the confidence in the discrepancy between theory and experiment of nearly four standard deviations. The Run 1 publication is expected to be complete in the spring of 2020. Data has already been gathered for Run 2 of E989 and Run 3 will begin in the late fall of 2019, with Run 4 planned for 2021. With the rate improvements seen in Run 2 and the expected increases for Run 3 and Run 4, the target uncertainty goal of 140 ppb is a likely reality. Assuming the same central value for  $a_\mu$  is measured as was done in the previous E821 experiment, then the statistical significance of the discrepancy between the theory and experiment would be pushed over five standard deviations, confirming the existence of new physics.

# Appendix A

## Tracking Derivations

### A.1 Straw measurement angular correction

The tracker straws don't measure U and V coordinates directly, but instead measure the DCA radii deriving from measured hit times. In order to utilize the minimization procedure described in Section 4.2.2 on measured track parameters these radii must first be converted to U and V parameters, and similarly for the U and V errors. To first order the measured DCAs can be used identically as the U and V positions, but it was found that there were slight biases in the truth pulls due to this.

In order to improve the results, angular corrections were made to the DCAs to give more accurate estimates of the “measured” positions. It was found that for the error correction, assuming a straight particle path was sufficient for ideal results. For the position correction, it was found that assuming a circular particle path (constant field) correction for the curved tracks was sufficient. These corrections are dependent on the angle of the track, so it's important to note that during each successive iteration of the track fitting, the “measured” parameters are adjusted by the latest “predicted” momenta which change the angle of the track. The correction depends on whether the track went to the left or right side of the wire. Note that the momentum perpendicular to the straw measurement axis can be ignored since it doesn't affect the U or V value. A summary of the calculation of the right side correction follows, with the left side correction being calculated in a similar manner. See Figure A.1.

To solve for the measured U (or V) value, first use the following trigonometric

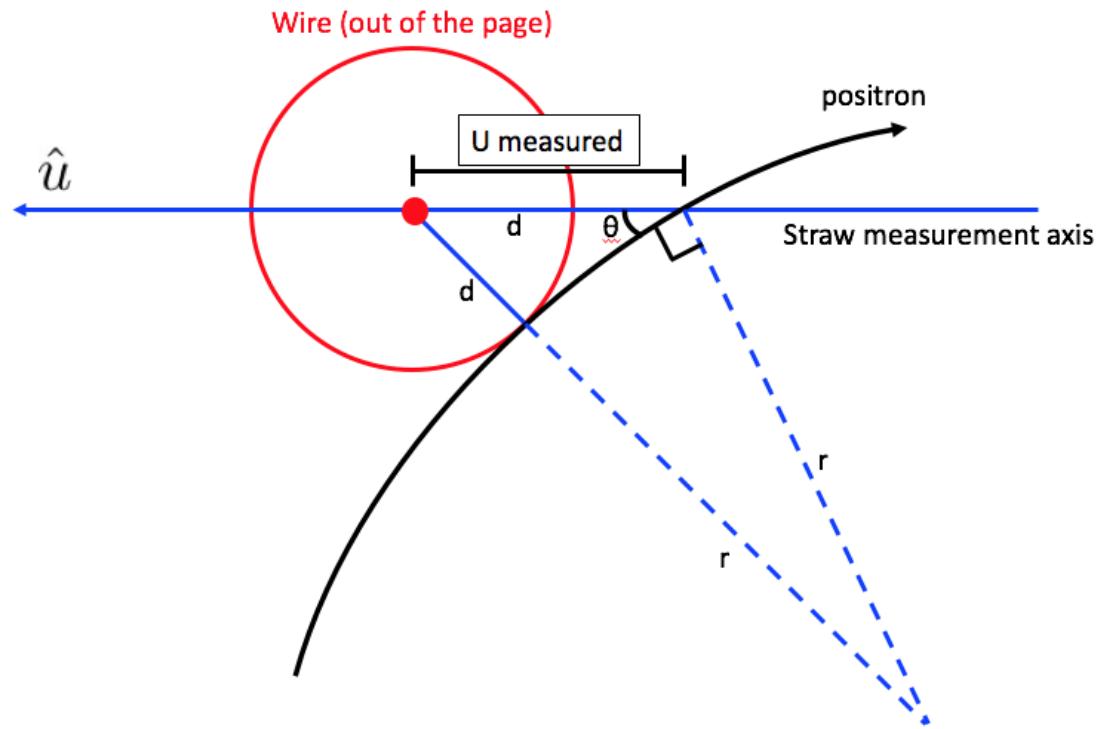


Figure A·1: A positron passing through a straw will produce a hit of radius  $d$ . The desired value is the U or V position along the straw measurement axis. The positron trajectory can be approximated as a circle in a constant magnetic field over the length of the path across the straw. The curvature for the high energy positrons is small such that  $r \gg d$  and the angle between the trajectory and the center of the circle can be approximated as  $90^\circ$ . Sizes and angles are exaggerated. A similar diagram can be drawn for positrons passing to the left of the wire.

identity:

$$(r + d)^2 = r^2 + u^2 - 2ru \cos(90 + \theta), \quad (\text{A.1})$$

where  $u$  is the parameter of interest. The angle  $\theta$  can be determined from

$$\hat{z}_\parallel \cdot \hat{p}_\parallel = \cos \theta, \quad \theta = \cos^{-1} \frac{p_\parallel}{p}, \quad (\text{A.2})$$

where  $p_\parallel$  is the positron momentum anti-parallel to the U measurement axis at the wire plane. Using some further trigonometric identities and solving for  $u$  gives

$$u = -r \sqrt{1 - \left(\frac{p_\parallel}{p}\right)^2} + \sqrt{d^2 + 2dr + r^2 \left(1 - \left(\frac{p_\parallel}{p}\right)^2\right)}, \quad (\text{A.3})$$

for the right side correction. Similarly,

$$u = +r \sqrt{1 - \left(\frac{p_\parallel}{p}\right)^2} - \sqrt{d^2 + 2dr + r^2 \left(1 - \left(\frac{p_\parallel}{p}\right)^2\right)}, \quad (\text{A.4})$$

for the left side correction. The radius  $r$  can be calculated from the momentum and magnetic field at the predicted hit position, and the momentum components can be determined within the Geant4 simulation. The straight line correction to the errors is done simpler manner using the Pythagorean theorem,

$$\sigma'_{uv} = \frac{\sigma_{uv}}{\sqrt{1 - \left(\frac{p_\parallel}{p}\right)^2}}, \quad (\text{A.5})$$

where  $\sigma'_{uv}$  is the improved error from original U or V error  $\sigma_{uv}$  on the DCA.

