# Studies of Optimal Track-Fitting Techniques for the DarkLight Experiment 

by

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#### Abstract

The DarkLight experiment is searching for a dark force carrier, the $\mathrm{A}^{\prime}$ boson, and hopes to measure its mass with a resolution of approximately $1 \mathrm{MeV} / \mathrm{c}^{2}$. This mass calculation requires precise reconstruction to turn data, in the form of hits within the detector, into a particle track with known initial momentum. This thesis investigates the appropriateness of the Billoir optimal fit to reconstruct helical, low-energy lepton tracks while accounting for multiple scattering, using two separate track parameterizations. The first method approximates the track as a piecewise concatenation of parabolas in three-dimensions, and (wrongly) assumes that the $y$ and $z$ components of the track are independent. When tested using simulated data, this returns a track which geometrically fits the data. However, the momentum extracted from this geometrical representation is an order of magnitude higher than the true momentum of the track. The second method approximates the track as a piecewise concatenation of helical segments. This returns a track which geometrically fits the data even better than the parabolic parameterization, but which returns a momentum which depends on the seeds to the algorithm. Some further work must be done to modify this fitting method so that it will reliably reconstruct tracks.


Thesis Supervisor: Peter Fisher
Title: Professor

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## Chapter 1

## Introduction

Track reconstruction is essential for analyzing data produced by tracking particle detectors. While they are ubiquitous, different experiments require different types of track-fitting methods.

This chapter will describe the motivations behind the DarkLight experiment and the need for an accurate particle track reconstruction algorithm. Chapter two will cover prior work done in simulating and fitting these tracks. Chapter three will describe the problems with current track-fitting solutions, and requirements for a more appropriate one for this experiment. Chapters four and six will cover two different approaches to this problem, while chapters five and seven will cover their respective results. Lastly chapter eight contains conclusions and a discussion of possible future work.

### 1.1 Dark Matter and DarkLight

There is significant evidence that the universe contains what is called dark matter: matter that we have not yet observed because it does not interact electromagnetically, but is around in large enough quantities that we can detect its cosmological effects. There are currently many theories for what dark matter consists of - axions, weakly interacting massive particles (WIMPS), neutrinos, supersymmetric particles, and more - and many more experiments searching for them, either directly or in-
directly. New theories, motivated in part by evidence for the excess production of electron/positron pairs in our galaxy, predict a dark force, mediated by a neutral light (low-mass) particle, called the $\mathrm{A}^{\prime}$ boson. The $\mathrm{A}^{\prime}$ is produced in dark matter interactions, but also decays via electron/positron pair production [4].

The DarkLight experiment (Detecting A Resonance Kinematically with eLectrons Incident on a Gaseous Hydrogen Target) is currently being developed to create and observe this new boson through the following process:

$$
\begin{equation*}
e^{-}+p \rightarrow e^{-}+p+e^{-}+e^{+} \tag{1.1}
\end{equation*}
$$

DarkLight will be rum using the Jefferson Lab Free Electron Laser (FEL), using a $1 \mathrm{MW}, 100 \mathrm{MeV}$ electron beam incident on a gaseous hydrogen (proton) target. This collision will cause pair production through electromagnetic processes, which are the most significant source of background, but should also pair produce through $\mathrm{A}^{\prime}$. An excess of electrons and positrons at a specific total energy would allow us to determine the mass $m_{A^{\prime}}$ of the particle and coupling strength $\alpha^{\prime}$ of the corresponding force. DarkLight is designed to detect $\mathrm{A}^{\prime}$ in the mass range of $10-100 \mathrm{MeV}$ and coupling strength range of $10^{-9}-10^{-7}$.

### 1.2 Tracking Chambers

One of the challenging design aspects of the DarkLight detector is the ability to detect all four final-state particles in order to fully reconstruct an event [4]. Since the interactions that we wish to study happen on very small time scales, we cannot observe particles traversing the detector in real-time. Instead, particle detectors gain information about decay products in two ways, through calorimetry and tracking. Calorimeters measure the total amount of energy deposited by particles as they slow down and stop in the material, while tracking chambers record position and possibly timing of hits, where particle interact with the material.

The type of tracker that will be used for DarkLight is not currently fixed. However,
the current design calls for a time-projection chamber. It is a cylindrical, grounded chamber, filled with a mixture of argon gas and carbon dioxide. The cylinder has an electrode in the middle, producing an electric field along its axis pointing towards the ends, which contain a number of high-voltage wires. As particles travel through this chamber, they knock into the gas molecules, ionizing them. These freed electrons then accelerate towards the charged wires, ionizing even more molecules and creating a shower. When these ions hit the charged wires, they create a change in voltage along them, proportional to the number of ions and therefore proportional to energy of the original particle. The wires lie along the beam axis $(\hat{z})$, so the detector measures a triple ( $x, y, t$ ) for each "hit", or interaction with the gas. Here, $x$ and $y$ are discrete, set by the location of the wires, and $t$ is the time of the signal. The $z$-coordinate of the hit is calculated based on the drift time, as ions from farther away will take longer to reach the wires. Other possible designs include GEM (Gas Electron Multiplier) trackers, which also operate using the principle of detecting electron showers.

Independent of the chamber design, an approximately homogenous, constant magnetic field is applied to the detector volume, also along the beam axis $(\vec{B}=B \hat{z})$. A charged particle of charge $q$ traveling with a momentum $p_{T}$ in the transverse ( $\mathrm{x}-\mathrm{y}$ ) plane will move in a circle in that plane with a radius of

$$
\begin{equation*}
R=\frac{p_{T}}{q B} \tag{1.2}
\end{equation*}
$$

If the particle also has some momentum $p_{Z}$ parallel to the field, its overall trajectory becomes a helix. The projection in the $x$ - $y$ plane is still a circle, and the amount traversed in $z$ as a function of the arclength $s$ along the helix is linear.

### 1.3 Track Reconstruction

Simply knowing the points of interaction of the particle with the detector is fairly useless. In order to extract any meaningful physical quantities, the hits have to be reassembled into the tracks that they came from. For early detectors, such as bubble
chambers, which detect single tracks continuously, this task was done geometrically by hand. For more complex experiments, where data is recorded digitally and hits are sparser, this becomes more complicated.

