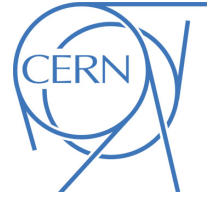




ATLAS Note

EXOT-2017-XX

March 4, 2018



Draft version 0.1

Search for long-lived resonance decaying to a dilepton pair in pp collisions at $\sqrt{s} = 13$ TeV with the ATLAS detector

Siinn Che^a, K.K. Gan^a, Christopher B. Martin^a

^a*The Ohio State University*

A search for long-lived neutral massive particle decaying to a $\mu\mu$, ee , or $e\mu$ pair is presented using the ATLAS detector with 32.8 fb^{-1} of pp collisions at $\sqrt{s} = 13$ TeV at the LHC. Upper limits are presented on the production cross section times branching ratio for resonances decaying to a lepton pair. Also presented is the detection efficiency as a function of p_T and η for a resonance with mass of 0.1–2.0 TeV and lifetime ($c\tau$) of 100–500 mm.

13

Contents

14

1 Introduction 3

15

2 Data and MC samples 4

16

2.1 Data samples 4

17

2.2 MC samples 5

18

2.2.1 Signal samples 5

19

2.2.2 Background MC samples 6

20

3 Data reprocessing 7

21

3.1 Event preselection 7

22

3.2 Reconstruction 8

23

3.2.1 Large radius tracking 8

24

3.2.2 Lepton reconstruction and identification 9

25

3.2.3 Secondary vertex reconstruction 9

26

4 Signal selection 10

27

4.1 Event selection 10

28

4.2 Muon and electron requirements 10

29

4.3 Vertex selection 11

30

5 Signal efficiency 12

31

6 Trigger efficiency 13

32

7 Lepton reconstruction efficiency 14

33

8 Overall reconstruction efficiency 15

34

8.1 Event and vertex cut flow 15

35

8.2 Efficiency distribution 16

36

8.3 Efficiency reweighting 17

37

9 Background estimation 17

38

9.1 Random-crossing background 18

39

9.1.1 MC study 18

40

9.1.2 Estimating random-crossing background with data sample 19

41

9.1.3 Systematic uncertainty in the track flipping method 21

42

9.2 Cosmic background 23

43

9.3 Low-mass background 23

44

10 Systematic uncertainties 24

45

10.1 Systematic uncertainties on tracking and vertexing efficiency 24

46

10.1.1 Data and MC comparison 24

47

10.2 Systematic uncertainties on lepton identification 28

48

10.3 Systematic uncertainties on trigger efficiency 28

49

11 Results 28

50	12 Conclusion	28
51	Appendices	32
52	A Truth-level p_T and η distributions of Signal MC samples	32
53	B K_S yields in data and MC samples used for tracking and vertexing systematic uncertainties	33
54	C K_S and Z' comparison	34
55	D Track flipping extrapolation method	35
56	D.1 Extrapolation from control region	35
57	D.2 Extrapolation from validation region	35
58	D.3 Scale factors	35

59 1 Introduction

60 The search for long-lived particles (LLP) is an important part of the program in searching for new physics
61 at the LHC. Many extensions to the Standard Model (SM) such as split SUSY [1] [2], MSSM with R-parity
62 violation [3], or Hidden Valley [4] predict the production of neutral, weakly-coupled particles with long
63 lifetimes compatible with the dimension of the ATLAS detector. In particular, several theories, including
64 R-parity violation, Hidden valley, or Z' models with long-lived neutrinos [5], predict the existence of
65 LLPs that can decay to final-states containing a displaced vertex with a pair of leptons.

66 This paper presents the search for a heavy, long-lived neutral particle decaying to a dilepton pair, $\mu\mu$,
67 ee , or $e\mu$ within the ATLAS Inner Detector (ID). The LLP is referred as Z' but with no assumption
68 on Z' production mechanism for a model-independent search. For the purpose of establishing a signal
69 benchmark, the LLP is singly produced in Drell-Yan process with Z' mass ranges from 100 GeV to 1 TeV
70 and $c\tau$ between 100 mm and 500 mm. The displaced vertex provides a clean signature with minimum
71 backgrounds from the SM processes.

72 There have been several searches for the LLPs produced in pp collisions in Run I at $\sqrt{s} = 8$ TeV, including
73 the search for displaced hadronic jet [6], displaced heavy flavors [7], or multi-track displaced vertex [8],
74 and no significant excess was observed. This paper presents the search for a different signature, and it
75 is one of the first efforts¹ in the ATLAS experiment to search for a genetic displaced vertex signature
76 decaying to a dilepton pair.

77 This analysis uses 32.8 fb^{-1} of pp collision data at $\sqrt{s} = 13$ TeV collected in 2016 using the ATLAS
78 detector. In order to gain sensitivity for the non-conventional signature of LLPs, a special setup of data
79 reprocessing and reconstruction, called *Large radius tracking*, is used. This setup is described in 3.2.
80 The special setup allows the reconstruction of tracks with large impact parameters and secondary vertices
81 significantly displaced from primary vertices.

82 The analysis shares the technical setup and the analysis framework [9] with SUSY displaced multi-track
83 vertex search [10] and SUSY displaced dilepton search []. The former searches for the LLPs decaying to

¹ SUSY displaced dilepton search in Run II is looking for a displaced dilepton signature in the context of supersymmetric models.

displaced vertices with high track multiplicity and large missing energy, and the latter looks for the same displaced dilepton signature as this search but in the context of supersymmetric models.

This analysis focuses on interpreting the LLPs decaying to displaced dilepton vertices in the context of model-independent, exotic resonance search.

2 Data and MC samples

2.1 Data samples

The analysis uses the full 2016 pp collisions data (periods A-L) with the integrated luminosity of 32.8 fb^{-1} . In this search, because the standard ATLAS track reconstruction does not provide good sensitivity for long-lived particles, a dedicated stream, DRAW_RPVLL, is used to reconstruct events using the non-standard reconstruction algorithm discussed in Section 3.2.1. The stream is used in several Exotics and SUSY analyses, searching for long-lived particles.

In DRAW_RPVLL stream, a subset of events from the main physics stream is selected by RPVLL filters. The filters select events using High-Level Triggers (HLT) and offline selections configured for each analysis. The triggers and offline selection used in this search is discussed in Section 4. The selected events are passed downstream for reconstruction. The data is in RAW format so that low-level information such as detector hits can be used for the special reconstruction algorithms to reconstruct displaced tracks and vertices.

The selected events are centrally processed with AMI tag r8669. The dedicated track reconstruction algorithm, the *large radius tracking*, and the secondary vertex reconstruction algorithm, VrtSecInclusive, are used to reconstruct displaced tracks and vertices, respectively. The output of DRAW_RPVLL stream is in DAOD_RPVLL format which is a standard xAOD data format with additional displaced tracks and secondary vertices reconstructed.

The DAOD_RPVLL is further processed to produce the DAOD_SUSY15 derivation with AODfix and data reduction as recommended by the Analysis Model Study Group (AMSG) [11]. Table 1 summarizes datasets used in this search.

Format	Dataset
DRAW_RPVLL	data16_13TeV.*.physics_Main.merge.DRAW_RPVLL.f*_m*
DAOD_RPVLL	data16_13TeV.*.physics_Main.recon.DAOD_RPVLL.f*_r8669
DAOD_SUSY15	data16_13TeV.*.physics_Main.recon.DAOD_RPVLL.f*_r8669_p2950

Table 1: Dataset used in DRAW_RPVLL, DAOD_RPVLL, and DAOD_SUSY15 format.

This search uses a modified version of the standard GoodRunsList because a small number of events selected by DRAW_RPVLL was not reconstructed successfully. The corresponding lumi blocks were removed from the GoodRunsList².

² data16_13TeV.periodAllYear_DetStatus-v83-pro20-15_DQDefects-00-02-04_PHYS_StandardGRL_All_Good_25ns_DAOD_RPVLL_r8669.xml

2.2 MC samples

2.2.1 Signal samples

The long-lived Z' is generated using PYTHIA 6.4 [12] in which Z' is singly produced from $q\bar{q}$ scattering and decays to a $\mu\mu$, ee , or $e\mu$ pair. The proper lifetime, $c\tau$, is set to 100 mm, 250 mm, or 500 mm. The mass of Z' is set between 100 and 1000 GeV. A width based on relativistic Breit-Wigner is assumed for the new resonance. A sample of 20k events are generated for each mass and lifetime. Table 2 summarizes dataset identifiers (DIDs), mass, and lifetime of the signal MC samples used in this search.

$m_{Z'}$ (GeV)	Γ (GeV)	$c\tau$ (mm)	DID		
			$\mu\mu$	ee	$e\mu$
100	2.8	100	308264	309539	309554
100	2.8	250	308265	309540	309555
100	2.8	500	308266	309541	309556
250	6.9	100	301911	309542	309557
250	6.9	250	301912	309543	309558
250	6.9	500	301913	309544	309559
500	14.7	100	301914	309545	309560
500	14.7	250	301915	309546	309561
500	14.7	500	301916	309547	309562
750	23.0	100	308285	309548	309563
750	23.0	250	308286	309549	309564
750	23.0	500	308287	309550	309565
1000	31.0	100	301917	309551	309566
1000	31.0	250	301918	309552	309567
1000	31.0	500	301919	309553	309568

Table 2: Mass, lifetime, and DID of the signal MC samples.

The signal MC samples generated by PYTHIA are processed to include detector simulation using the AMI tags s2698 and s2726. The samples are overlaid with simulated minimum-bias events to model multiple interactions (pile-up) in data samples. In the signal MC samples, the average number of pile-ups, $\langle\mu\rangle$, ranges from 10 to 40 with small number of events having $\langle\mu\rangle < 10$. The difference in the $\langle\mu\rangle$ distributions between MC and data samples are corrected for by pile-up reweighting in Section 8.3. The resulting MC samples, in HTS format, are reconstructed using AMI tag r8788.

In the reconstruction process, the large radius tracking and VrtSecInclusive algorithms are used with the same configuration as data samples discussed in Section 2.1, to reconstruct displaced tracks and vertices. The reconstructed events are stored in DAOD_RPVLL, and the samples are processed to produce the DAOD_SUSY15 derivation with AODfix and data reduction.

The representative plots of truth-level p_T and η distributions of Z' and the muons from the decay of Z' , referred as *signal* muons, are shown in Figure 1 using the signal MC samples with $m = 500, 1000$ GeV and $c\tau = 100$ mm. The signal MC samples with ee and $e\mu$ final states produce similar distributions as shown in Appendix A.

The η distribution of signal muons shows that most of the signal muons are produced within the detector acceptance ($\eta < 2.7$). The characteristic upper edge in the p_T spectrum is related to the Z' mass.