## A.2 Track fitting correlation matrices

Define  $\epsilon_{MN}$  as the deviation in average predicted track parameters from the true ones on plane  $M$  from track parameters propagated from plane  $N$ ,

$$\epsilon_{MN} = \tilde{p}_M(p_N) - p_M, \quad (\text{A.6})$$

where  $N < M$  can be any plane numbers. From the definition of transport matrices in Equation 4.6,

$$\epsilon_{M0} = T_{MN}\epsilon_{N0} + \epsilon_{MN}, \quad (\text{A.7})$$

where  $\epsilon_{MN}$  is any additional deviation between planes  $N$  and  $M$  due to material effects. By definition, the correlation between error matrices on planes  $M$  and  $N$  is given by

$$\sigma_{MN} = \langle \epsilon_{M0}\epsilon_{N0} \rangle = \langle \epsilon_{M0} \cdot \epsilon_{N0} \rangle - \langle \epsilon_{M0} \rangle \cdot \langle \epsilon_{N0} \rangle. \quad (\text{A.8})$$

The last term in Equation A.8 is zero since  $\langle \epsilon_{M0} \rangle$  and  $\langle \epsilon_{N0} \rangle$  are zero. Therefore

$$\sigma_{MN} = \langle (T_{MN}\epsilon_{N0} + \epsilon_{MN}) \cdot \epsilon_{N0} \rangle, \quad (\text{A.9})$$

and since  $\epsilon_{MN}$  and  $\epsilon_{N0}$  are independent such that their product is zero,

$$\sigma_{MN} = T_{MN}\sigma_N, \quad (\text{A.10})$$

with  $\sigma_N = \langle \epsilon_{N0} \cdot \epsilon_{N0} \rangle$ .

# Appendix B

## Ratio Method Derivation

### B.1 Ratio form and function

Consider the 5 parameter function:

$$N_5(t) = N_0 e^{-t/\tau} (1 + A \cos(\omega_a t + \phi)), \quad (\text{B.1})$$

which describes some ideal dataset in histogram format. Here  $\phi$  will be set to zero for simplicity. Now define the variables  $u_+(t)$ ,  $u_-(t)$ ,  $v_1(t)$ , and  $v_2(t)$  as

$$\begin{aligned} u_+(t) &= \frac{1}{4} N_5(t + T/2) \\ u_-(t) &= \frac{1}{4} N_5(t - T/2) \\ v_1(t) &= \frac{1}{4} N_5(t) \\ v_2(t) &= \frac{1}{4} N_5(t), \end{aligned} \quad (\text{B.2})$$

where the  $1/4$  out front reflects randomly splitting the whole dataset into 4 equally weighted sub-datasets, and  $T$  is the g-2 period known to high precision,  $\mathcal{O}(10^{-6})$ . This corresponds to a weighting of 1:1:1:1 between the datasets. To be explicit here regarding the signs, the counts that are filled into the histogram described by  $u_+$  have their times shifted as  $t \rightarrow t - T/2$ , which is what the function  $N_5(t + T/2)$  describes,

and vice versa for  $u_-$ . To form the ratio define the variables:

$$\begin{aligned} U(t) &= u_+(t) + u_-(t) \\ V(t) &= v_1(t) + v_2(t) \\ R(t) &= \frac{V(t) - U(t)}{V(t) + U(t)}. \end{aligned} \tag{B.3}$$

Plugging in and dividing the common terms ( $N_0 e^{-t/\tau}/4$ ),

$$R(t) = \frac{2(1 + A \cos(\omega_a t)) - e^{-T/2\tau}(1 + A \cos(\omega_a t + \omega_a T/2)) - e^{T/2\tau}(1 + A \cos(\omega_a t - \omega_a T/2))}{2(1 + A \cos(\omega_a t)) + e^{-T/2\tau}(1 + A \cos(\omega_a t + \omega_a T/2)) + e^{T/2\tau}(1 + A \cos(\omega_a t + \omega_a T/2))}. \tag{B.4}$$

Now set  $\omega_a T/2 = \delta$ , and note that  $T$  is really

$$\begin{aligned} T &= T_{guess} = \frac{2\pi}{\omega_a} + \Delta T, \\ \Delta T &= T_{guess} - T_{true}. \end{aligned} \tag{B.5}$$

Being explicit,

$$\delta = \frac{\omega_a}{2} T_{guess} = \frac{\omega_a}{2} \left( \frac{2\pi}{\omega_a} + \Delta T \right) = \pi + \pi \frac{\Delta T}{T_{true}} = \pi + \pi(\delta T), \tag{B.6}$$

and  $\delta$  can be redefined as

$$\delta = \pi(\delta T), \tag{B.7}$$

by flipping the sign of any cosine terms that contain  $\delta$ .