First, when there are multiple tracks which overlap each other, the hits need to be separated out and grouped based on which particle they originated from. Assuming that this can be done, the hits can be fit by a continuous track, in order to determine particle type, initial momentum, and initial position.

Using the initial momentum and masses of all four final-state particles (two electrons, one positron, and one proton), we can calculate the center of mass energy of the particle that created it, whether that be a photon or an $\mathrm{A}^{\prime}$. After many events (of the interaction described in equation 1.1), there should be a slight excess of events with invariant mass $m_{A^{\prime}}$. The coupling strength can be extracted by comparing this excess (the signal) to the number of background events (electromagnetic noise).

The precision, or resolution of these measurements depend on the error of measured momentum, which in turn depends on the errors in the fitting algorithm. It also depends on the detector's resolution in position and energy, which depends on design and material properties. A more precise fitting algorithm allows more leeway in the design of the detector.

The goal of my thesis was to write a track-fitting program that could be used to evaluate and optimize detector design in order to have the mass resolution required for the experiment. Finalization of the detector design is necessary in order to secure funding, and in order to start physically building and testing detector components.

## Chapter 2

## Prior Work

Tracking algorithms are ubiquitous in particle physics experiments. They are extremely important for extracting any information from the huge number of hits that are generated in any accelerator-based particle detector. However, there are many problems with tracking algorithms that make them very difficult to reuse.

First and foremost, all tracking programs are either specific to or optimized for the detector geometry of the experiment they are written for. This means that they are written for specific shapes of trajectories (helices vs. lines), specific types of detectors (wire chambers vs. spark chambers), or specific detector materials (argon vs. helium), and are difficult, if not impossible, to generalize to other experiments. In addition, different experiments have different requirements; one experiment may need a very fast but rudimentary algorithm for discarding unneeded data (triggering), while another may need a very accurate fit but have no time constraints.

Even if existing code is appropriate for reuse, it can be very difficult to adapt. My first attempt to create a track-fitter was trying to adapt the Kalman Filter from the Mu2e collaboration, which was originally written for the BaBar experiment. During this attempt, I found that the Mu2e code was so dependent on other collaboration-specific software - the uncommon build system, statistical libraries, modeling libraries, and more - that the optimizations and accuracy guaranteed by reusing their software was not worth the very long time that it would take to untangle it and adapt it to our group's needs. For this reason, it was necessary to write a
track fitter from scratch, one that can be used to extract physical parameters of the particle using information about the detector, and not just a geometric description of the track.

### 2.1 Geant4 Detector Simulation

In order to test any track-fitters, there has to be a track with known parameters. These tracks were generated by a simulation of the detector.

I wrote the simulation in C++ using ROOT (a statistical analysis framework) and Geant4 (a particle simulation toolkit). The simulation takes particles and moves them through the detector, simulating the interactions with the materials in the detector. Depending on the interaction, the particle can deposit energy in the detector. This is recorded as a hit, and the position ( $x, y, z$ ), energy deposited $d E$, and volume (tracking chamber vs. calorimeter) are recorded. These positions are exact, up to truncation, because they are the positions of interaction within the gas, not of the hit along the chamber wires.

The simulation can be seeded in two ways. First, a particle can be shot out of a "particle gun", meaning that the event simply contains a particle with a given initial position and momentum. Alternatively, the simulation can read in input from MadGraph, which is a program which numerically calculates cross-sections of different processes given their Feynman diagrams, and also generates events. It can produce possible initial momenta for the final-state particles produced either by photons of various energy (background) or by $\mathrm{A}^{\prime}$ of a specific mass (signal). An accurate trackfitter should be able to fit a track produced by the particle gun and calculate an accurate initial momentum. It should then be able to fit tracks produced by MadGraph and use the fitted momenta to calculate an accurate center-of-mass energy for the event.

The simulation outputs "events", which contain all of the tracks produced by a single interaction. Each track contains information about the initial position and momentum, as well as a vector of hits. Each hit includes the coordinates $(x, y, z)$,
deposited energy $\mathrm{d} E$, and what part of the detector those coordinates lie in (e.g. tracker, scintillator, beam dump, etc.). In reality, only a portion of this information can be detected; the other (such as initial momentum) must be calculated from the data. However, this allows us to test the track-fitters.

The type of hits generated by the simulation depend on the specific geometry of the detector. Because the detector design is so fluid, the simulation has been modified by other members of the group to be able to read in geometry from an XML file. This allows things like the magnetic field strength, inner and outer radii of the tracker, or gas concentration, to be easily changed and compared. A good track-fitter will be able to give concrete values for the errors associated with track parameters for different geometries.

### 2.2 Karimäki Circle Fit

A preliminary track fitter was written by Ray Cowan in ROOT. It fits tracks in two consecutive steps. First, it takes the projection of the hits in the $x-y$ plane and fits the hits to a circle. It then calculates the arclength $s$ for each point and fits the s-z projection to a line.

The circular $x$-y fit is implemented using the Karimäki fit. This fit performs a least-squares minimization in $x$ and $y$ to return the radius $R$ and center point ( $x_{0}, y_{0}$ ) of the circle that best fits the points [6]. There is no order to the points, so all points are equally weighted.

The second step is to use the circular fit parameters to calculate the arclength $s$ along the circular projection at each point. A standard linear least-squares regression is performed to fit $z$ linearly to $s$.

## Chapter 3

## Problem Statement

Although the Karimäki circe fitter does a decent job of getting good estimates for initial momentum, it is not very precise. This is because the technique does not take into account a very important physical process that occurs in particle detectors: multiple scattering.

### 3.1 Multiple Scattering

As a high-energy ( $E \gtrsim 10 \mathrm{MeV}$ ) electron (or positron) moves through a medium, it interacts electromagnetically. Coulomb repulsion with atomic electrons causes the particle to scatter, ionizing the atoms. It is this interaction that allows the detector to function, but it also changes the particle trajectory. First, the particle scatters, and second, the particle loses the energy required for ionization. This changes both the direction and magnitude of the momentum, and changes the expected smooth helical trajectory to one that has kinks (from scattering) and spirals inward (from energy loss).

The amount of scattering that occurs is characterized by a property of the material called the radiation length, $X_{0}$. The energy of a particle that has travelled a single radiation length drops to $1 / e$ of its initial value. A particle with initial energy $E_{0}$, after traveling a distance $x$ through a material, has an energy $E=E_{0} e^{-x / X_{0}}$.