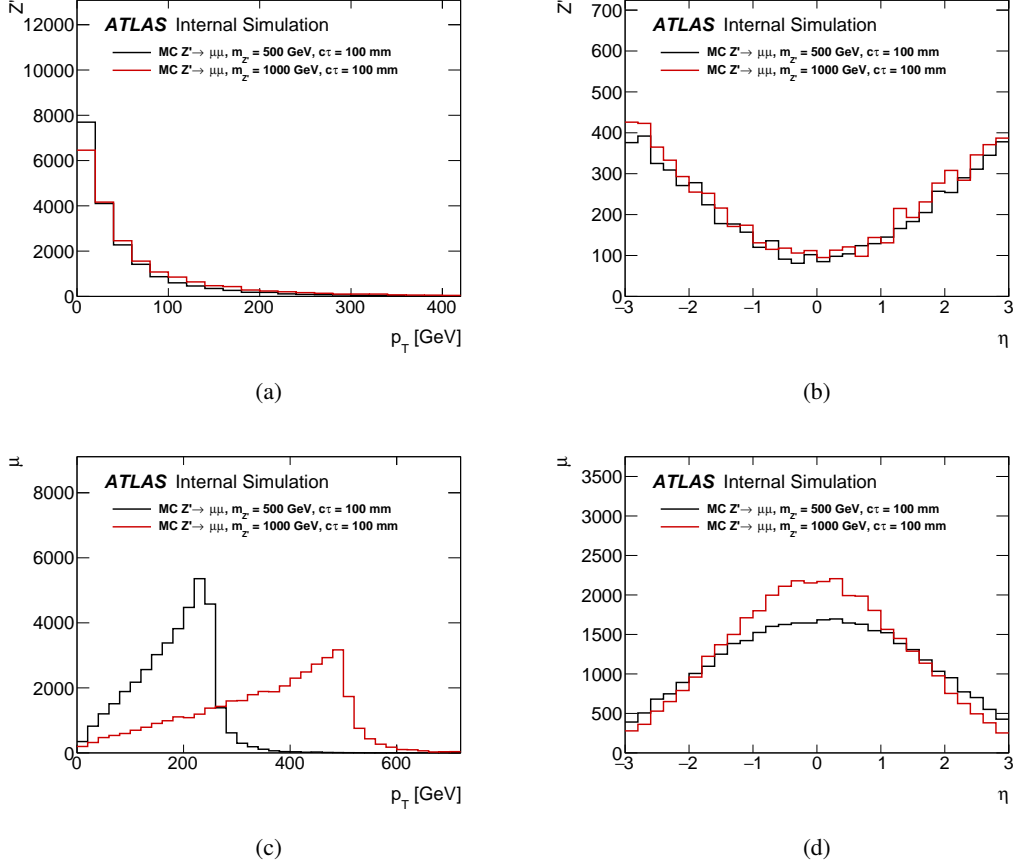


Figure 1: The representative plots of truth-level (a) p_T and (b) η distributions of Z' , and (c), (d) are the corresponding distributions for the signal muons. The signal MC samples are generated with $m = 500, 1000$ GeV, and $c\tau = 100$ mm.

2.2.2 Background MC samples

In this analysis, backgrounds are estimated from data because most of the backgrounds are expected to be originated from non-collision processes such as cosmic rays or random-crossing of tracks.

However, SM background samples are used to study the performance of random-crossing background estimation in Section 9 and to estimate the systematic uncertainties in vertexing and tracking in Section 10. The SM background samples are reprocessed from HITS using the same configuration as the signal MC sample for consistency.

The background MC samples used for background and systematic uncertainty estimations are summarized in Table 3.

Process	DID	σ (pb)	Events (10^6)	$\mathcal{L}_{Int}(\text{fb}^{-1})$
$t\bar{t}$	410252	87.8	0.70	7.97
$ZZ \rightarrow \ell\ell\ell\ell$	361063	11.7	0.12	10.3
$W^-Z \rightarrow \ell\ell\ell\nu$	361064	1.68	0.020	10.2
$W^+Z \rightarrow \ell\ell\ell\nu$	361066	2.33	0.70	30.0
$WW \rightarrow \ell\ell\nu\nu$	361068	12.8	0.025	1.95
JZ3W	361023	$8.45 \cdot 10^3$	0.20	0.0247
JZ4W	361024	135	0.20	1.48
JZ5W	361025	4.20	0.20	47.6
JZ6W	361026	$2.42 \cdot 10^{-2}$	0.20	826

Table 3: Background MC samples used in the study of random-crossing background and in the estimation of tracking and vertexing systematic uncertainty.

3 Data reprocessing

3.1 Event preselection

In the first step of data reprocessing, events are selected by DRAW_RPVLL filters as described in Section 2. The filters are designed to select events of interest while maintaining reasonably low filter rates. Single photon (γ), single electron (e), di-photon ($\gamma\gamma$), di-electron (ee), and the combination of photon and electron ($e\gamma$) filters are used to select events with ee or $e\mu$ candidates of interest. Single μ filter is used to select events with $\mu\mu$ or $e\mu$ candidates of interest.

Events are required to pass one of the HLTs listed in Table 4. In order to increase the sensitivity to electrons and muons with large transverse and longitudinal impact parameters, d_0 and z_0 , the triggers with no requirement on ID tracks are used. Consequently, muon trigger with Muon Spectrometer (MS) information only and photon triggers are used.

Description	Trigger
Single photon	HLT_g140_loose
Di-photon	HLT_2g50_loose
Single muon	HLT_mu60_0eta105_msonly

Table 4: HLTs used to select events in DRAW_RPVLL filter. Single photon trigger requires one photon with $p_T > 140$ GeV. Di-photon trigger requires two photons with $p_T > 50$ GeV. Single muon trigger requires one muon with $p_T > 60$ within $0 < |\eta| < 1.05$ using MS information only.

In addition to the HLT requirements, each filter requires offline selection on particles such as p_T , η , and d_0 . Single photon (γ) or electron (e) filter requires a leading photon or electron, respectively, with $p_T > 150$ GeV, $\eta < 2.5$, and $d_0 > 2.0$ mm. These filters also require a second photon or lepton with $p_T > 10$ GeV and $\eta < 2.5$ to keep the filter rate reasonably low. Single muon μ filter requires a muon with $p_T > 60$ GeV, $\eta < 2.5$, and $d_0 > 1.5$ mm. Di-photon ($\gamma\gamma$), di-electron (ee), and the combination of photon and electron ($e\gamma$) filters require two photons/leptons with $p_T > 50$ GeV, $\eta < 2.5$, and $d_0 > 2.0$ mm. The offline selection is summarized in Table 5.

Filter	Leading			Second		
	p_T (GeV)	$ \eta $	d_0 (mm)	p_T (GeV)	$ \eta $	d_0 (mm)
γ, e	> 150	< 2.5	> 2.0	> 10	< 2.5	-
μ	> 60	< 2.5	> 1.5	-	-	-
$\gamma\gamma, ee, e\gamma$	> 50	< 2.5	> 2.0	> 50	< 2.5	> 2.0

Table 5: RPVLL filter offline selection on photon and leptons. Single photon (γ) or electron (e) filter requires a leading photon or electron, respectively, with $p_T > 150$ GeV, $\eta < 2.5$, and $d_0 > 2.0$ mm and a second photon or lepton with $p_T > 10$ GeV, $\eta < 2.5$. Single muon μ filter requires a muon with $p_T > 60$ GeV, $\eta < 2.5$, and $d_0 > 1.5$ mm. Di-photon ($\gamma\gamma$), di-electron (ee), and the combination of photon and electron ($e\gamma$) filters require a photon or lepton with $p_T > 50$ GeV, $\eta < 2.5$, and $d_0 > 2.0$ mm.

The event selected by the RPVLL filters are passed downstream as DRAW_RPVLL for the special track and vertex reconstruction.

3.2 Reconstruction

3.2.1 Large radius tracking

The standard track reconstruction in ATLAS is optimized³ for the reconstruction of particles that are originating from the primary pp interaction point (IP). The requirements on d_0 and z_0 limit the tracking efficiency at large impact parameters ($d_0 > 2$ mm). To improve the tracking efficiency at large impact parameter, the large radius tracking algorithm is used for track reconstruction.

The large radius tracking is performed as the third tracking sequence following the standard *inside-out* and *outside-in* track reconstruction [13]. It follows the similar track reconstruction strategy as the standard *inside-out* track reconstruction, but there are a few important differences between the two tracking algorithms.

- The large radius tracking only uses un-used hits from the standard *inside-out* and *outside-in* track reconstruction for track seed creation.
- The requirements on tracks such as d_0 , z_0 , and number of hits are relaxed.

The track requirements in the standard track reconstruction and the large radius tracking are compared in Table 6. More details on the large radius tracking can be found in Ref. [14].

The collection of tracks reconstructed by the large radius tracking, referred as *large radius tracks*, is merged with the track collection from the standard track reconstruction. The combined track collection is used as an input for the lepton reconstruction and identification and secondary vertex reconstruction.

³ There are dedicated algorithms for reconstructing displaced decays such as photon conversion and b -hadron decays, but their usage is limited to the particular topologies.

	Standard	Large radius
Maximum d_0 (mm)	10	300
Maximum z_0 (mm)	250	1500
Maximum $ \eta $	2.7	5
Maximum shared silicon modules	1	2
Minimum unshared silicon hits	6	5
Minimum silicon hits	7	7
Seed extension	Combinatorial	Sequential

Table 6: Comparison of track requirements between the standard and large radius trackings.

3.2.2 Lepton reconstruction and identification

In this search, the standard ATLAS electron and muon reconstruction and identification algorithms are used. The combined track collection, including both standard and large radius tracks, is used as an input. Therefore, standard muon or electron working points such as **Tight**, **Medium**, or **Loose** are available after the lepton identification. A few changes are implemented in the muon reconstruction algorithm to improve the muon reconstruction efficiency at large impact parameters;

- requirements on d_0 and Pixel hits of muon tracks are removed.
- minimum SCT hits on muon tracks are lowered to 2.

Details of these algorithms are discussed in Refs. [15] and [16].

3.2.3 Secondary vertex reconstruction

Secondary vertices are reconstructed by `VrtSecInclusive` algorithm. The algorithm was originally developed for the material mapping of the ID in Run I, but updated for several long-lived particle searches in Run 2.

The secondary vertex reconstruction starts by selecting tracks that satisfy the requirements on track parameters and hit patterns as shown in Table 7. Both large radius tracks and standard tracks are used as input, and tracks passing the requirements are stored for the next step of the vertex reconstruction.

Variable	Cut
p_T (GeV)	> 1.0
χ^2/DOF	< 50.0
d_0 (mm)	$2.0 - 300.0$
z_0 (mm)	< 1500.0
SCT hits	≥ 2
Si shared hits	≤ 2
Pixel and TRT hits	TRT hits > 0 or Pixel hits ≥ 2

Table 7: Track requirements for secondary vertex reconstruction.