Then, using the trig identity

$$\cos(a \pm b) = \cos(a) \cos(b) \mp \sin(a) \sin(b) \tag{B.8}$$

so that

$$\begin{aligned}
\cos(\omega_a t \pm \delta) &= \cos(\omega_a t) \cos \delta \mp \sin(\omega_a t) \sin \delta \\
&\approx \cos(\omega_a t)(1 - \delta^2) \mp \sin(\omega_a t)\delta \\
&\approx \cos(\omega_a t),
\end{aligned} \tag{B.9}$$

since  $\delta \sim O(10^{-5})$ , the ratio becomes

$$R(t) \approx \frac{2(1 + A \cos(\omega_a t)) - (1 - A \cos(\omega_a t))(e^{-T/2\tau} + e^{T/2\tau})}{2(1 + A \cos(\omega_a t)) + (1 - A \cos(\omega_a t))(e^{-T/2\tau} + e^{T/2\tau})}. \tag{B.10}$$

Expanding

$$e^{\pm T/2\tau} = 1 \pm \frac{T}{2\tau} + \frac{1}{2} \left( \frac{T}{2\tau} \right)^2 \pm \dots, \tag{B.11}$$

replacing and simplifying,

$$R(t) \approx \frac{A \cos(\omega_a t) - C(1 - A \cos(\omega_a t))}{1 + C(1 - A \cos(\omega_a t))}, \tag{B.12}$$

where

$$C = \frac{1}{16} \left( \frac{T}{\tau} \right)^2 \approx 2.87 * 10^{-4}. \tag{B.13}$$

Using the expansion

$$f(x) = \frac{1}{1+x} = 1 - x + x^2 - \dots, \quad |x| < 1, \tag{B.14}$$

and since  $C$  is small, the denominator can be manipulated such that

$$\begin{aligned}
R(t) &\approx (A \cos(\omega_a t)) - C(1 - A \cos(\omega_a t))(1 - C(1 - A \cos(\omega_a t))) \\
&\approx A \cos(\omega_a t) - C + CA^2 \cos^2(\omega_a t),
\end{aligned} \tag{B.15}$$

after dropping terms of  $\mathcal{O}(C^2)$  and higher. In practice the last term is omitted since

it has a minimal effect on the fitted value of  $\omega_a$  [92], and one arrives at

$$R(t) \approx A \cos(\omega_a t) - C, \quad (\text{B.16})$$

the conventional 3 parameter ratio function.

In order to avoid approximations one can instead weight the counts in the histograms as

$$u_+(t) : u_-(t) : v_1(t) : v_2(t) = e^{T/2\tau} : e^{-T/2\tau} : 1 : 1, \quad (\text{B.17})$$

so that

$$\begin{aligned} u_+(t) &= \frac{e^{T/2\tau}}{2 + e^{T/2\tau} + e^{-T/2\tau}} N_5(t + T/2) \\ u_-(t) &= \frac{e^{-T/2\tau}}{2 + e^{T/2\tau} + e^{-T/2\tau}} N_5(t - T/2) \\ v_1(t) &= \frac{1}{2 + e^{T/2\tau} + e^{-T/2\tau}} N_5(t) \\ v_2(t) &= \frac{1}{2 + e^{T/2\tau} + e^{-T/2\tau}} N_5(t). \end{aligned} \quad (\text{B.18})$$

(These factors out front aren't so far off from 1/4 since  $e^{\pm T/2\tau} \approx e^{\pm 4.35/2*64.4} \approx 1.034, .967$ .) Then instead  $R(t)$  becomes

$$R(t) = \frac{2(1 + A \cos(\omega_a t)) - (1 - A \cos(\omega_a t + \delta)) - (1 - A \cos(\omega_a t - \delta))}{2(1 + A \cos(\omega_a t)) + (1 - A \cos(\omega_a t + \delta)) + (1 - A \cos(\omega_a t - \delta))}, \quad (\text{B.19})$$

where the  $e^{\pm T/2\tau}$  terms out front now cancel. Using Equation B.9 again and this time avoiding approximations in  $\delta$ ,

$$R(t) = \frac{2A \cos(\omega_a t)(1 + \cos \delta)}{4 + 2A \cos(\omega_a t)(1 - \cos \delta)}, \quad (\text{B.20})$$

after simplifying. In the limit that

$$\delta = \pi(\delta T) \rightarrow 0 \quad (\text{B.21})$$

since  $\delta T$  is small,

$$R(t) \approx A \cos(\omega_a t), \quad (\text{B.22})$$

with the only approximation being made at  $\mathcal{O}(\delta^2) \sim \mathcal{O}(10^{-10})$ .

Finally, while the 3 parameter ratio function suffices for fits to data containing slow modulations, it does not suffice for faster oscillation features. In that case it is more useful to fit with the non-approximated or simplified version of the ratio,

$$\begin{aligned} R(t) &= \frac{v_1(t) + v_2(t) - u_+(t) - u_-(t)}{v_1(t) + v_2(t) + u_+(t) + u_-(t)}, \\ &= \frac{2f(t) - f_+(t) - f_-(t)}{2f(t) + f_+(t) + f_-(t)}, \end{aligned} \quad (\text{B.23})$$

where

$$\begin{aligned} f(t) &= C(t)(1 + A \cos(\omega_a t + \phi)) \\ f_{\pm}(t) &= f(t \pm T_a/2), \end{aligned} \quad (\text{B.24})$$

and  $C(t)$  can encode any other effects in the data that need to be fitted for, such as the CBO,

$$C(t) = 1 + A_{cbo} \cdot e^{-t/\tau_{cbo}} \cdot \cos(\omega_{cbo}t + \phi_{cbo}). \quad (\text{B.25})$$

Additionally, any other fit parameters such as  $A$  or  $\phi$  can be made a function of  $t$ . Using the non-approximated form for the final fit function gives greater confidence in the fit results for the high precision  $\omega_a$  extraction necessary for the experimental measurement.

## B.2 Ratio errors

In order to determine the errors on the points in the formed ratio, Equation B.3, we use standard error propagation:

$$\sigma_R(t)^2 = \left( \frac{\partial R(t)}{\partial V(t)} \right)^2 \delta V(t)^2 + \left( \frac{\partial R(t)}{\partial U(t)} \right)^2 \delta U(t)^2 \quad (\text{B.26})$$

This works because  $V(t)$  and  $U(t)$  are statistically independent datasets. Using standard error propagation again,

$$\begin{aligned} \delta V(t)^2 &= \delta v_1(t)^2 + \delta v_2(t)^2 = v_1(t) + v_2(t) = V(t), \\ \delta U(t)^2 &= \delta u_+(t)^2 + \delta u_-(t)^2 = u_+(t) + u_-(t) = U(t). \end{aligned} \quad (\text{B.27})$$

Calculating out and simplifying the partial derivatives, (and this time dropping the  $t$ 's),

$$\begin{aligned} \frac{\partial R}{\partial V} &= \frac{2U}{(V+U)^2}, \\ \frac{\partial R}{\partial U} &= \frac{-2V}{(V+U)^2}. \end{aligned} \quad (\text{B.28})$$

Combining and simplifying, we arrive at the error formula:

$$\sigma_R^2 = \frac{4UV}{(V+U)^3} = \frac{1-R^2}{(V+U)} \quad (\text{B.29})$$

## Appendix C

# Pileup Modified Errors

In the pileup subtraction method detailed in Section 5.2.1, pileup events are statistically constructed and then subtracted from the data. Because of this, the errors on the bins need to be adjusted appropriately. Reference [119] describes the modified errors, but is not quite correct. Here is provided an improved calculation that is easier to understand. While we are mainly interested in the errors on the histogram bins after pileup subtraction, it first helps to examine the errors of the pileup histogram itself. Here we only consider doublets.