Radiation lengths for different materials are well known. For a mixture, the radi-
ation length is given by

$$
\begin{equation*}
\frac{1}{X_{0, \text { mix }}}=\sum_{j \text { in mix }} \frac{w_{j}}{X_{0, j}} \tag{3.1}
\end{equation*}
$$

where $w_{j}$ is the fraction by weight of component $j[5]$. The current iteration of the DarkLight tracker is filled with an $80: 20$ mixture of argon gas ( $X_{0}=19.55 \mathrm{~g} / \mathrm{cm}^{2}$ ) and carbon dioxide ( $X_{0}=36.20 \mathrm{~g} / \mathrm{cm}^{2}$ ) [5] at a density of $\rho=.0018 \mathrm{~g} / \mathrm{cm}^{3}$; this mixture has a radiation length of 119.6 m . This is much larger than the length of the detector (a single meter), so the energy loss is less significant than the direction change.

The scattering angle can be approximated as a Gaussian variable $\alpha$ with standard deviation

$$
\begin{equation*}
\sigma_{\alpha}=\frac{13.6 \mathrm{MeV}}{p v} \sqrt{\xi}(1+.038 \ln \xi) \tag{3.2}
\end{equation*}
$$

where $\xi=\frac{x}{X_{0}}$ is the number of radiation lengths of the material the particle has passed through, $p$ is the total momentum, and $v$ is the total velocity [5]. Because the electrons produced in the DarkLight experiment are much lower energy than those produced in many high-energy experiments, they can scatter at much larger angles. This makes an effect that is negligible at high energies very important, especially because our goal is precision.

### 3.2 Karimäki Fit Problems

This fitter was tested using both "particle gun" and MadGraph simulated data. Although it does a good job at providing rough estimates of the particle momentum, and fits the tracks fairly well, as shown in figure 3-1, the fit fails to be valid over the entire track. This is due to not accounting for multiple scattering.

This discrepancy holds especially true for longer tracks, or ones that loop around multiple times. Because they are in the tracker for longer, they scatter more and the effects on the track become more significant. Loopers especially are not very optimally fit using the Karimäki fitter, because as the electron stays in the tracker longer, it spirals inward and creates multiple circles in the $x-y$ projection. This causes the overall calculated radius to decrease, and the calculated initial momentum will


Figure 3-1: Fitted counterclockwise track of an electron produced by an $\mathrm{A}^{\prime}$. The hits farther along in $z$ do not lie on the fitted track because of energy loss (y-z projection, right), and the hits do not lie perfectly along a circle because of scattering (x-y projection, left).
always be less than in reality. In addition, the scattering causes the calculated circle center to move, which will change the initial fitted direction of the track.

## Chapter 4

## Separable Billoir Fit

In order to get an accurate reconstruction of the initial track momentum and position, we require a method that takes into account multiple scattering. One such method is Pierre Billoir's method for track fitting, developed in the 1980's for high-energy physics at CERN [3].

### 4.1 Billoir Recursive Method

Billoir's method is a recursive one. Given a set of ordered hits and the fitted parameters and their errors at hit $n$, it calculates the fit parameters and errors at hit $n-1$. This allows for the change in various parameters (such as the radius of the helix) as the track goes along. Note that ordering hits and identifying particles can be done using less precise, faster fits, such as the Karimäki fit already written, but this algorithm is designed to be precise.

The optimized fit parameters for a track starting at hit $n$ are represented by a vector $y_{n}^{\text {opt }}$, and the errors are represented by the information matrix $I_{n}$, which is the inverse of the covariance matrix, $I_{n}=\left(V_{n}\right)^{-1}$. For every dependent variable ( $y$ in two dimensions, $y$ and $z$ in three), one of the parameters has to be the fitted value of that variable at $x_{n}$, in order to easily take into account measured values. The parameters at previous hits are calculated in three steps.

First, the effects of scattering at point $n$ must be taken into account and removed.

Because the scattering angle is uniformly distributed around 0 , scattering only affects the errors. The errors prior to scattering are defined as

$$
\begin{equation*}
I_{n}^{*}=\left(I_{n}^{-1}+A_{n}\right)^{-1} \tag{4.1}
\end{equation*}
$$

where $A_{n}$ is a matrix which depends on the specific track parameterization and on the scattering angle variance $\sigma_{\alpha}^{2}$, defined in equation 3.2.

Next, the parameters are propagated to the previous point. Depending on the parameterization of the track, the parameters can stay exactly the same or change. For example, a line is always defined by two parameters, but what these two parameters are is variable. A line can be defined by its slope and one of its intercepts, and these are constant along the entire line. However, a line can also be defined by its slope and the value at $x_{n}$. At each point $n$, the second parameter changes, but its can be calculated from its value at any of the other points. Likewise, the propagated parameters and their errors are

$$
\begin{align*}
& y_{n-1}^{*}=D_{n} y_{n}^{\mathrm{opt}}  \tag{4.2}\\
& I_{n-1}^{*}=\left(D_{n}^{T}\right)^{-1} I_{n}^{*}\left(D_{n}\right)^{-1} \tag{4.3}
\end{align*}
$$

where $D_{n}$ is a matrix, again dependent on the specific choice of parameterization.
Lastly, the method incorporates the actual measurements at hit $n-1$ through the following equation

$$
\left(I_{n-1}^{*}+M_{n}\right)\left(y_{n-1}^{\mathrm{opt}}-y_{n-1}^{*}\right)=M_{n}\left[\begin{array}{c}
y_{n-1}^{\mathrm{m}}-y_{n-1}^{*}  \tag{4.4}\\
\left(z_{n-1}^{\mathrm{m}}-z_{n-1}^{*}\right) \\
0 \\
\vdots
\end{array}\right]
$$

where $M_{n}$ is a matrix which depends on $\sigma_{y}$. The vector on the right-hand side of equation 4.4 has values $y_{n-1}^{\mathrm{m}}-y_{n-1}^{*}$ in location corresponding to the parameter $y$ (the same for $z$, if it is part of the fit), and zeroes elsewhere. The solution for $y_{n-1}^{\mathrm{opt}}$
is the vector of optimal parameters post-scattering at hit $n-1$. The errors of the parameters are

$$
\begin{equation*}
I_{n-1}=I_{n-1}^{*}+M_{n} \tag{4.5}
\end{equation*}
$$

Because this is a recursive method, it requires initial values for the parameters and their errors. Billoir is ambiguous about what these initial values should be, but suggest assigning very large arbitrary variances (small information) and claims that "the fitted values are nearly independent of the initial ones."