The selected tracks are used for the creation of two-track *seed* vertices. From the *seed* vertices, fake vertices are rejected by considering the location of a vertex and hit patterns of the tracks from the vertex. Tracks are not allowed to have any hits at radius smaller than the vertex position. *seed* vertices passing the location and hit patterns requirement are used to create N-track vertices. Ambiguity solving is applied to N-track vertices to improve the purity of the reconstructed vertices. More details of the algorithm can be found in Ref. [17].

The two-track secondary vertices reconstructed by `VrtSecInclusive` is the primary analysis object, and Section 4.3 discuss the vertex selection applied to these two-track vertices to select displaced vertices. Lepton identification requirements are applied to the tracks from secondary vertices after applying analysis level vertex selection, so secondary vertices reconstructed can have any combination of muon, electron, or non-leptonic tracks.

4 Signal selection

In this section, the analysis level selections applied to events, leptons, and and vertices are described.

4.1 Event selection

Minimum requirements are placed on events based on the quality of events, completeness of the corresponding luminosity blocks, primary vertex, and HLTs used in this analysis. In addition, cosmic veto is applied to reject events with back-to-back muons. The event selection is described below.

- `GoodRunsList` removes events from incomplete luminosity blocks.
- Event cleaning removes corrupted/bad events due to problems in TileCal, LAr noise bursts, SCT recovery, or TTC restarts [18].
- Events are required to pass one of the HLTs listed in Table 4.
- Events are required to have at least one primary vertex along the beam line ($z < 200$ mm).
- Events are rejected if there is a pair of leptons with $R_{\text{CR}} < 0.01$ where $R_{\text{CR}} = \sqrt{(\Delta\phi - \pi)^2 + (\Sigma\eta)^2}$.

The event selection is summarized in Table 10.

4.2 Muon and electron requirements

Prior to applying the vertex level selections, tracks from vertices are required pass the requirements on kinematics, overlap removal, and muon and electron identification criteria.

Electron requirements are based on the recommendations from EGamma group [18] with a few optimization for electrons with large impact parameters. Electrons are rejected if there is a bad cluster associated with an electron. Basic kinematic cuts are applied to electrons, $|\eta| < 2.47$ and $p_T > 7$ GeV. The electron LooseLH working point is used, but the requirements on d_0 and Pixel hits are removed to improve electron detection efficiency at large impact parameters.

Muon requirements are based on the recommendations from MuonCombinedPerformance group [19]. Muon Loose working point is used for the identification criteria, and a fiducial cut, $|\eta| < 2.5$, and kinematic cut, $p_T > 10$ GeV, are applied to muons. The requirements on Pixel hits are removed to improve muon detection efficiency at large impact parameters. In addition, muons are required to be CombinedMuons to ensure that they have associated ID tracks for vertex reconstruction. In case of MC samples, muon momentum resolution and scale correction are applied to the simulated muons for better agreement between data and simulation [16].

Overlap removal is applied to both muons and electrons to ensure that a ID track is associated with only one muon or electron. The muon and electron requirements are summarized in Table 8.

Muon	Overlap removal
	Muon Loose
	$ \eta < 2.5$
	$p_T > 10.0$ GeV
Electron	Combined Muon
	Overlap removal
	Bad cluster removal
	Electron LooseLH (no requirement related to d_0 , Pixel hits)
	$ \eta < 2.47$
	$p_T > 7.0$ GeV

Table 8: Muon and electron requirements applied at analysis level.

4.3 Vertex selection

The vertex selection is applied to two-track secondary vertices found in Section 3.2.3. Secondary vertices with minimum displacement of 2 mm from the primary vertex are selected. The selected displaced vertices are made of two tracks which can be any combination of muon, electron, and non-lepton tracks. Therefore, vertices are separated into three vertex types, control, validation, and signal regions.

In the control region, vertices are required to have two non-leptonic tracks (xx). In the validation region, vertices are required to have a muon or an electron and another non-leptonic track (μx , ex). In signal region, vertices are required to have a muon pair, an electron pair, or a muon-electron pair ($\mu\mu$, ee , $e\mu$). The control region and the validation region are used for background (Section 9) and systematic uncertainty (Section 10) estimations. The control, validation, and signal regions are summarized in Table 9.

Region	Vertex Type
Control	xx xx
Validation	μx , ex
Signal	$\mu\mu$, ee , $e\mu$

Table 9: The control, validation, and signal regions defined by the vertex type.

In all regions, vertices are required to pass a common set of vertex selections described as follows. Vertices are required to have $\chi^2/\text{DOF} < 5$ to reject poorly reconstructed vertices. A minimum transverse displacement of 2 mm from the primary vertex is required to suppress background from prompt decays.

Two tracks from a vertex are required to have opposite charges. Vertices are rejected if they are within the volume of disabled Pixel module [20]. Hadronic interaction of charged particles with detector material is a major source of backgrounds. Therefore, the vertices are rejected if they are within dense detector material [21]. The material veto is not applied to $\mu\mu$ type vertex due to low probability of muon interaction with detector material. Vertices are also required to be in the detector volume covered by the material mapping ($r < 300$ mm, $z < 300$ mm). The vertices are required to have $m > 10$ GeV to suppress backgrounds from low mass SM particles such as J/Ψ . The vertex mass is calculated by the secondary vertex reconstruction algorithm with the assumption that all tracks have pion mass. Cosmic veto is applied to vertices by requiring $R_{CR} > 0.01$. The cosmic veto is very effective in rejecting cosmic muons reconstructed as back-to-back muon vertices, and the details are discussed in Section 9.2

In addition to the common vertex selection, at least one electron or muon from the vertex is required to be matched with one of the triggers listed on Table 4 and the filters listed on Table 5 in the signal region. The vertex selection is summarized in Table 10.

Event	GoodRunsList (Section 3.1) Trigger filter (Section 3.1) Event cleaning (Section 4.1) Cosmic veto (Section 4.1) $z_{PV} < 200$ mm (Section 4.1)
Vertex	Trigger matching (signal region only) $\chi^2/\text{DOF} < 5$ $r > 2$ mm Opposite charge Disabled module veto Material veto (excluding $\mu\mu$) $m > 10$ GeV $R_{CR} > 0.01$ $r < 300$ mm, $z < 300$ mm Filter matching (signal region only)

Table 10: Event and vertex selections applied to select displaced vertices.

5 Signal efficiency

The signal efficiency of finding displaced dilepton vertex is defined by the ratio of the number of events passing the signal selection (Section 4) to the total number of events processed. The signal efficiency can be written as Eq. 1.

$$\mathcal{E}_{\text{overall}} = \mathcal{E}_{\text{filter}} \cdot \mathcal{E}_{\text{trigger}} \cdot (\mathcal{E}_{\text{tracking}} \cdot \mathcal{E}_{\text{leptonID}})^2 \cdot (\mathcal{E}_{\text{vertexTrack}})^2 \cdot \mathcal{E}_{\text{vertexFit}}. \quad (1)$$

$\mathcal{E}_{\text{filter}}$ and $\mathcal{E}_{\text{trigger}}$ together represent the efficiency of RPVLL filter, the ratio of the events passing RPVLL filter to the total events processed. RPVLL filter has the trigger filter as one of its requirements, and because it is desirable to study the trigger efficiency independently from the filter efficiency, RPVLL filter efficiency is factorized into the filter efficiency and the trigger efficiency. $\mathcal{E}_{\text{tracking}}$ represents the efficiency

to reconstruct ID tracks from signal particles, and $\varepsilon_{\text{leptonID}}$ represents the efficiency to identify the signal particles with reconstructed ID tracks as leptons. $\varepsilon_{\text{vertexTrack}}$ represents the efficiency for the reconstructed signal leptons to be selected for secondary vertex reconstruction, and $\varepsilon_{\text{vertexFit}}$ represents the efficiency to reconstruct a displaced vertex using two signal leptons and pass the vertex selection.

In order to understand the source of signal efficiency loss, the trigger efficiency is studied in Section 6, and the tracking and lepton identification efficiencies are studied in Section 7.

6 Trigger efficiency

The trigger efficiency is defined as the ratio of the events passing one of the triggers used in this analysis to the total events processed. No RPVLL filter is applied when estimating the trigger efficiency.

The analysis uses three triggers listed on Table 4 to select the events with displaced dilepton vertex candidates. The single muon trigger is sensitive to the events with a $\mu\mu$ or $e\mu$ vertex. The di-photon trigger is mainly used select the events with an ee vertex, but a small number of events with an $e\mu$ vertex pass this trigger. The single photon trigger is sensitive to the events with ee or $e\mu$ vertex, but its efficiency is relatively low in comparison with the other two triggers.

Figure 2 shows the efficiency of each trigger and the combined trigger efficiency on the signal MC samples of Z' decaying to all three channels at $m = 250$ GeV and $c\tau = 250$ mm. The sample with ee channel shows the highest combined trigger efficiency due to the high efficiency in di-photon trigger, and the sample with $e\mu$ channel shows the reduced combined trigger efficiency because $e\mu$ vertices have only one track that can satisfy either the single muon or photon trigger.

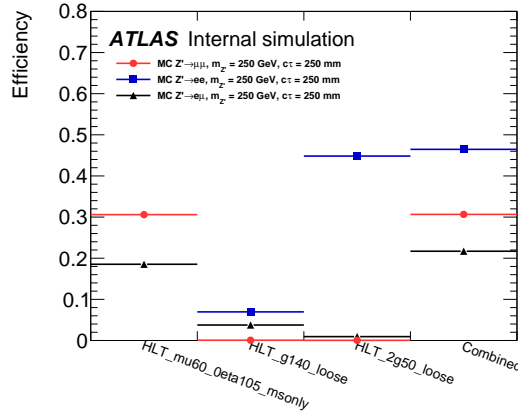


Figure 2: Trigger efficiency of single muon, single photon, di-photon, and the combined triggers of the signal MC samples of $Z' \rightarrow \mu\mu, ee$, and $e\mu$ generated with $m = 250$ GeV and $c\tau = 250$ mm.

The trigger efficiency on all $\mu\mu$ signal MC samples is shown in Table 11. It is evident that at low Z' mass (~ 100 GeV), the combined trigger efficiency on the signal MC sample is significantly reduced because the typical p_T of the signal muons is lower than the p_T threshold of the single muon trigger.

The trigger study indicates that there is a substantial loss in the signal efficiency at trigger level before reconstruction, and developing dedicated, more efficient triggers for long-lived particles will provide

potential improvement in sensitivity to long-lived particles. The systematic uncertainties in trigger efficiency is estimated by tag-and-probe method in Section 10.3.