In the asymmetric shadow window pileup method, shadow doublets are constructed from two singlets. The pileup histogram is then filled as the sum of the doublets minus the singlets,

$$P = D - S, \tag{C.1}$$

where  $D$  or  $S$  are only added or subtracted when they are above some energy threshold. If the threshold is set to zero, then for every doublet one entry will be added and two will be subtracted. Since these entries are exactly correlated, the error in each time bin will be

$$\sigma_P = \sqrt{N_D}, \tag{C.2}$$

where  $N_D$  is the number of doublets in that time bin. If the energy threshold is above zero, then we can determine whether the counts in the pileup histogram increase or

	$E_1 < E_{th}$	$E_1 > E_{th}$
$E_2 < E_{th}$	$N_1(+1)$	$N_2(0)$
$E_2 > E_{th}$	$N_3(0)$	$N_4(-1)$

Table C.1: Table of doublets above threshold. Here  $E_1$  and  $E_2$  are the energies of the two singlets,  $E_{th}$  is the energy threshold, and  $N_i$  are the number of doublets above threshold for the different combinations of  $E_1$  and  $E_2$ . ( $N_1$  is assumed above threshold here.) The numbers in the parentheses indicate the number of counts gained or lost in the pileup histogram.

decrease based on whether the singlets and doublets are above threshold or not. Table C.1 shows the different combinations of counts put into the pileup histogram. The counts that go into  $P$  will be

$$\begin{aligned}
 P &= \sum_i N_i - \text{singlets above threshold} \\
 &= (N_1 + N_2 + N_3 + N_4) - (N_2 + N_4) - (N_3 + N_4) \\
 &= N_1 - N_4
 \end{aligned} \tag{C.3}$$

and the errors are

$$\sigma_P = \sqrt{N_1 + N_4}. \tag{C.4}$$

Consider the individual cases: In the cases for  $N_1$ , you will gain a count from the doublet above threshold, and lose no counts since both singlets are below threshold. In the cases for  $N_2$  and  $N_3$ , you will gain a count from the doublet, and lose a count from one of the singlets which is above threshold. In the cases for  $N_4$ , you will gain a count from the doublet and lose two counts from the singlets which are both above threshold. Since the doublet and singlets are exactly correlated, the  $N_1$  and  $N_4$  cases naturally result in a single weight being added into the error, while the  $N_2$  and  $N_3$  cases result in no additions to the error.

Consider now the pileup subtracted time spectrum? The corrected spectrum can be written as

$$N_{\text{corrected}} = N_{\text{measured}} - P. \quad (\text{C.5})$$

The content in  $N_{\text{measured}}$  doesn't matter exactly. What matters is what is in  $N_{\text{measured}}$  that is also within  $P$ , for that is where the correlations come from. Since  $N_{\text{measured}}$  is the sum of all singlets above threshold, we can write it as

$$N_{\text{measured}} = N_{\text{other}} + N_2 + N_3 + 2N_4 \quad (\text{C.6})$$

since we know that those cases  $N_i$  listed come from singlets above threshold, and  $N_{\text{other}}$  is anything in the measured hits that was not included in the pileup shadow construction. We can then replace  $P$  and simplify to get

$$N_{\text{corrected}} = N_{\text{other}} - N_1 + N_2 + N_3 + 3N_4. \quad (\text{C.7})$$

The error on the corrected histogram is then

$$\sigma_{N_{\text{corrected}}} = \sqrt{N_{\text{other}} + N_1 + N_2 + N_3 + 9N_4}. \quad (\text{C.8})$$

Replacing  $N_{\text{other}}$  as

$$N_{\text{other}} = N_{\text{corrected}} + N_1 - N_2 - N_3 - 3N_4, \quad (\text{C.9})$$

we can remove the dependence of the corrected histogram errors on the unknown quantity and arrive at

$$\begin{aligned} \sigma_{N_{\text{corrected}}} &= \sqrt{N_{\text{corrected}} + 2N_1 + 6N_4}, \\ &= \sqrt{N_{\text{corrected}}} \cdot \sqrt{1 + (2N_1 + 6N_4)/N_{\text{corrected}}}. \end{aligned} \quad (\text{C.10})$$

(This argument might seem circular at the end, but it works because of the squaring that occurs when calculating the error.) In the end we have a form for the bin errors of the pileup corrected histogram which only depend on  $N_1$  and  $N_4$  in addition to the number of counts in the corrected histogram. As shown it can be re-factored into a form equal to the naive errors (just the bin content) times some correction factor. Since  $N_1$  and  $N_4$  are much smaller than  $N_{\text{corrected}}$  at all times, and because they decay away at about twice the rate as the pileup diminishes, the change to the errors is small, of the order 1 or 2% at 30  $\mu\text{s}$ .