### 4.2 Separable Track Parameterization

As noted in the previous section, the tracks we wish to fit are helical in shape. Fitting this helix can be separated into two problems: fitting the $x-y$ projection by a circle, and fitting the s-z projection by a line. The Karimäki circle fit works best for many points, but for small arclengths, it isn't terribly effective. Instead, we can approximate the circle as a concatenation of parabolic segments between each hit. This fits into the linear model of the Billoir method, and also makes it still possible to calculate the curvature at each point, which would not be so if the circle were approximated with line segments; the curvature of a line is definitionally zero.

At each point, the track can be described by:

$$
\begin{equation*}
y=Y+a x+\frac{c}{2} x^{2} \tag{4.6}
\end{equation*}
$$

The curvature, $\kappa=1 / R$, for a curve is

$$
\begin{equation*}
\kappa=\frac{x^{\prime} y^{\prime \prime}-y^{\prime} x^{\prime \prime}}{\left(x^{2}+y^{\prime 2}\right)^{3 / 2}}=\frac{y^{\prime \prime}}{\left(1+y^{\prime 2}\right)^{3 / 2}}=\frac{c}{\left(1+(a+c x)^{2}\right)^{3 / 2}} \tag{4.7}
\end{equation*}
$$

By rearranging equation 1.2, it is clear that

$$
\begin{equation*}
p_{T}=\frac{q B}{\kappa} \tag{4.8}
\end{equation*}
$$

We can then separate out the components of the transverse momentum:

$$
\begin{align*}
& p_{x}=\frac{p_{T}}{\sqrt{1+y^{\prime 2}}}=\frac{p_{T}}{\sqrt{1+(a+c x)^{2}}}  \tag{4.9}\\
& p_{y}=\sqrt{p_{T}^{2}-p_{x}^{2}}=\frac{(a+c x) p_{T}}{\sqrt{1+(a+c x)^{2}}} \tag{4.10}
\end{align*}
$$

For simplification, instead of calculating the arclength, I approximated $z(x)$, which is a decaying sinusoid, as a set of linear segments between each hit. For close enough hits, these line segments are approximately parallel to the tangent lines. Each segment can be described by:

$$
\begin{equation*}
z=Z+b x \tag{4.11}
\end{equation*}
$$

At each point, the $z$-momentum can be calculated as

$$
\begin{equation*}
p_{z}=b p_{x} \tag{4.12}
\end{equation*}
$$

### 4.2.1 Matrices for the Billoir Method

As stated earlier, the scattering, propagation, and measurements matrices $\left(A_{n}, D_{n}\right.$, and $M_{n}$ respectively) used in the Billoir linear method are dependent on the specific parameterization of the track. The track is separated into $y(x)$ and $z(x)$, each described by the parameters

$$
y_{n}^{\mathrm{opt}}=\left[\begin{array}{l}
y  \tag{4.13}\\
a \\
c
\end{array}\right] \quad z_{n}^{\mathrm{opt}}=\left[\begin{array}{l}
z \\
b
\end{array}\right]
$$

where $y$ and $z$ are the values of the fitted track at $x_{n}$, not the measured coordinates, and $a, b$, and $c$ are as defined in equations 4.6 and 4.11.

The scattering matrices are

$$
A_{n}^{(y)}=\left[\begin{array}{ccc}
0 & 0 & 0  \tag{4.14}\\
0 & \sigma_{\alpha}^{2} & 0 \\
0 & 0 & 0
\end{array}\right] \quad A_{n}^{(z)}=\left[\begin{array}{cc}
0 & 0 \\
0 & \sigma_{\alpha}^{2}
\end{array}\right]
$$

where $\alpha$ is the scattering angle in each plane. It is as defined in equation 3.2, but with the momentum projected noto the respective plane of interest (either $\mathrm{x}-\mathrm{y}$ or $\mathrm{x}-\mathrm{z}$ ).

The propagation matrices depend on $\delta x_{n}=x_{n}-x_{n-1}$, the x -distance between two consecutive hits, as

$$
D_{n}^{(y)}=\left[\begin{array}{ccc}
1 & -\delta x_{n} & \delta x_{n}^{2}  \tag{4.15}\\
0 & 1 & -\delta x_{n} \\
0 & 0 & 1
\end{array}\right] \quad D_{n}^{(z)}=\left[\begin{array}{cc}
1 & -\delta x_{n} \\
0 & 1
\end{array}\right]
$$

The measurement matrices are

$$
M_{n}^{(y)}=\left[\begin{array}{ccc}
1 / \sigma_{y}^{2} & 0 & 0  \tag{4.16}\\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right] \quad M_{n}^{(z)}=\left[\begin{array}{cc}
1 / \sigma_{z}^{2} & 0 \\
0 & 0
\end{array}\right]
$$

where $\sigma_{y}$ and $\sigma_{z}$ are the errors in the measurement of $y$ and $z$ respectively. The matrices for the parabolic parameterization are taken from Billoir's paper, bottom of page 360 [3], stated here in a slightly different notation.

### 4.3 Implementation

I implemented this fitter as a standalone ROOT script, in $\mathrm{C}++$. This enables the program to read in the simulated input data, which is stored in TFiles, a type of ROOT storage object. All matrix operations were written using the TMatrixD (matrix of doubles) and TDecompSVD (singular value matrix decomposition) classes [1].