$m_{Z'}$ (GeV)	$c\tau$ (mm)	Single muon	Single photon	Di-photon	Combined
100	100	0.047	< 0.001	0	0.047
100	250	0.043	0	0	0.043
100	500	0.039	0	0	0.039
250	100	0.343	< 0.001	< 0.001	0.344
250	250	0.306	< 0.001	< 0.001	0.307
250	500	0.230	< 0.001	< 0.001	0.230
500	100	0.454	0.010	< 0.001	0.459
500	250	0.410	0.009	< 0.001	0.415
500	500	0.331	0.008	0.001	0.336
750	100	0.541	0.026	0.002	0.553
750	250	0.470	0.023	0.003	0.481
750	500	0.391	0.022	0.001	0.402
1000	100	0.570	0.039	0.004	0.586
1000	250	0.512	0.036	0.003	0.526
1000	500	0.430	0.034	0.004	0.444

Table 11: Trigger efficiency of single muon, single photon, di-photon triggers, and the combined trigger efficiency on the signal MC samples of $Z' \rightarrow \mu\mu$.

7 Lepton reconstruction efficiency

The tracking efficiency, $\varepsilon_{\text{track}}$, and the lepton identification efficiency, $\varepsilon_{\text{leptonID}}$, are studied together as a lepton reconstruction efficiency. The lepton reconstruction efficiency is defined and estimated as follows. From a signal MC sample, the leptons decaying from Z' are collected at truth-level, referred as *truth* signal leptons. For each truth signal lepton, if there is a reconstructed lepton with its ID track matched to the ID track of the truth signal lepton by a hit-based truth matching scheme, it is marked as reconstructed. The ratio of reconstructed signal leptons to the total number of signal leptons produced in the sample is taken as the lepton reconstruction efficiency. No RPVLL or trigger filter is applied in estimating the lepton reconstruction efficiency.

Figure 3 shows the representative plot of the lepton reconstruction efficiency as a function of track parameters using the combined signal MC samples of Z' decaying to all three channels, generated with $m = 250$ GeV and $c\tau = 250$ mm.

It is evident that the efficiency drops drastically at $\eta > 2.0$ where the Pixel barrel region ends due to the minimum silicon hits requirement on tracks as shown in Table 6. The efficiency is not very sensitive to p_T except low p_T region ($p_T < 20$).

The lepton reconstruction efficiency decreases for large d_0 and z_0 . In the signal MC samples, most of Z' decay within the Pixel barrel region, $r < 122.5$ mm and $z < 400.5$ mm, where the lepton reconstruction efficiency is high.

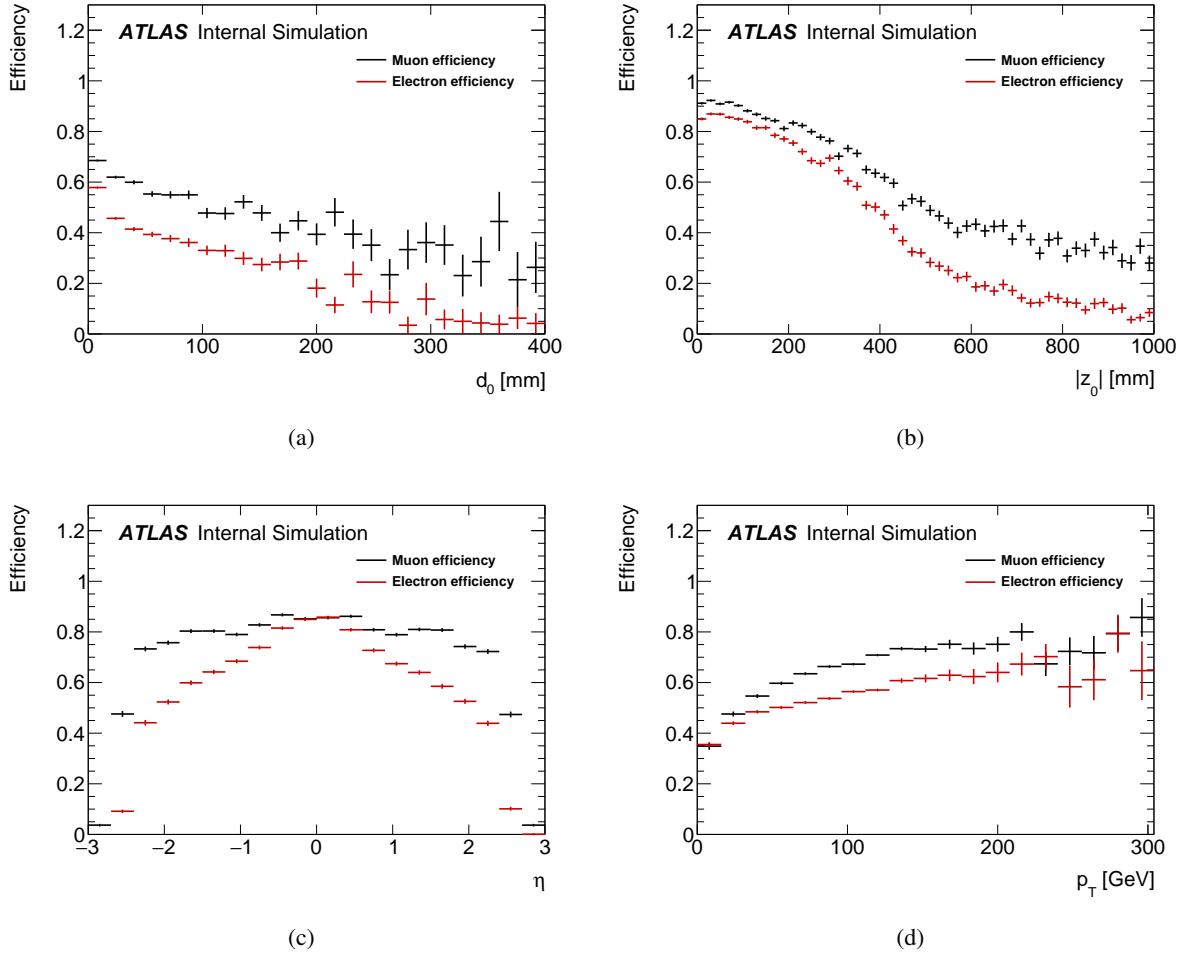


Figure 3: Lepton reconstruction efficiency as a function of (a) d_0 , (b) z_0 , (c) η , and (d) p_T of signal leptons using the signal MC sample generated with $m = 250$ GeV and $c\tau = 250$ mm.

8 Overall reconstruction efficiency

The overall reconstruction efficiency is defined as the ratio of Z 's reconstructed as displaced vertices in the signal region to the total Z ' produced in the sample. In this section, representative plots of event cut flow, vertex cut flow (Section 8.1), and overall reconstruction efficiency distributions (Section 8.2) are presented using the signal MC samples generated with $m = 500, 1000$ GeV, and $c\tau = 100$ mm.

In Section 8.3, the overall reconstruction efficiency is reweighted to reproduce pile-up distribution in the data sample. The reconstruction efficiency for all MC samples is also presented.

8.1 Event and vertex cut flow

The MC samples are processed with steps described in Section 3.1. The analysis level cuts, the event and the vertex selections described in Table 10, are applied to the processed samples. Representative plots

of the event cut flow and the vertex cut flow are shown in Figure 4 using the signal MC samples of Z' decaying to $\mu\mu$ generated with $m = 500, 1000$ GeV for $c\tau = 100$ mm.

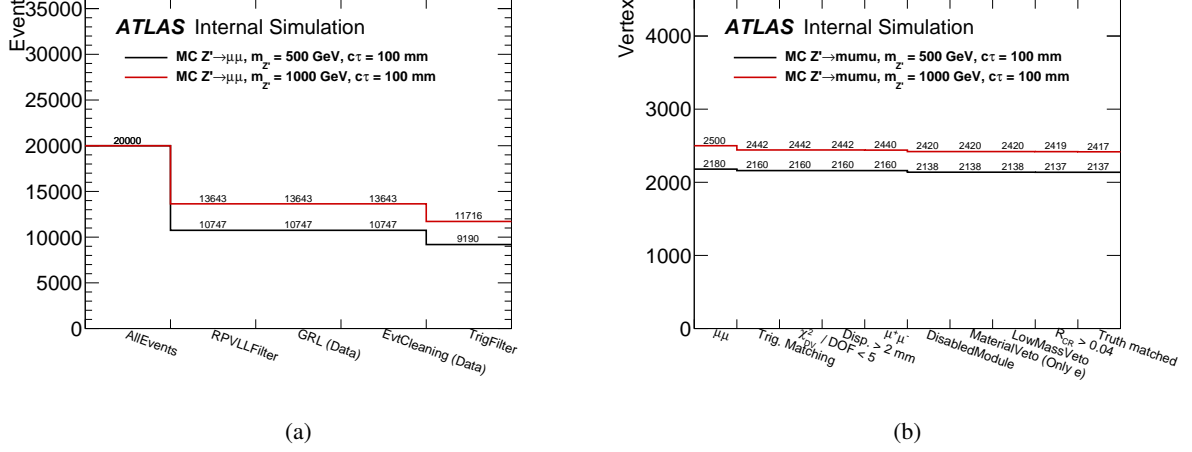


Figure 4: (a) Event cut flow, and (b) vertex cut flow using the signal MC samples of $Z' \rightarrow \mu\mu$ generated with $m = 500, 1000$ GeV for $c\tau = 100$ mm.

In the event cut flow, RPVLL filter is applied during the sample processing. GoodRunsList filter and Event cleaning are shown as place holders as they are only applied to data sample.

In the vertex cut flow, displaced vertices are reconstructed in about $\sim 25\%$ of the signal events passing the event selection, indicating that there is a significant loss of signal efficiency in the reconstruction process. The following selection criteria, $\chi^2 / \text{DOF} < 5$ and the minimum displacement cut ($r > 2$ mm), are applied, but the effect is expected to be very small as the same requirements are applied in the secondary vertex reconstruction algorithm. Material veto is applied to all vertex types except $\mu\mu$ vertex. The minimum dilepton mass requirement and cosmic veto cuts have minimum impact on the signal efficiency.

8.2 Efficiency distribution

The overall reconstruction efficiency is studied by examining the efficiency distributions in the transverse (r), longitudinal (z) vertex position, and the angular distributions of the reconstructed vertices. The representative efficiency distributions are shown in Figure 5 using the signal MC samples generated with $m = 500, 1000$ GeV for $c\tau = 100$ mm.

The overall efficiency shows a significant dependence on vertex position which decreases at large r and z due to the minimum silicon hits requirement on tracks. The first bins in r and z distributions have lower efficiency due to the minimum displacement requirement ($r > 2$ mm) on secondary vertices. The r distribution shows features that reflects the physical structure of the ID. The η distribution has higher efficiency in the central region, and the ϕ distribution is uniform as expected.