## C.1 Pileup errors for the ratio function

Equation C.10 applies to the corrected errors for a pileup subtracted histogram, but what about the modifications to the ratio errors? If we parameterize that equation as

$$\sigma_{N_{\text{corrected}}} = \sqrt{N_{\text{corrected}}} \cdot \sqrt{\gamma(t)}, \quad (\text{C.11})$$

where the correction factor  $\gamma(t) \approx \gamma e^{-t/\tau_\mu}$  is small and decays at approximately the muon lifetime, we can recast the errors on the individual ratio sub-datasets as

$$\begin{aligned} \delta V(t)^2 &= \delta v_1(t)^2 \cdot \gamma(t) + \delta v_2(t)^2 \cdot \gamma(t) = (v_1(t) + v_2(t)) \cdot \gamma(t) = V(t) \cdot \gamma(t), \\ \delta U(t)^2 &= \delta u_+(t)^2 \cdot \gamma(t + T/2) + \delta u_-(t)^2 \cdot \gamma(t - T/2) \\ &\approx u_+(t) \cdot \gamma(t) e^{-T/2\tau} + u_-(t) \cdot \gamma(t) e^{+T/2\tau} \\ &\approx (u_+(t) + u_-(t)) \cdot \gamma(t) \cdot \left(1 + \frac{1}{2} \left(\frac{T}{2\tau}\right)^2\right) \\ &\approx U(t) \cdot \gamma(t), \end{aligned} \quad (\text{C.12})$$

where in the last step the  $\frac{1}{2} \left(\frac{T}{2\tau}\right)^2$  term has been neglected because it's small. With these approximations having been made, the modified errors on the ratio points simply

become

$$\sigma_R^2 \rightarrow \sigma_R^2 \cdot \gamma(t), \quad (\text{C.13})$$

with the correction being the same as that on the pileup subtracted histogram. Credit to Reference [93] for this derivation.

## **Appendix D**

### **Fit Result Correlation Matrices**

Correlation matrices for the 60h, HighKick, 9d, and Endgame datasets. It was found that there were stronger correlations between higher order CBO parameters in the HighKick and 9d datasets than in the 60h and Endgame datasets.

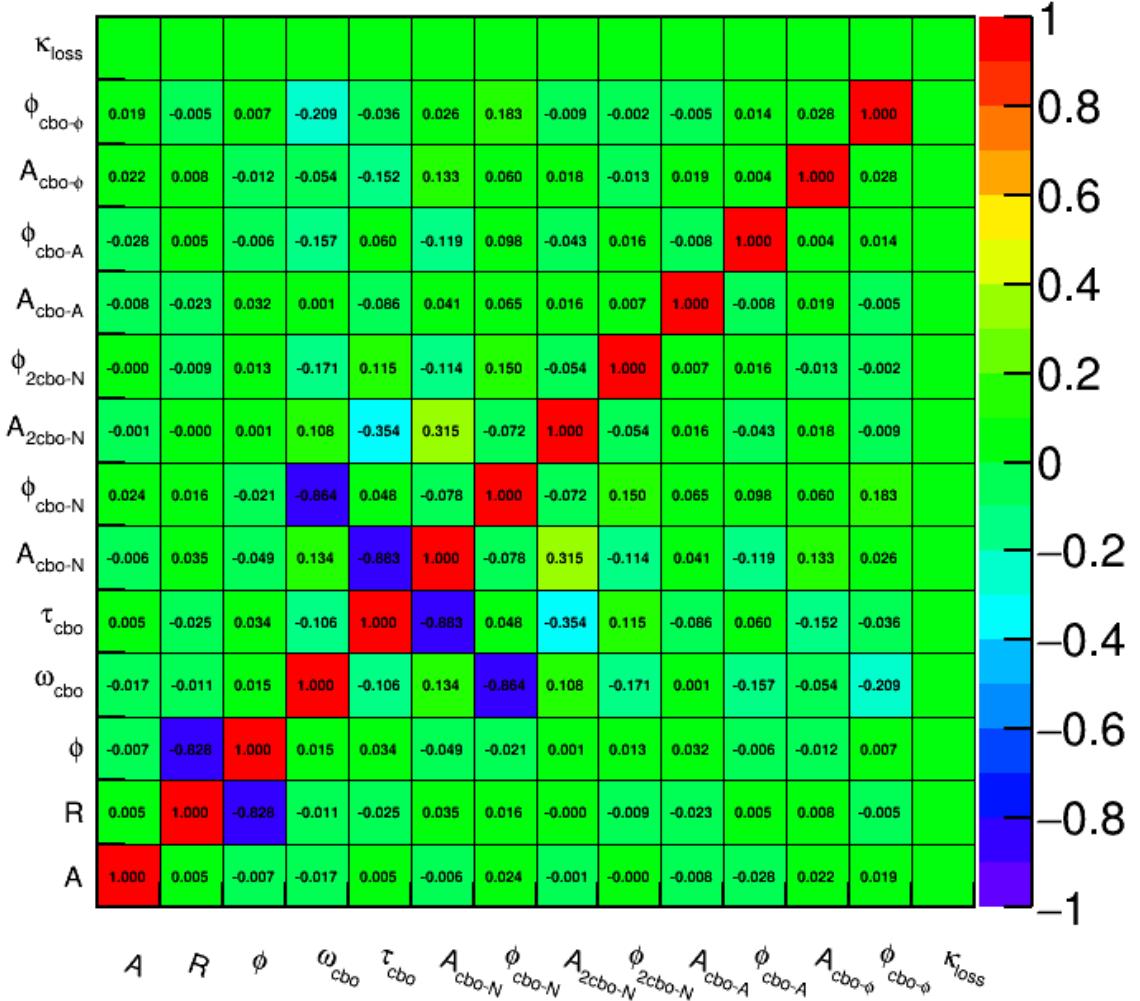


Figure D.1: Correlation matrix for a single seed ratio fit to the 60h dataset. The only significant correlation with  $R$  is the  $g - 2$  phase.  $\kappa_{loss}$  is fixed, hence the corresponding empty row and column.