### 4.3.1 Parameter Seeding

Because three points completely define a parabola, and two points completely define a line, the $\mathrm{x}-\mathrm{y}$ fit was seeded using the exact parameters for the last three points, and the $\mathrm{x}-\mathrm{z}$ fit using the last two points.

$$
\begin{align*}
& \underbrace{\left[\begin{array}{c}
Y_{N-2} \\
a_{N-2} \\
c_{N-2}
\end{array}\right]}_{Y}=\underbrace{\left[\begin{array}{ccc}
1 & x_{N-2} & \frac{1}{2} x_{N-2}^{2} \\
1 & x_{N-1} & \frac{1}{2} x_{N-1}^{2} \\
1 & x_{N} & \frac{1}{2} x_{N}^{2}
\end{array}\right]^{-1} \underbrace{\left[\begin{array}{c}
y_{N-2} \\
y_{N-1} \\
y_{N}
\end{array}\right]}_{Q}}_{P} \begin{aligned}
{\left[\begin{array}{c}
Z_{N-1} \\
b_{N-1}
\end{array}\right] }
\end{aligned}=\underbrace{\left[\begin{array}{cc}
1 & x_{N-1} \\
1 & x_{N}
\end{array}\right]^{-1} \underbrace{\left[\begin{array}{c}
y_{N-1} \\
y_{N}
\end{array}\right]}_{Q}}_{P} \tag{4.17}
\end{align*}
$$

### 4.3.2 Error Seeding

The Geant 4 simulation outputs "exact" positions for each of the hits. This is because the simulation doesn't model the actual wires in the wire chamber; instead, it just records the positions of interactions with the gas. However, the output is not infinitely precise - in fact, it outputs to four decimal places - so I assumed measurement errors of $\sigma_{y}=\sigma_{z}=5 \times 10^{-5} \mathrm{~mm}$.

Billoir is ambiguous about how to initialize the errors, so I tried a few different methods in order to see which one would be most accurate.

1. I calculated the errors directly, propagating the measurement errors

$$
\begin{align*}
\frac{\partial Y}{\partial x_{n}}=\frac{\partial P^{-1}}{\partial x_{n}} Q & =-P^{-1} \frac{\partial P}{\partial x_{n}} P^{-1} Q  \tag{4.19}\\
\frac{\partial Y}{\partial y_{n}} & =P^{-1} \frac{\partial Q}{\partial y_{n}} \tag{4.20}
\end{align*}
$$

$$
\begin{gather*}
\sigma_{a_{N-2}}^{2}=\sum_{n=N-2}^{N}\left(\left(\frac{\partial a_{N-2}}{\partial x_{n}} \sigma_{x}\right)^{2}+\left(\frac{\partial a_{N-2}}{\partial y_{n}} \sigma_{y}\right)^{2}\right)  \tag{4.21}\\
I_{N-2}^{(y 1)}=\left[\begin{array}{ccc}
1 / \sigma_{y}^{2} & 0 & 0 \\
0 & 1 / \sigma_{a_{N-2}}^{2} & 0 \\
0 & 0 & 1 / \sigma_{c_{N-2}}^{2}
\end{array}\right] \quad I_{N-1}^{(z 1)}=\left[\begin{array}{cc}
1 / \sigma_{z}^{2} & 0 \\
0 & 1 / \sigma_{b_{N-2}}^{2}
\end{array}\right] \tag{4.22}
\end{gather*}
$$

2. I assumed arbitrarily large errors, and therefore arbitrarily small information

$$
I_{N-2}^{y(2)}=\left[\begin{array}{lll}
i & 0 & 0  \tag{4.23}\\
0 & i & 0 \\
0 & 0 & i
\end{array}\right]
$$

where $i=$ some arbitrarily small number, like $10^{-10}$.
3. I assumed arbitrary large errors, except in $y$ and $z$, because those errors are only from measurement

$$
I_{N-2}^{y(3)}=\left[\begin{array}{ccc}
\sigma_{y}^{2} & 0 & 0  \tag{4.24}\\
0 & i & 0 \\
0 & 0 & i
\end{array}\right]
$$

4. I assumed arbitrarily small errors, and therefore arbitrarily large information

$$
I_{N-2}^{y(4)}=\left[\begin{array}{lll}
i & 0 & 0  \tag{4.25}\\
0 & i & 0 \\
0 & 0 & i
\end{array}\right]
$$

where $i=$ some arbitrarily large number, like $10^{10}$.
5. I assumed arbitrary small errors, except in $y$ and $z$

$$
I_{N-2}^{y(5)}=\left[\begin{array}{ccc}
\sigma_{y}^{2} & 0 & 0  \tag{4.26}\\
0 & i & 0 \\
0 & 0 & i
\end{array}\right]
$$

Initial values 1 and 3 are the most likely to yield the correct answer. This is because 1 is calculated analytically, and should therefore account for all errors, and 4 should give errors that are large enough that the parameters are allowed to change significantly.

### 4.3.3 Iteration

Instead of actually performing this method recursively, which would be very memoryintensive, I implemented it iteratively, starting from the last hit and working backwards. After each iteration, the momentum was recalculated, and used to calculate the new scattering angle variance and the new fit parameters. The two fits, linear and parabolic, were performed in parallel so that this could be done, since the momentum in $z$ cannot be calculated from $z(x)$ alone.

## Chapter 5

## Results of separable fit

The separable fit was tested against "particle gun" simulated data, where the tracks contained hits only in the tracker region. The following results are for an electron track inputted at the origin with $\vec{p}=(0,50,40) \mathrm{MeV} / \mathrm{c}$.

### 5.1 Position

Figures 5-1 and 5-2 show a comparison between the measured hits and their projections onto the fitted track, with the measured points in blue and the fitted points in red. These were run with error method 1, direct calculation of the information matrix.

Visually, when compared with the length scales of the entire track, the points match up very well; by the first hit of the track, the fitted points are only 2.5 mm away measured ones. However, the algorithm computes errors of about .02 mm at each hit, meaning that if the fit were accurate, a 2 mm discrepancy would be very unlikely.

### 5.2 Momentum

Table 5.1 summarizes the effects of the different initial errors on the calculated momentum. The listed momenta are those calculated at the first hit in the tracker,

Figure 5-1: Comparison of measured and parabolically fitted $y$-coordinates. Discrepancy between the two is approximately 2 mm at each point. The crosses are for easier viewing of the data, not error bars.

## Separable Fit Y



Figure 5-2: Comparison of measured and linearly fitted $z$-coordinates. Discrepancy between the two is approximately 1.5 mm at each point.