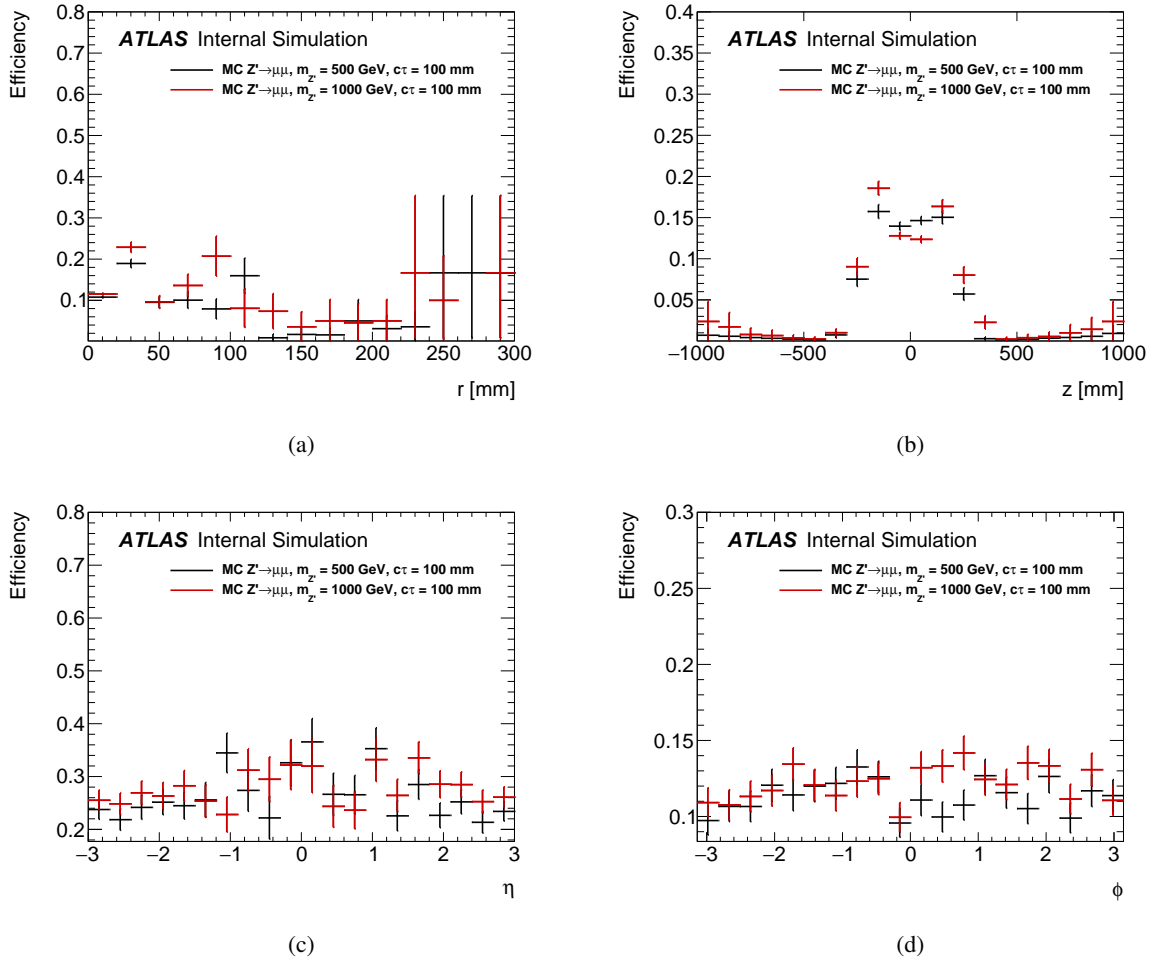


Figure 5: The overall efficiency distributions in (a) r , (b) z , (c) η , and (d) ϕ of the signal MC samples of $Z' \rightarrow \mu\mu$ with $m = 500, 100$ GeV for $c\tau = 100$ mm.

8.3 Efficiency reweighting

9 Background estimation

Due to the lifetime ($c\tau > 2$ mm) and mass ($m > 10$ GeV) requirements applied at vertex selection, no SM background is expected in the signal region in search for displaced dilepton resonance. Therefore, two non-collision backgrounds and background from low-mass vertices are considered in this search. In Section 9.1, background from *random-crossing* of two uncorrelated tracks are estimated. The cosmic ray background is estimated in Section 9.2, and the low-mass background which results from low-mass vertices is estimated in Section 9.3

9.1 Random-crossing background

The random-crossing of two uncorrelated tracks is a major source of the backgrounds in the search for displaced dilepton vertices. This background is expected to increase with more pile up in Run 2.

This random-crossing background is estimated by the *track flipping* method in which secondary vertex reconstruction is performed on each pair of tracks from all possible combinations of tracks after one random track from each pair is flipped with respect to the primary vertex. Because one track is flipped in each pair of tracks, the resulting vertices provide good estimation for random-crossing background.

The track flipping method is tested on the background MC samples. As an additional check, the vertices found from this method are compared with vertices found from another random-crossing background estimation method, the *event mixing*. In Section 9.1.2, the random-crossing background are estimated using the track flipping method with 32.8 fb^{-1} of 2016 data sample, and its systematic uncertainties are estimated in Section 9.1.3.

9.1.1 MC study

The track flipping method is tested using the background MC sample with 2.4 M events as described in Section 2. Events are selected by the same requirement described in Section 5. From the selected events, tracks identified as muon, electron, or neither, referred as muon, electron, or non-leptonic track, respectively, are selected with the track criteria (Table 7) as in the secondary vertexing algorithm for consistency. Leptons are required to pass the same selection criteria described in Table 8. Non-leptonic tracks are required to pass the minimal kinematic selection ($p_T > 10 \text{ GeV}$, $\eta < 2.5$) to match with the kinematic selection for leptons.

Track pairs are created from all possible combination of muon, electron, or non-leptonic tracks which fall into one of the six categories, $\mu\mu$, ee , $e\mu$, ex , μx , or xx track pair, where x represents a non-leptonic track. For each pair of tracks, one track is randomly flipped with respect to the beam spot ($d_0 \rightarrow -d_0$, $z_0 \rightarrow -z_0$, $\phi \rightarrow \phi - \pi$, $\theta \rightarrow \pi - \theta$). The flipped track and the other track in the pair are used to estimate the background from uncorrelated tracks.

The same secondary vertex algorithm used in the reconstruction of data or MC sample is performed on the track pairs with one track flipped to reconstruct secondary vertices. Vertex selection cuts similar to the cuts listed in Table 7 are applied to the vertices found by the track flipping. The only differences in the vertex cuts are:

- Trigger matching is only required for $\mu\mu$, ee , and $e\mu$ vertices because non-leptonic tracks cannot be matched to lepton triggers,
- Filter matching is only required for $\mu\mu$, ee , and $e\mu$ vertices for the same reason.

Because trigger and filter matchings are not required in the control and validation region, the track flipping method provides conservative background estimation. Figure 6 shows the vertex cut flow applied on xx vertices from the background MC samples.

The resulting vertices, referred as track-flipping vertices, are compared with the vertices reconstructed by the reconstruction described in Section 3.2 in the MC samples. The same event selection and the vertex selection described in Section 5 are applied to the vertices in the data sample. Because of minimum

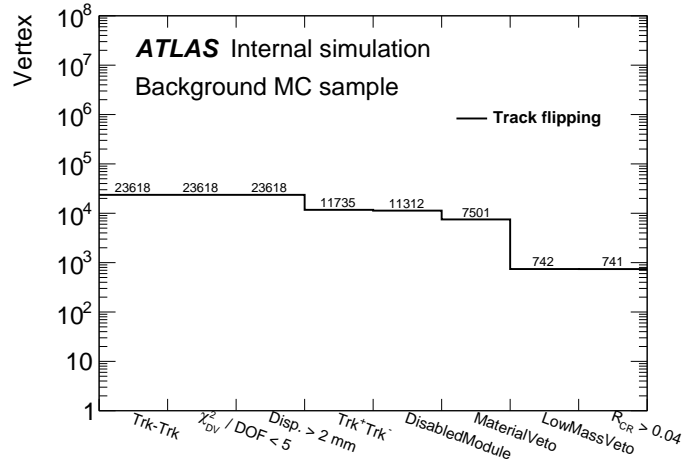


Figure 6: Vertex cut flow applied on xx vertices from the track flipping method

invariant mass ($m > 10$ GeV), minimum displacement ($r_{DV} > 2$ mm), and cosmic veto ($R_{CR} > 0.01$) cuts, the reconstructed vertices in the MC sample are expected to be purely from random-crossing.

In addition, the track-flipping are compared with another random-crossing background estimation method, the *event mixing*. In this method, tracks are sampled from data or MC sample, and the probability, denoted by p_{rc} , for a pair of tracks to form a vertex by random-crossing is estimated by randomly mixing tracks from different events. Using p_{rc} and the total number of pairs of tracks in data or MC sample, random-crossing background is estimated. The details on this method can be found in []. The same event, track, and vertex selections are applied for consistency with the track flipping.

The xx vertices found from the track flipping, reconstructed in the MC sample, and the estimation from the *event mixing* are compared in Table 12. No $\mu\mu$, ee , or $e\mu$ vertices was found or expected from these methods.

Vertex Type	Track flipping	<i>Event Mixing</i>	Background MC Samples
μx	1	1.3	0
$e x$	0	0.3	0
xx	741	714.0	676

Table 12: Comparison of the number of xx vertices found in the track flipping, estimated from the *event mixing*, and reconstructed in the background MC samples.

The xx vertex yields from the track flipping and the *event mixing* methods agree within the statistical uncertainty. The xx vertex distributions of the vertices from the track flipping, *event mixing*, and from reconstruction are shown in Figure 7. The distributions of the track flipping and the *event mixing* agree reasonably well with the distributions of vertices found from the reconstruction.

9.1.2 Estimating random-crossing background with data sample

Random-crossing background is estimated using 32.8 fb^{-1} of 2016 data sample described in Section 2.