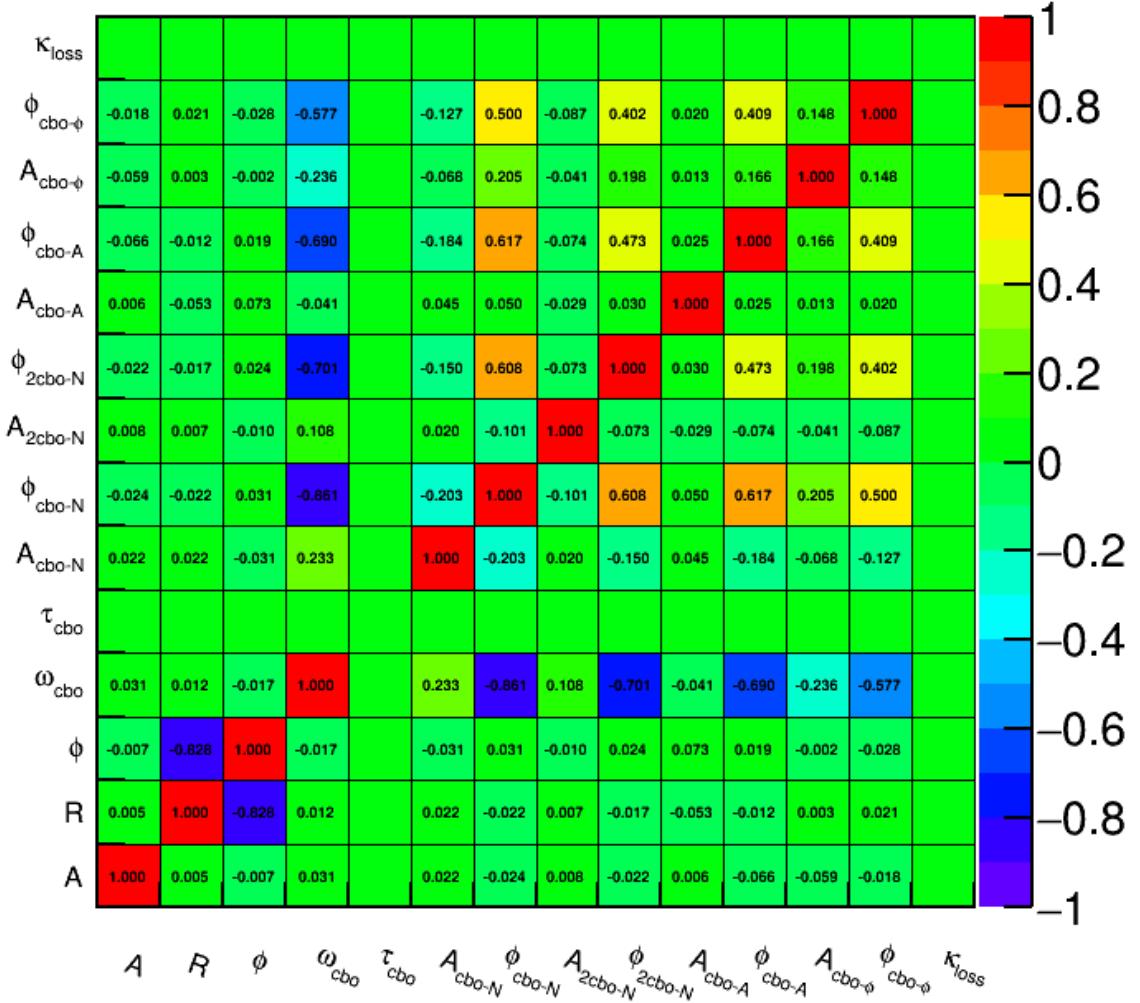


Figure D·2: Correlation matrix for the single seed ratio fit to the High-Kick dataset. The only significant correlation with  $R$  is the  $g - 2$  phase.  $\tau_{cbo}$  and  $\kappa_{loss}$  are fixed, hence the corresponding empty rows and columns.