Separable Fit Z


Table 5.1: Dependence of momentum, calculated from separable fit, on initial parameter error.

| initial information | $p_{Z}(\mathrm{MeV} / \mathrm{c})$ | $p_{T}(\mathrm{MeV} / \mathrm{c})$ |
| ---: | :---: | :--- |
| (1) calculated | -729.908546 | 906.176933 |
| (2) $i=10^{-10}$ | -734.475075 | 911.868107 |
| (3) $i=10^{-10}$ | -734.475064 | 911.868094 |
| $(4) i=10^{10}$ | -83.593212 | 93.114463 |
| (5) $i=10^{10}$ | -119.978075 | 139.146108 |

located at $\vec{r}=(-9.6,79.4,64.2) \mathrm{mm}$, not the origin.
The fitted momenta span an order of magnitude, and depend greatly on the initial information. Only initial values (1) and (3) have any justification - calculated errors should be accurate, and starting with arbitrarily large errors allows the fit parameters not to get stuck on an inaccurate value - and these are the ones which result in calculated momenta farthest from the true value. The track initially has a total momentum of $64 \mathrm{MeV} / \mathrm{c}$, which lies nowhere in the range, and as momentum can only decrease as the particle moves through the detector, all of the calculated momenta are impossibly high.

### 5.3 Conclusions

There are many reasons why this fit is not very good, and they have to do with our assumptions about the track. First of all, the multiple scattering causes scattering not just in the plane, but in three dimensions. This couples the $y$ and $z$, meaning that this problem is not truly separable.

Secondly, a parabola is a fine approximation of a circle only if they are tangent at the vertex of the parabola. However, this algorithm imposes no constraints on the vertices of the parabolas, and furthermore, constrains the parabolas to be vertical. There is no way to describe a horizontal parabola with equation 4.6, even though visually, from figure $5-1$, we can see that it is probably a better approximation to the track.

The curvature of a parabola is maximized at the vertex, but because none of the hits are at vertices, the curvature at any hit is lower than the curvature of the helix that is being approximated. This means that the calculated momenta will always be higher than the true momentum of the particle. Even though this method fits the points in the track correctly, there is no way to extract any physical information from the fit, and it is therefore useless.

This method could be improved by constraining the parabola at hit $n$ to have its vertex at hit $n$, and removing the directional constraint, but figuring out how to parameterize the track, figuring out how to propagate those parameters, and then deriving the matrices required for calculations would be much more complicated and still more inaccurate than simply implementing a helical parameterization of the track.

## Chapter 6

## Helical 5-Parameter Fit

Although parabolas and lines are easy to calculate, it's clear that in this case, they do not result in a useful fit of a helical track. Instead, it is better to simply fit using a helix. Instead of separating out the variables, the three-dimensional helical track is described by five parameters.

### 6.1 Helical Parameterization

The five parameters which describe a helix, given a hit with x -coordinate $x$, are: $y(x)$, $z(x)$, the radius, pitch (wavelength), and phase (see figure 6.1).


Figure 6-1: The 5 parameters of a helix are the radius $R$, dip angle $\lambda$, azimuthal angle $\phi$ (related to the phase), $y$ and $z$ [2]. Note that the center of the helix is a line parallel to the $z$-axis, but not necessarily going through the origin. $\phi$ and $y$ contain the same information as the center coordinates $x_{0}$ and $y_{0}$.

The five parameters at hit $n$ are represented by the vector

$$
y_{n}^{\mathrm{opt}}=\left[\begin{array}{l}
y  \tag{6.1}\\
z \\
a \\
b \\
d
\end{array}\right]
$$

where $a, b$, and $d$ are functions of $\phi, \lambda$, and $R$ that geometrically describe the components of the particle's momentum:

$$
\begin{align*}
& a \equiv \frac{p_{y}}{p_{x}}=\tan \phi  \tag{6.2}\\
& b \equiv \frac{p_{z}}{p_{x}}=\frac{\tan \lambda}{\cos \phi}  \tag{6.3}\\
& d \equiv \frac{q}{p}=\frac{1}{R B} \sqrt{\frac{1+a^{2}}{1+a^{2}+b^{2}}} \tag{6.4}
\end{align*}
$$

and $p \equiv \sqrt{p_{x}^{2}+p_{y}^{2}+p_{z}^{2}}$ is the total momentum. Instead of having to do a roundabout calculation of curvature, the momenta are easily extracted from these parameters:

$$
\begin{align*}
& p=\frac{q}{d}  \tag{6.5}\\
& \vec{p}=\frac{p}{\sqrt{1+a^{2}+b^{2}}}(1, a, b) \tag{6.6}
\end{align*}
$$

The scattering is described by

$$
A_{n}=\left(1+a^{2}+b^{2}\right) \sigma_{\alpha}^{2}\left[\begin{array}{ccccc}
0 & 0 & 0 & 0 & 0  \tag{6.7}\\
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 1+a^{2} & a b & 0 \\
0 & 0 & a b & 1+b^{2} & 0 \\
0 & 0 & 0 & 0 & 0
\end{array}\right]
$$

where $\alpha$ is the total scattering angle, and therefore depends on the total momentum $p$ instead of a projection in any single plane.

Although the helix parameters are not linear from one point to another, if the hits are close enough together, they can be linearized. Then, the propagation matrix $D_{n}$, which depends on the magnetic field $\vec{B}=B \hat{z}$ is as defined in equation 13 , page 362 of Billoir's paper [3]

Lastly, because $y$ and $z$ are measured independently of each other, the measurement matrix is

$$
M_{n}=\left[\begin{array}{ccccc}
1 / \sigma_{y}^{2} & 0 & 0 & 0 & 0  \tag{6.8}\\
0 & 1 / \sigma_{z}^{2} & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0
\end{array}\right]
$$

### 6.2 Implementation

I implemented the helix fit as an extension to the separable Billoir fit. The only changes I had to make were to the size of the matrices (changed to $5 \times 5$ ), the initialization of parameters, the various calculation matrices $A, D$, and $M$, and the fact that I only had to call one method per iteration, instead of one each for $y$ and $z$.

The parameters were initialized as

$$
y_{N}^{\mathrm{ppt}}=\left[\begin{array}{c}
y_{N}^{\mathrm{m}}  \tag{6.9}\\
z_{N}^{\mathrm{m}} \\
\delta y_{N} / \delta x_{N} \\
\delta z_{N} / \delta x_{N} \\
-.3 / p_{N}
\end{array}\right]
$$

where I assumed an electron with $p_{N}=50$ (in $\mathrm{MeV} / \mathrm{c}$ ), which is on the order of magnitude of the momenta of tracks we expect to see in our detector. The .3 is to account for units of $c$ and powers of ten so that if magnetic fields are expressed in Teslas, the resulting calculated distances are in millimeters.