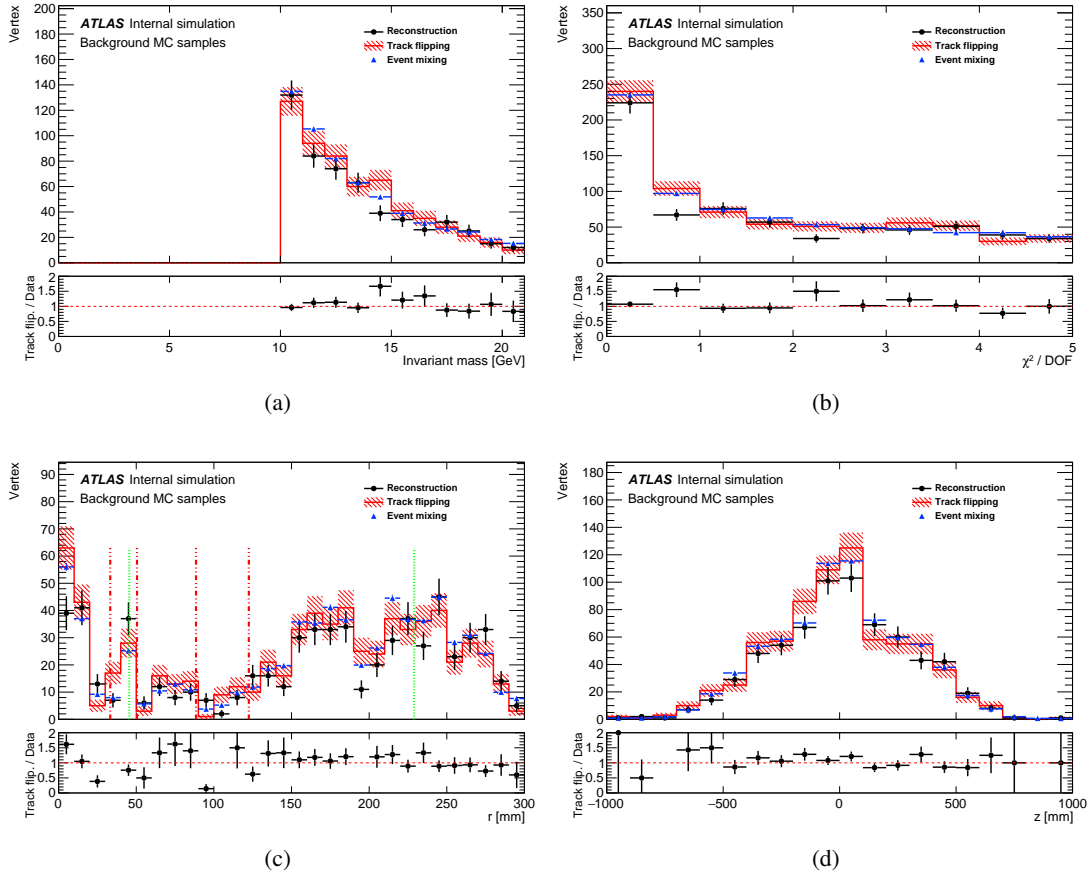


Figure 7: Comparison of (a) vertex mass, (b) χ^2/DOF , (c) transverse, and (d) longitudinal position of xx vertices reconstructed, found from the track flipping, and estimated using the *event mixing*. In (c), the red dashed lines indicate the four Pixel layers and the first layer of SCT. The green dotted lines indicate the Inner Support Tube (45.5 mm) and Pixel Support Tube (229 mm).

Vertex distribution In control region, the xx vertices found from the track flipping method are compared to the vertices reconstructed in the data sample in Figure 8. The track flipping method reproduces the data reasonably well including some of the structures.

Because of the limited statistics in lepton pairs in the data sample, track-flipping vertex yields in the signal region cannot be directly used as random-crossing background estimation. Instead, track-flipping vertex yields in the control region and validation region are extrapolated to estimate random-crossing background in the signal region using lepton probability, defined as follows:

- $P(e)$ is defined as the ratio of electrons to inner detector tracks in an entire sample,
- $P(\mu)$ is defined as the ratio of muons to inner detector tracks in an entire sample,

where track requirements described in Section 9.1.1 is required to both leptons and inner detector tracks.

Extrapolation from control region The track-flipping vertex yields in the control region is extrapolated to the signal and validation regions to estimate vertex yields using Eq. 3. The estimated μx and ex vertex yields are compared with the observed track-flipping vertex yield in validation region to calculate scale

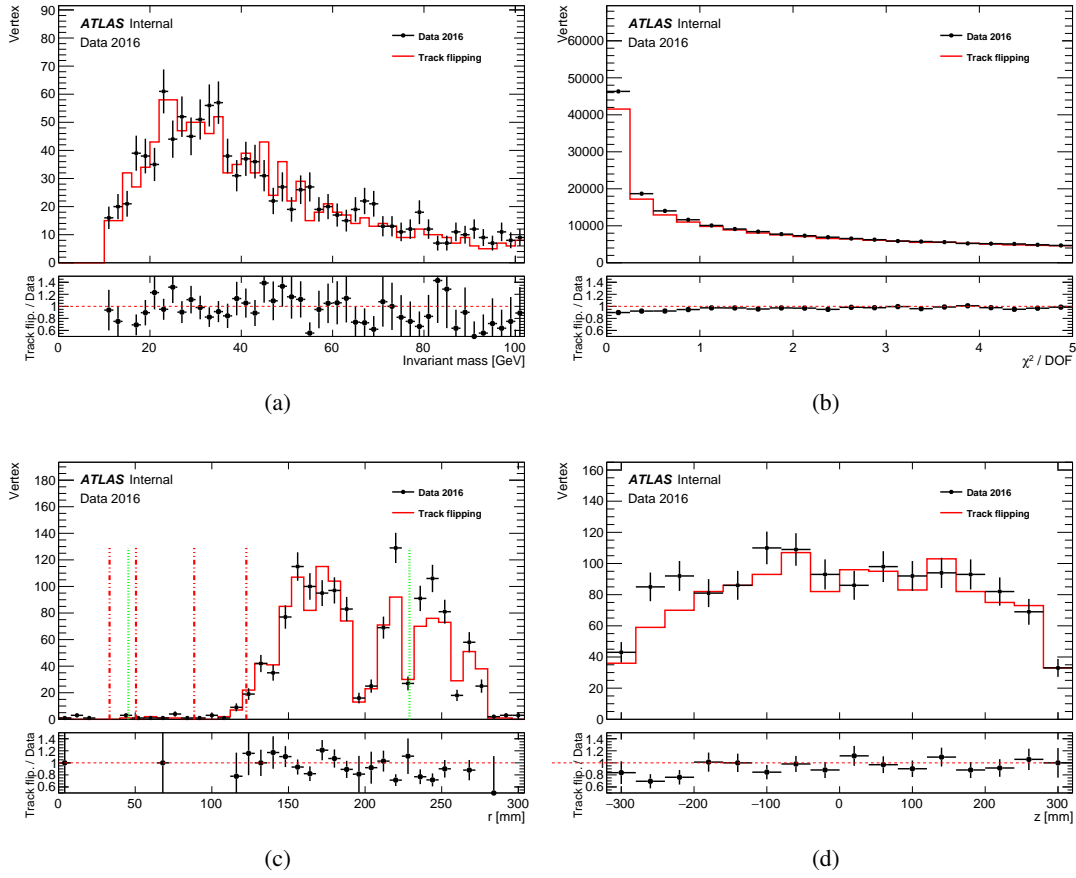


Figure 8: Comparison of (a) vertex mass, (b) χ^2/DOF , (c) transverse, and (d) longitudinal position of vertex found in 32.8 fb^{-1} of 2016 data sample with those found from the track flipping in the control region of the data sample. In (c), the red dashed lines indicate the four Pixel layers and the first layer of SCT. The green dotted lines indicate the Inner Support Tube (45.5 mm) and Pixel Support Tube (229 mm).

factors. The scale factor estimated from the extrapolation $xx \rightarrow \mu x$ ($xx \rightarrow ex$) is 0.82 (0.19). The estimated scale factors are applied to the extrapolation to the signal region using Eq. 5 to obtain the random-crossing background estimation.

Extrapolation from validation region Similarly, the track-flipping vertex yields in the validation region is extrapolated to the signal to estimate vertex yields using Eq. 4. The scale factors are applied to the extrapolation to obtain the background estimation.

The lepton probability, track-flipping and reconstructed vertex yields, scale factors, and estimated random-crossing background are summarized in Table 13.

9.1.3 Systematic uncertainty in the track flipping method

The systematic uncertainty in the track flipping method is estimated by studying the variance in background estimation from different variations of the track flipping method. In addition to the transformation

Tracks			P(ℓ)		
x	2.47×10^7	-			
μ	5.23×10^4	2.10×10^{-3}			
e	3.63×10^4	1.46×10^{-3}			
Sum	2.48×10^7	-			
(a) Lepton probability					

Tracks-flipping			Data		
xx	1255	1346			
μx	3	4			
ex	1	0			
(b) Vertex yields					

Type		SF	
$S_{xx \rightarrow \mu x}$			0.82
$S_{xx \rightarrow ex}$			0.19
(c) Scale factors			

Estimation			Applying SF		
$N_{\mu x}$	4	-			
N_{ex}	5	-			
$N_{\mu\mu}$	2.69×10^{-3}	1.79×10^{-3}			
N_{ee}	5.56×10^{-3}	1.99×10^{-4}			
$N_{e\mu}$	7.73×10^{-3}	1.96×10^{-3}			
(d) Extrapolation from control region					

Estimation			Applying SF		
$N_{\mu\mu}$	2.19×10^{-3}	1.79×10^{-3}			
N_{ee}	1.05×10^{-3}	1.99×10^{-4}			
$N_{e\mu}$	3.89×10^{-3}	1.96×10^{-3}			
(e) Extrapolation from validation region					

Table 13: Random-crossing background by track flipping

described in the previous section ($d_0 \rightarrow -d_0, z_0 \rightarrow -z_0, \phi \rightarrow \phi - \pi, \theta \rightarrow \pi - \theta$), the following track transformations are considered.

- **Same sign d0:** $d_0 \rightarrow d_0, z_0 \rightarrow -z_0, \phi \rightarrow \phi - \pi, \theta \rightarrow \pi - \theta$
- **Same sign θ :** $d_0 \rightarrow -d_0, z_0 \rightarrow -z_0, \phi \rightarrow \phi - \pi, \theta \rightarrow \theta$
- **90 degrees rotation in ϕ :** $d_0 \rightarrow -d_0, z_0 \rightarrow -z_0, \phi \rightarrow \phi - \frac{\pi}{2}, \theta \rightarrow \theta$

NEED TO ADD RESULTS HERE

9.2 Cosmic background

Cosmic background is the dominant source of backgrounds in $\mu\mu$ channel as a cosmic muon can be reconstructed as a back-to-back $\mu\mu$ vertex with large displacement from the primary vertex. Cosmic veto is introduced to suppress the background from such vertices by requiring $R_{CR} = \sqrt{(\Delta\phi - \pi)^2 + (\Sigma\eta)^2} > 0.04$ where $\Delta\phi$, $\Sigma\eta$ are the difference and sum of track parameters from two tracks at a vertex, respectively. For a back-to-back vertex, R_{CR} is expected to be close to 0. The vertex cut flow in Figure 4 shows that the efficiency loss due to the cosmic veto is negligible.