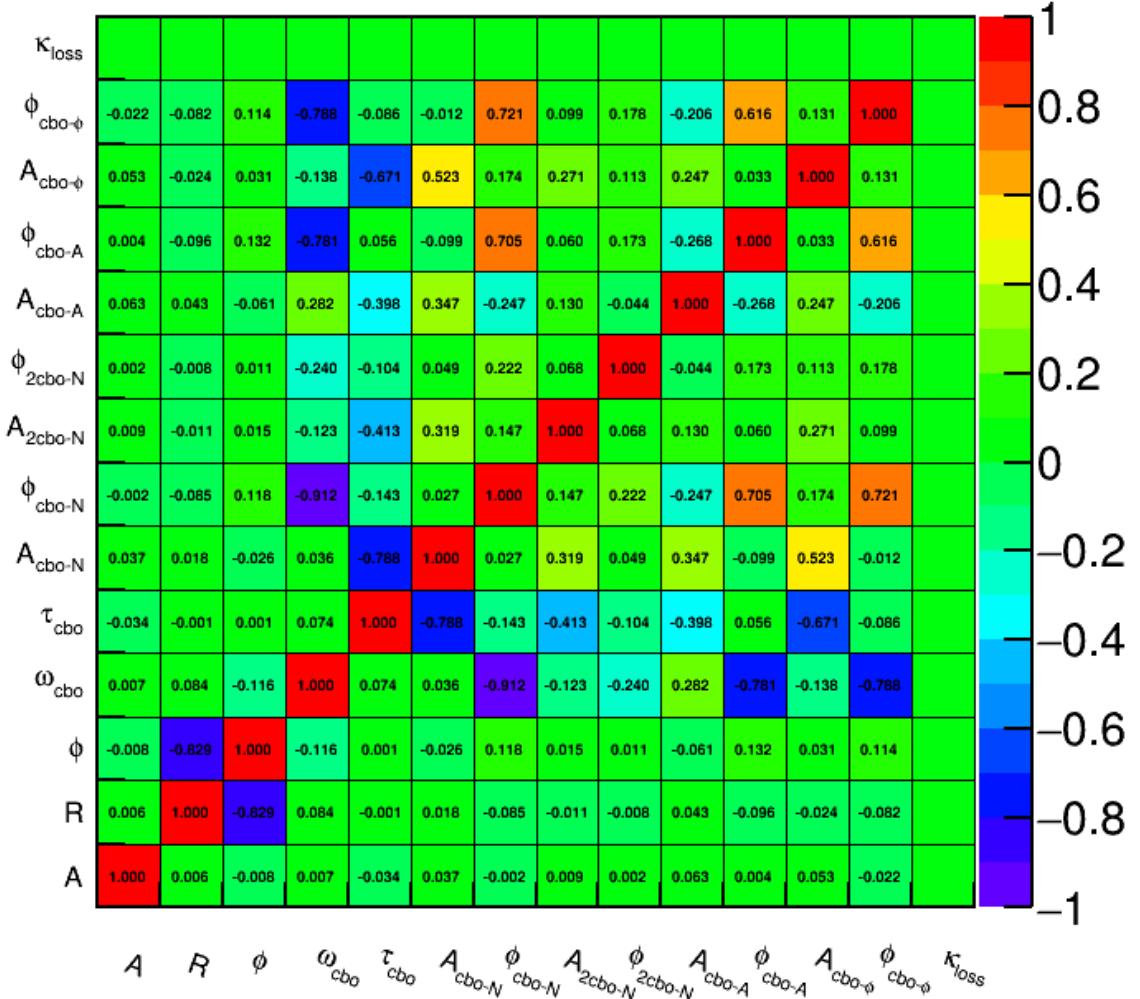


Figure D.3: Correlation matrix for the single seed ratio fit to the 9d dataset. The only significant correlation with  $R$  is the  $g-2$  phase.  $\kappa_{loss}$  is fixed, hence the corresponding empty row and column.

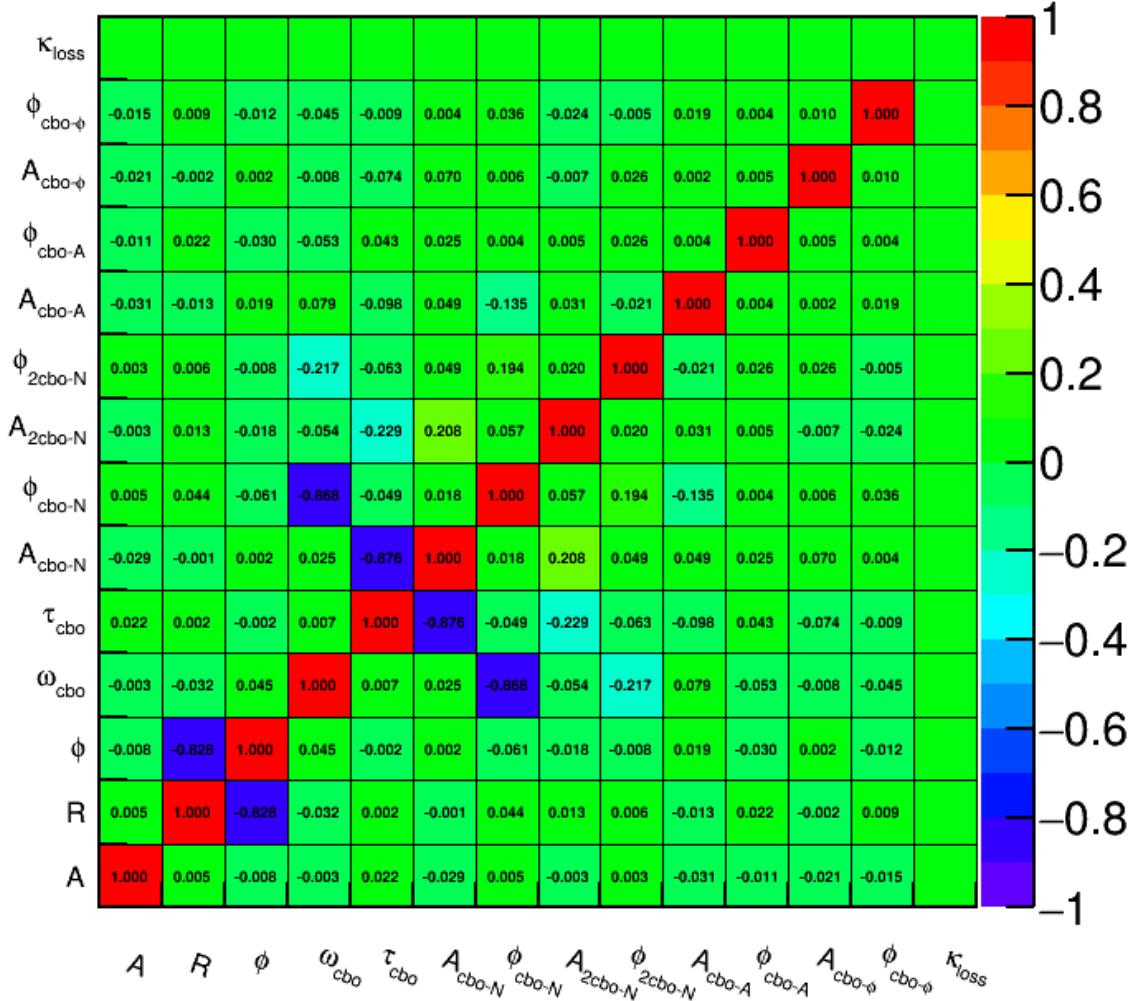


Figure D·4: Correlation matrix for the single seed ratio fit to the Endgame dataset. The only significant correlation with  $R$  is the  $g - 2$  phase.  $\kappa_{loss}$  is fixed, hence the corresponding empty row and column.

## Appendix E

# Ratio Method – T-Method Randomization Comparison

The final value of  $\omega_a$  and by extension  $a_\mu$  will be some type of combination or average between different analyzers and their different fit types. Ratio Method fits due to their additional level of randomization will result in fitted  $R$  values that are consistent but differ from T-Method fits to the same data. In order to assist in combining the Ratio Method results with T-Method results, it is desirable to understand the expected difference between the two fit types.

In order to investigate this a toy MC study was performed with 100 pseudo-experiments. Each pseudo-experiment consisted of a single set of “positron” data, generated using a 1D `ROOT` function corresponding to an energy threshold time histogram. The amount of stats in each pseudo-experiment was chosen to be comparable to the 60h dataset. This single set of hits was then time-randomized and ratio-randomized with 50 different random seeds, where the times of the generated hits were randomized in the same way as was done for data. In this way each pseudo-experiment corresponds to an idealized approximation of the 60h dataset, with 50 different random seeds applied before fitting.