The errors were initialized as

$$
I_{N}=\left[\begin{array}{ccccc}
1 / \sigma_{y}^{2} & 0 & 0 & 0 & 0  \tag{6.10}\\
0 & 1 / \sigma_{z}^{2} & 0 & 0 & 0 \\
0 & 0 & i & 0 & 0 \\
0 & 0 & 0 & i & 0 \\
0 & 0 & 0 & 0 & i
\end{array}\right]
$$

where $i$ is again an arbitrary value for the initial information known. In this case, I could not exactly calculate the initial errors, because the difference between a tangent and the chord between two points depends on the radius of the helix, which is an unknown.

Additionally, because we want to be able to calculate the initial momentum of the track, at the origin, I added the additional steps of de-scattering the parameters at the first hit and propagating them to $x=0$.

## Chapter 7

## Results of 5-parameter fit

The 5-parameter fit was again test using "particle gun" data with hits only in the tracker region. The following results are for the exact same electron track as in chapter 5 , starting at the origin with $\vec{p}=(0,50,40) \mathrm{MeV} / \mathrm{c}$.

### 7.1 Position

Figures 7-1 and 7-2 show a comparison between the measured hits and their projections onto the fitted helical track, with measured points in blue and fitted points in red. These were run for initial information $i=10^{-10}$, an arbitrarily small number.

The helix points match up much better than those fit parabolically, being an average distance of .02 mm away from the measured points. However, the computed errors also shrank to about $.5 \mu \mathrm{~m}$ at each hit, which is much too small. The point propagated to $x=0$ from the first hit is quite far away from the origin, well outside either of these bounds.

### 7.2 Momentum

Table 7.1 summarizes the effects of the different initial errors on the calculated momentum. The listed momenta are those calculated at the first hit, located at $\vec{r}=(-9.6,79.4,64.2) \mathrm{mm}$.

Figure 7-1: Comparison of measured and helically fitted $y$-coordinates. Discrepancy between the two is approximately .02 mm at each point.

Helical Fit $Y$


Figure 7-2: Comparison of measured and helically fitted $z$-coordinates. Discrepancy between the two is approximately .01 mm at each point.

Helical Fit Z


Table 7.1: Dependence of momentum, calculated from helix fit, on initial parameter error.

| $i$ | $p_{z}(\mathrm{MeV} / \mathrm{c})$ | $p_{T}(\mathrm{MeV} / \mathrm{c})$ |
| ---: | :---: | :--- |
| $10^{-10}$ | 43.864 | 53.181 |
| $10^{-9}$ | 43.864 | 53.181 |
| $10^{-8}$ | 68.922 | 83.361 |
| $10^{-7}$ | 78.340 | 94.680 |
| $10^{-6}$ | 74.141 | 89.634 |
| $10^{-5}$ | 17.972 | 24.948 |
| $10^{-4}$ | 18.185 | 25.252 |
| $10^{-3}$ | 18.389 | 25.548 |
| $10^{-2}$ | 18.381 | 25.536 |
| $10^{-1}$ | 18.380 | 25.535 |
| 1 | 18.380 | 25.535 |

The fitted momenta widely vary with initial information $i$. Those seeded with the smallest $i$ are closest to the actual value ( $p_{T}=53.2 \mathrm{MeV} / \mathrm{c}$ vs. $50, p_{z}=43.9 \mathrm{MeV} / \mathrm{c}$ vs. 40), but still too high. The total momentum grows until $i=10^{-6}$, then jumps down and keeps increasing. The proportion $p_{z} / p_{T}=.82$ is constant, meaning that this dependence on initial information only changes the total momentum (described by $d$ ), but not any of their ratios (described by $a$ and $b$ ).

### 7.3 Propagation of Parameters to Origin

Table 7.2 shows the effect of initial information on the propagation of the first hit backwards to $x=0$. Again, the smallest $i$ yields the point closest to the origin, which is the true initial position of the particle. Interestingly, after $i=10^{-6}$, where the total momentum suddenly drops, the tracks start to overshoot the origin, and then approach it from the other side. This could be an artifact dependent on the values for $\sigma_{y}$ and $\sigma_{z}$ that I assumed, or specific to this track. The exact origin is unknown.

Table 7.2: Dependence of calculated "initial" position and momentum on initial parameter errors.

| $i$ | $y_{0}(\mathrm{~mm})$ | $z_{0}(\mathrm{~mm})$ | $\overrightarrow{p_{0}}(\mathrm{MeV} / \mathrm{c})$ |
| ---: | :---: | :---: | :--- |
| $10^{-10}$ | 11.6 | 20.07 | $(-5.78,59.59,34.18)$ |
| $10^{-9}$ | 11.6 | 20.07 | $(-5.78,59.59,34.18)$ |
| $10^{-8}$ | 24.6 | 25.60 | $(-11.88,91.03,57.20)$ |
| $10^{-7}$ | 27.3 | 26.70 | $(-14.38,102.60,66.09)$ |
| $10^{-6}$ | 26.2 | 26.25 | $(-13.25,97.45,62.11)$ |
| $10^{-5}$ | -231.6 | -69.03 | $(-0.49,28.57,11.34)$ |
| $10^{-4}$ | -226.4 | -67.05 | $(-0.50,28.91,11.49)$ |
| $10^{-3}$ | -219.8 | -64.52 | $(-0.52,29.24,11.64)$ |
| $10^{-2}$ | -220.1 | -64.62 | $(-0.52,29.23,11.63)$ |
| $10^{-1}$ | -220.1 | -64.62 | $(-0.52,29.23,11.63)$ |
| 1 | -220.1 | -64.62 | $(-0.52,29.23,11.63)$ |

## Chapter 8

## Summary, Conclusions, and Future

 WorkTrack reconstruction is essential in particle detector experiments, and is used for many different things, such as particle identification or triggering. There are many different methods for analyzing tracks with different strengths - speed vs. precision, for example - and which method you should implement is greatly affected by what it will be used for.