In order to estimate the cosmic background, *cosmic control region* is defined by inverting the cosmic veto cut (i.e. $R_{CR} < 0.04$). All other event and vertex selections are kept the same as the signal region. Figure 9(a) shows the vertex cut flow of the cosmic control region in the data sample. There are 127 $\mu\mu$ vertices found with $R_{CR} < 0.04$, and no ee or $e\mu$ vertex was found. Figure 9(b) shows the R_{CR} distribution of these vertices. All vertices found in the cosmic control region have very small R_{CR} . The R_{CR} distribution is fitted with gaussian function and then extrapolated into the signal region, and the cosmic background is estimated to be negligible.

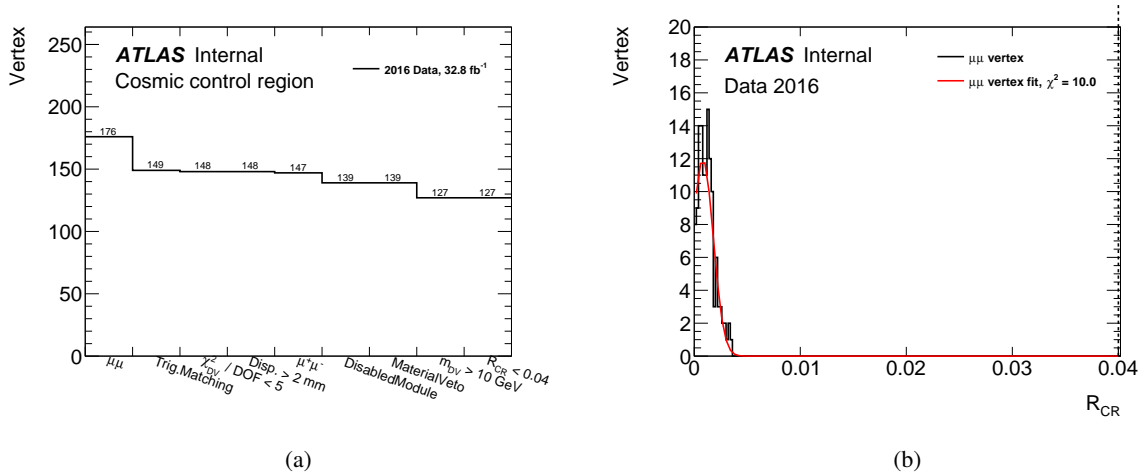


Figure 9: (a) Vertex cut flow of the cosmic control region, and (b) R_{CR} distribution of $\mu\mu$ vertex found in the data sample. The curve shows a fit with an exponential function.

9.3 Low-mass background

A SM process such as photon conversion or the decay of a low-mass particle, e.g. J/ψ , can be reconstructed as a low-mass, displaced vertex decaying to a dilepton final state. Low-mass veto is implemented to suppress background from these displaced vertices ($m > 10$ GeV). Due to large mass of Z' in the signal MC samples, there is no efficiency loss by the veto.

Low-mass control region is used to estimate the low-mass background by inverting the low-mass cut (i.e. $m < 10$ GeV). All other event and vertex selections are kept the same as the signal region.

Figure 10 shows the vertex cut flow and mass distribution of the vertices found in the low-mass control region in the data. There are 12 vertices with $m < 10$ GeV, passing all other signal selections. The mass spectrum is fitted with the sum of a Gaussian centered around 3 GeV, representing J/ψ resonance, and an

464 exponential for the background distribution. The extrapolation of the combined fit yields an estimate of
 465 ~ 0.09 vertices in the signal region.

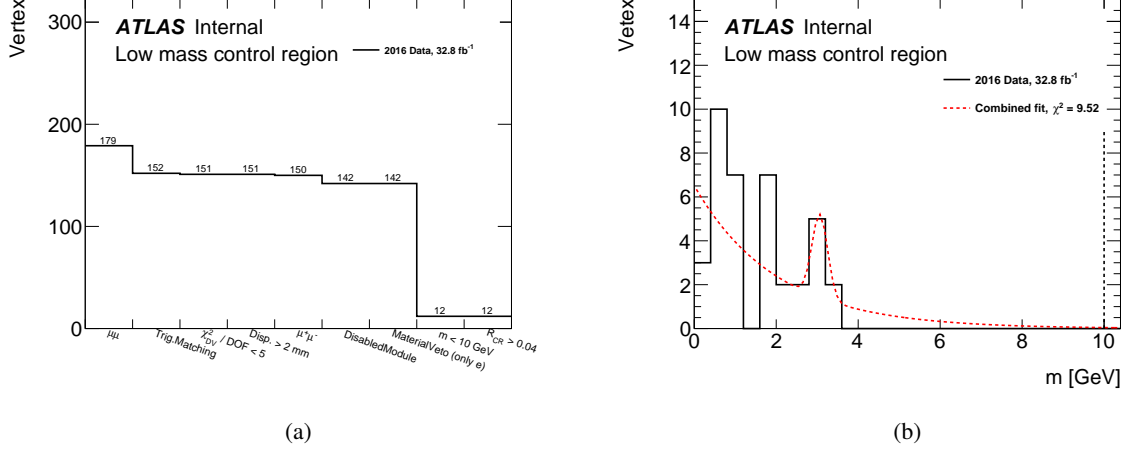


Figure 10: (a) Vertex cut flow and (b) mass distribution of the low-mass control region from the data sample. The combined fit of a Gaussian centered around 3 GeV, representing J/ψ resonance, and an exponential for the background is used to estimate the low-mass background in the signal region.

466 10 Systematic uncertainties

467 The systematic uncertainty in the signal efficiency is estimated by estimating the systematic uncertainty
 468 of each efficiency factor in Eq. 1.

469 The systematic uncertainties on $\varepsilon_{\text{track}}$, $\varepsilon_{\text{vertexTrack}}$, and $\varepsilon_{\text{vertexFit}}$ are studied using K_S in Section 10.1. The
 470 systematic uncertainties on $\varepsilon_{\text{lepton}}$ is studied in Section 10.2 using tag-and-probe method with $Z \rightarrow ee, \mu\mu$
 471 events. The systematic uncertainties on $\varepsilon_{\text{trigger}}$ and $\varepsilon_{\text{filter}}$ will be studied separately.

472 10.1 Systematic uncertainties on tracking and vertexing efficiency

473 10.1.1 Data and MC comparison

474 The systematic uncertainties on $\varepsilon_{\text{track}}$, $\varepsilon_{\text{vertexTrack}}$, and $\varepsilon_{\text{vertexFit}}$ are estimated by comparing the tracking
 475 and vertexing efficiencies between the data and the background MC samples described in Section 2.1
 476 using the process, $K_S \rightarrow \pi^+\pi^-$. In order to understand the validity and the limitation of this method, the
 477 kinematic distributions of K_S and Z' found in this method are compared in Appendix C.

478 Events are selected using the same event selection described in Section 4. From the selected events, K_S
 479 candidates, referred as K_S vertices, are selected by applying K_S vertex selection to the secondary vertices
 480 in the events. K_S vertex selection is similar to the Z' signal vertex selection, but for the consistency with
 481 K_S study in Run I and further background reduction, additional vertex cuts as described in Ref. [22] are
 482 applied in K_S vertex selection. The mass window of 0.35 to 0.65 GeV is used in the K_S vertex selection.

The difference between K_S and Z' vertex selections are summarized in Table 14. Figure 11 shows K_S vertex cut flow in the data and the background MC samples.

	Z'	K_S
Vertex type	$\mu\mu, e\mu, ee$	xx
Mass (GeV)	> 10.0	$[0.35, 0.65]$
Additional cut	-	K_S selection [22]

Table 14: Comparison of Z' and K_S vertex selections.

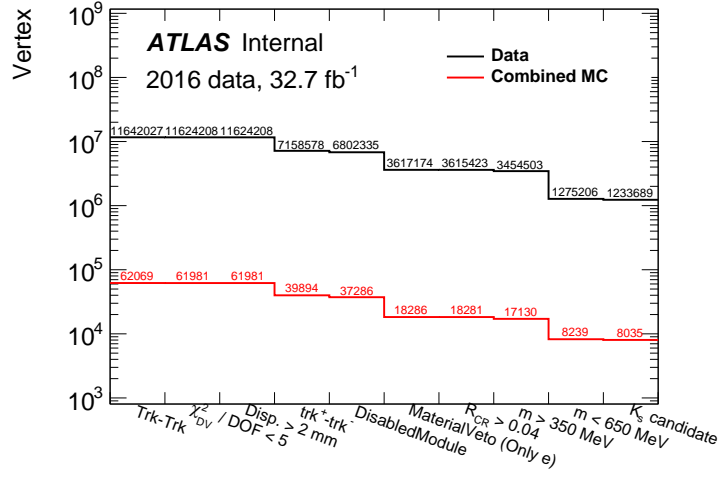


Figure 11: Vertex cut flow applied on K_S vertices in the data and MC samples

After applying the event and K_S vertex selection, the K_S vertex distributions in the data are compared to the MC samples in Figure 12. The data sample is normalized to the MC sample which has limited statistics. There are good agreements in the invariant mass, p_T , transverse, longitudinal position, and decay length of the vertices, except the pile-up distribution as expected.

K_S vertices found in the data and MC samples are binned in decay radius, r , and the K_S yields in each bin are estimated from a fit using Breit-wigner. Figure 13 shows a few representative K_S mass distributions (others are included in Appendix B). Background is negligible and hence not included in the fit. The estimated K_S yields are normalized to the number of K_S found in the lowest r bin since the expected number of K_S in the data samples is unknown.

The K_S yield, normalized to the yield with $r < 20$ mm, is compared to the MC in Figure 14. The lower pane shows the double ratio,

$$\frac{N_R^{\text{data}}}{N_{R_0}^{\text{data}}} \bigg/ \frac{N_R^{\text{MC}}}{N_{R_0}^{\text{MC}}} \quad (2)$$

which shows deviation from unity with increase of r . The difference in yield is then convoluted with the r distribution of Z' to estimate the systematic uncertainty in tracking and vertexing.