Five parameter fits and three parameter ratio fits were then performed on the histograms for all random seeds, and for all pseudo-experiments. The distribution of fitted  $R$  values for a single pseudo-experiment is shown in Figure E.1a. While the true  $R$  value is 0 as chosen in the simulation, the  $R$  values for a single pseudo-experiment

<b>T-R Method Randomization Comparison</b>			
Stat level and randomization	Error on $R$	Mean Difference RMS	Error on RMS
~ 60h stats	1322.3	70.1	5.0
~ $10 \times 60$ h stats	416.4	29.3	2.1
~ 60h stats w/ VW randomization	1374.9	84.6	6.0

Table E.1: RMS of mean difference between T and R Method fits with different levels of statistics. In each case times are randomized by  $\pm T_c/2$  where  $T_c$  is the cyclotron period or bin width. In the third row times are also randomized by  $\pm T_{VW}/2$  where  $T_{VW}$  is the VW period, here set to 485.709 ns corresponding to an  $n$  value of 0.120. Units are in ppb.

will be centered around some other value that is statistically consistent with 0. (This is satisfied since the statistical error on  $R$  is approximately 1.3 ppm.) Plots for all pseudo-experiment  $R$  distribution means and widths are shown in E.1b and E.1c respectively. In the former, the means of the plotted histograms should be and are consistent with 0, with widths corresponding to the statistical error on  $R$ . In the latter the widths of the T-Method and Ratio Method fits due to the randomization is shown. It can be seen that the Ratio Method, while having the same statistical fit precision as the T-Method, has a larger width due to the randomization of counts in the ratio histograms.

Finally, the differences in the mean of the  $R$  distributions between the T-Method fit and the Ratio Method fit for each pseudo-experiment is plotted in Figure E.1d. It is the width of this histogram that is the number of interest. What this plot shows is that for a dataset with approximately the same statistics as the 60h dataset, an average  $R$  value for 50 different random seeds for a T-Method fit that is 70 ppb different from that for a ratio fit, is statistically consistent to  $1\sigma$ . Table E.1 includes the expected differences for studies done with 10 times the stats of the 60h dataset, and with the VW randomization applied to the hit times.

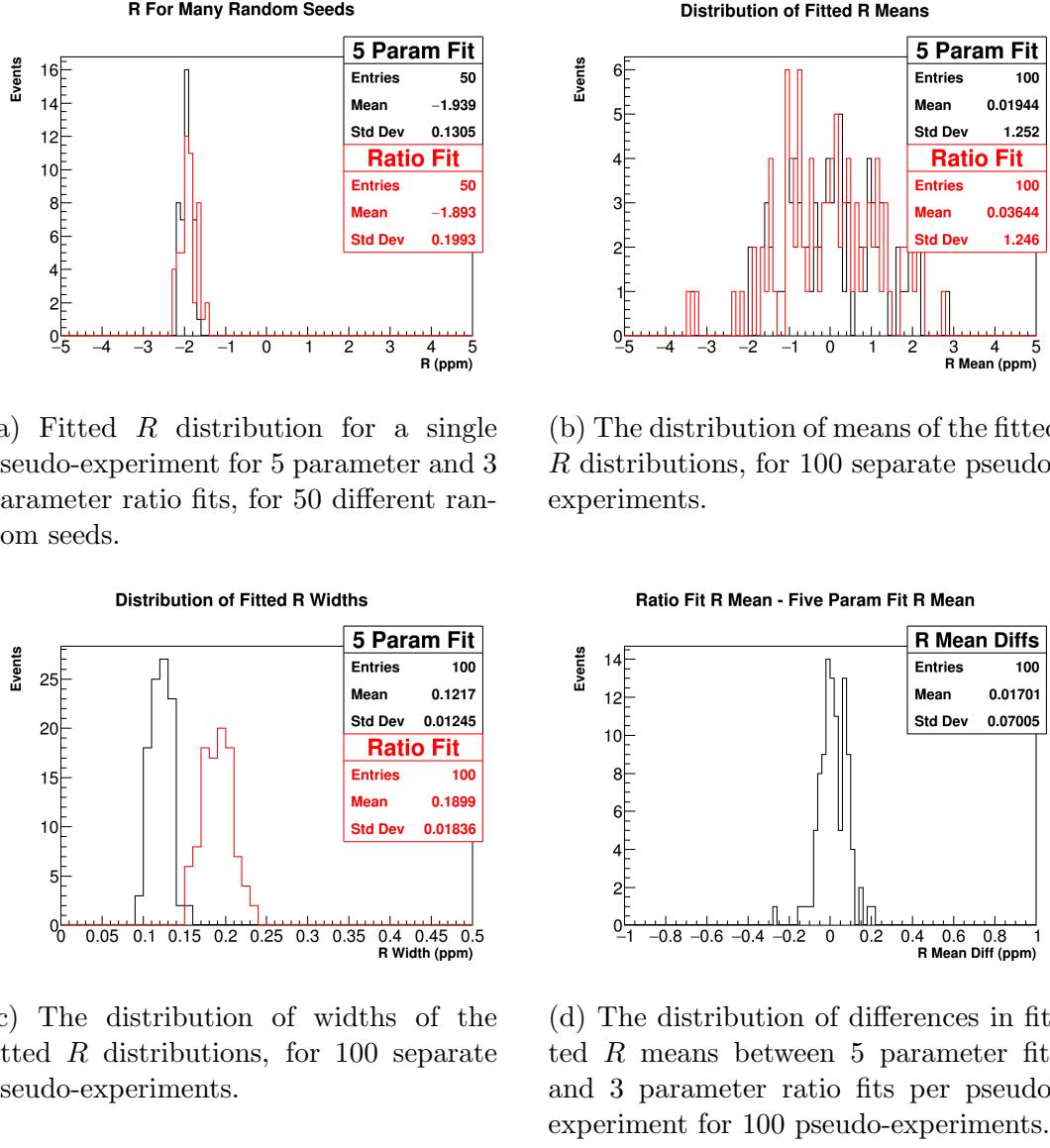


Figure E.1: Toy MC study investigating differences in expected  $R$  values between Ratio Method and T-Method fits due to different randomizations. The statistical precision on  $R$  for a single fit is approximately 1.3 ppm.

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