The goal of this project was to write a track reconstruction program for the DarkLight detector that would be able to reconstruct tracks and precisely extract their initial momenta. Reconstructing the tracks of all four final-state particles in the event of interest would allow us to calculate the center-of-mass energy, and from that the invariant mass of the $\mathrm{A}^{\prime}$ boson. Therefore, the track fitter must be precise enough to yield the required mass resolution of $1 \mathrm{MeV} / \mathrm{c}^{2}$.

Due to this precision requirement, and due to the cylindrical geometry of the detector, we require a track-fitting method that can fit a helical trajectory and accounts for multiple scattering, allowing the energy and momentum to change at each hit. The Billoir method offers a nice solution.

### 8.1 Billoir Method

I implemented the Billoir method using two different track parameterization: a separable parameterization, with parabolic $y$ and linear $z$, and a helical parameterization. Neither of them were accurate enough for our purposes.

### 8.1.1 Separable Fit

The separable fit approximated the track as a series of concatenated parabolas with an axis of symmetry in the y-z plane, and fit the tracks $y(x)$ and $z(x)$ separately. I used the curvature of the track to approximate the radius of curvature of the helical track, and then used this, along with the magnetic field strength, to calculate the momentum at each point. This resulted in a track that was fairly close to the measured points, but momenta that were way too large.

This occurred because a parabola was not an appropriate approximation for this track, and because they were not constrained to have their vertex at a hit. Like a circle approximated by its chords, these parabolas will always lie inside the helix, and therefore have a lower curvature than the actual track, accounting for this with kinks at each hit beyond those caused by multiple scattering. This lower curvature mimics the behavior of a particle with a much higher momentum moving in the same magnetic field. In addition, because multiple scattering couples the two directions, this problem cannot truly be separated. This warranted changing the track parameterization.

### 8.1.2 Helical Fit

The helical fit segments the track into sections of helices with slightly different parameters at each hit. This allows for energy loss and scattering, by modifying the radius and phase of the helix. One of the parameters of the fit was already the momentum, and I was able to decompose it into $x, y$, and $z$ components by using the other helix parameters. This fit resulted in a track that was much closer to the measured points than the parabolic fit, and much lower momenta, in the range of the true momentum for the track.

Because the Billoir method is a recursive one, it has to be seeded with an initial guess of the track parameters and their inverted errors. Billoir suggests assuming arbitrarily large errors, because infinite errors results in zero information, which leads to singularities when we try to invert matrices. However, the resulting fitted momentum depends heavily on this "arbitrary" initial error. A better solution might be to seed this fit with the results of another, more rudimentary fitter.

Another possible problem with the helical fit is that in the propagation step, the Billoir method assumes that the track parameters at one point are linearly related to the track parameters at the previous point. However, for a helix, which is circular and has sinusoidal dependence, this is simply not true, especially if the angle between the points is large. Billoir tested his method on particles with encrgies of hundreds of MeV , which is higher than any of the energies in our experiment. Higher energy tracks have a much lower curvature, so hits that are the same distance apart have a smaller angle between them. For such tracks, linearization may be valid, but for our tracks, with a higher curvature, this may simply not be true. The discrepancy caused by linearizing the parameters can be seen in the fact that propagating the helical parameters backwards from the first hit to $x=0$ does not lead back to the origin.

### 8.2 Future Work

### 8.2.1 Fixing the Billoir Helical Fit

There are many ways that we can possibly improve the accuracy of the helical fit. Firstly, the ambiguity in initial values can be fixed by seeding the fit with values from the Karimäki fit. This will require modifying both fits so that they can be run in sequence, and so that the Billoir fit can actually read in the proper seed values.

Another modification that can be made is to the propagation step in the algorithm. The linear propagation step involving the $5 \times 5$ matrix $D_{n}$ would have to be replaced by nonlinear, exact formulas relating parameters at different points along the curve. These can be derived from existing formulas for changing reference points along a
helical track [2].
Lastly, we can account for the energy loss from the interactions by introducing energy loss into the scattering matrix $A_{n}$. Because energy loss would affect total momentum, which is characterized by the parameter $d$, this would be done by having a non-zero $A_{n(d d)}$ dependent on the energy loss rate $\mathrm{d} E / \mathrm{d} x$ and the distance travelled.

### 8.2.2 Testing and Using the Billoir Helical Fit

Once the helical fit is deemed accurate and precise enough from tests on single particle gun tracks, it can be tested and used in other ways.

First, the algorithm should be tested for robustness with different types of tracks. This means testing with high-momentum tracks, low-momentum tracks, tracks which leave the tracker very quickly, tracks which stay in the tracker and loop around multiple times, tracks which go backwards, and more. The fitter is much more useful if it is general, and if it can be used to reconstruct all tracks that could possibly be produced in the DarkLight detector.

Next, the fitter can be tested on simulated signal events containing all of the finalstate particles. It should be able to reconstruct the momenta well enough that they can be combined to measure the invariant mass of the $\mathrm{A}^{\prime}$. Signal events generated with different possible masses $m_{A^{\prime}}$ should result in different calculated masses, with the necessary resolution.

Lastly, the hits can be discretized to model the placement of wires in the tracking chamber. This creates much larger measurement errors $\sigma_{x}$ and $\sigma_{y}$, which depend on the density of wires, and will change the mass resolution of the detector. Other design changes that will change the mass resolution include the density and type of gas inside the chamber (this changes the average distance between hits) and the strength of the magnetic field. These can all be studied using the Billoir helical fit in order to optimize the resolution of the DarkLight detector and increase our chances of possibly detecting the $\mathrm{A}^{\prime}$ boson.

## Bibliography

[1] ROOT user's guide, February 2011.
[2] J. Alcaraz. Helicoidal tracks. L3 Internal Note 1666, February 1995.
[3] P. Billoir. Track fitting with multiple scattering: A new method. Nuclear Instruments and Methods in Physics Research, A225:352-366, December 1894.
[4] J. Balewksi et al. A proposal for the DarkLight experiment at the Jefferson Llaboratory free electron laser. Proposal, PAC39, May 2012.
[5] J. Beringer et al. (Particle Data Group). Review of particle physics. Phys. Rev. D, 86:010001, Jul 2012.
[6] V. Karimäki. Effective circle fitting for particle trajectories. Nuclear Instruments and Methods in Physics Research, A305:187-191, December 1991.