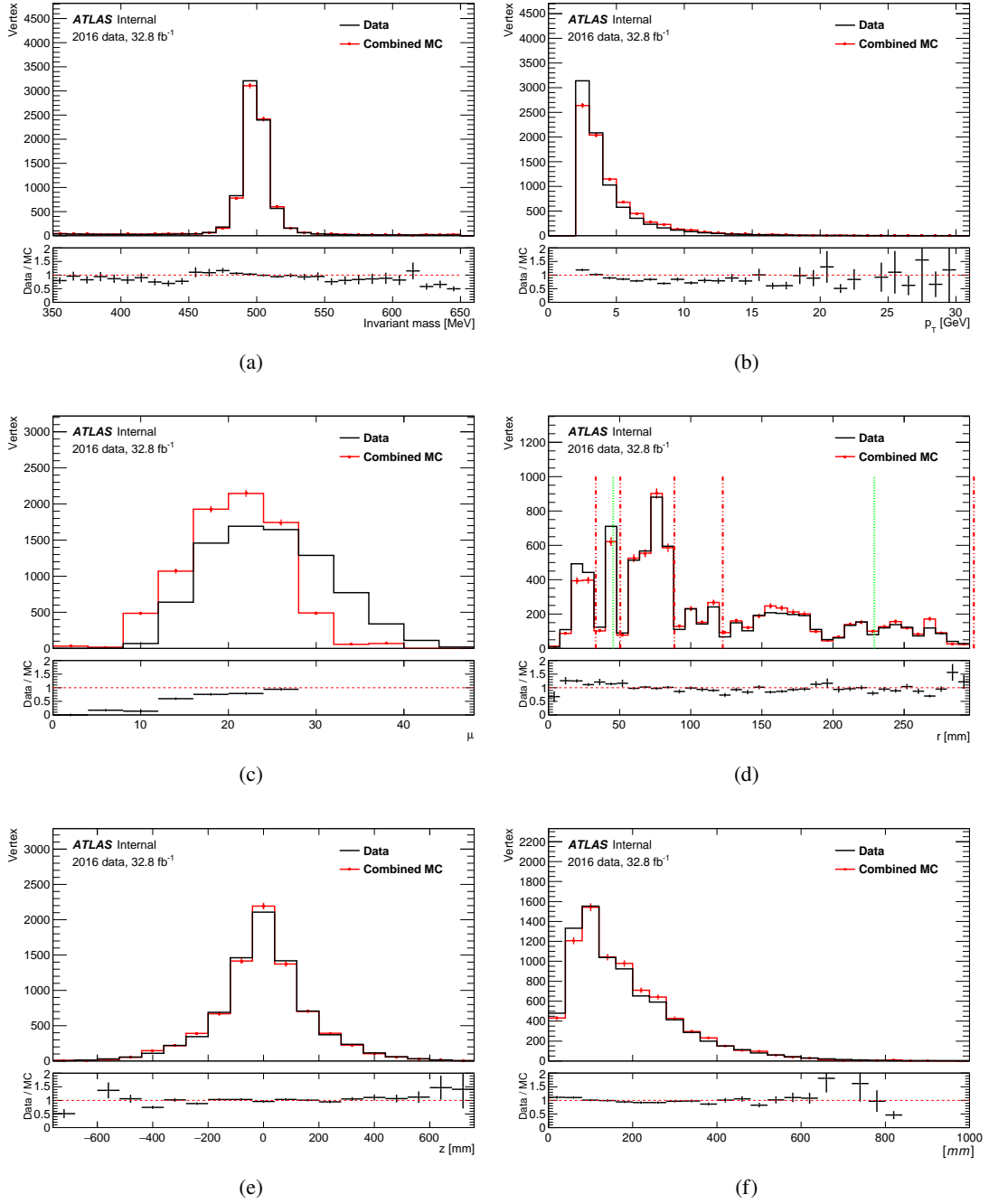


Figure 12: Comparison of the (a) invariant mass, (b) p_T , (c) μ , (d) transverse, (e) longitudinal position, and (f) decay length of K_S in the data with the MC samples. Data is normalized to MC. In (d), the red dashed lines indicate the four Pixel layers and the first layer of SCT. The green dotted lines indicate the Inner Support Tube (45.5 mm) and Pixel Support Tube (229 mm).

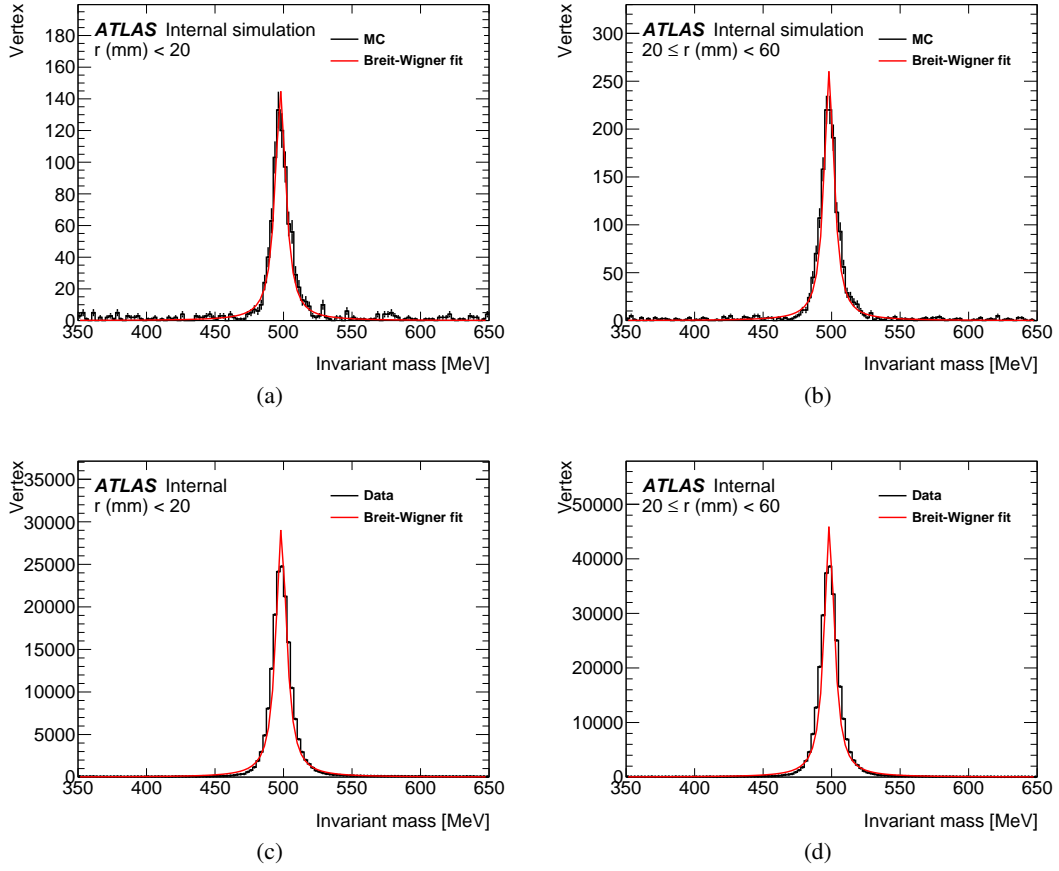


Figure 13: Representative distributions of the K_S invariant mass for (a) $r < 20$ mm, and (b) $20 < r < 60$ mm in the background MC sample. The corresponding plots from the data sample are shown in (c) and (d). The mass distributions are fitted with a Breit-Wigner.

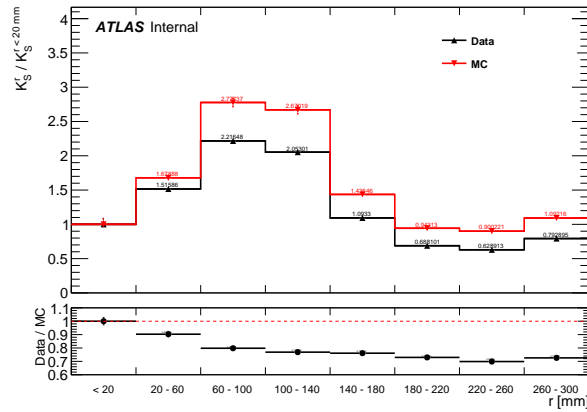


Figure 14: The radial distribution of K_S yield, normalized to the lowest r bin in the data and MC samples. The lower pane shows the double ratio as defined in the text.

498 **10.2 Systematic uncertainties on lepton identification**

499 **10.3 Systematic uncertainties on trigger efficiency**

500 **11 Results**

501 **12 Conclusion**

References

- [1] J. L. Hewett, B. Lillie, M. Masip, and T. G. Rizzo,
Signatures of long-lived gluinos in split supersymmetry, *JHEP* **09** (2004) p. 070,
arXiv: [hep-ph/0408248](https://arxiv.org/abs/hep-ph/0408248) [hep-ph].
- [2] N. Arkani-Hamed, S. Dimopoulos, G. F. Giudice, and A. Romanino,
Aspects of split supersymmetry, *Nucl. Phys.* **B709** (2005) p. 3,
arXiv: [hep-ph/0409232](https://arxiv.org/abs/hep-ph/0409232) [hep-ph].
- [3] R. Barbier et al., *R-parity violating supersymmetry*, *Phys. Rept.* **420** (2005) p. 1,
arXiv: [hep-ph/0406039](https://arxiv.org/abs/hep-ph/0406039) [hep-ph].
- [4] T. Han, Z. Si, K. M. Zurek, and M. J. Strassler,
Phenomenology of hidden valleys at hadron colliders, *JHEP* **07** (2008) p. 008,
arXiv: [0712.2041](https://arxiv.org/abs/0712.2041) [hep-ph].
- [5] L. Basso, A. Belyaev, S. Moretti, and C. H. Shepherd-Themistocleous,
Phenomenology of the minimal B-L extension of the Standard model: Z' and neutrinos,
Phys. Rev. **D80** (2009) p. 055030, arXiv: [0812.4313](https://arxiv.org/abs/0812.4313) [hep-ph].
- [6] D. Blackburn et al., “Search for long-lived, weakly-interacting particle that decay to displaced
hadronic jets in proton-proton collisions at $\sqrt{s} = 8$ TeV with the ATLAS detector”,
tech. rep. ATL-COM-PHYS-2013-683, CERN, 2013,
URL: <https://cds.cern.ch/record/1550730>.
- [7] O. Harris et al., “Searches for long-lived neutral particles decaying into heavy flavors in the
hadronic calorimeter of ATLAS at $\sqrt{s} = 8$ TeV”, tech. rep. ATL-COM-PHYS-2013-113,
EdBoard for EXOT-2012-28: Shlomit Tarem, Irene Vichou, Claudia Gemme, Masahiro Kuze.:
CERN, 2013, URL: <https://cds.cern.ch/record/1512932>.
- [8] G. Aad et al., *Search for massive, long-lived particles using multitrack displaced vertices or
displaced lepton pairs in pp collisions at $\sqrt{s} = 8$ TeV with the ATLAS detector*,
Phys. Rev. **D92** (2015) p. 072004, arXiv: [1504.05162](https://arxiv.org/abs/1504.05162) [hep-ex].
- [9] *DVxAODAnalysis: Analysis framework for Run 2 DV analysis*, (), URL: https://gitlab.cern.ch/atlas-phys-susy-secondary-vertex/DV%5C_xAODAnalysis.
- [10] J. Duarte-Campderros et al.,
“Search for long-lived, massive particles in events with displaced vertices and missing transverse
momentum in $\sqrt{s} = 13$ TeV *pp* collisions with the ATLAS detector”,
tech. rep. ATL-COM-PHYS-2016-508, CERN, 2016,
URL: <https://cds.cern.ch/record/2152010>.
- [11] J. Catmore, M. Elsing, E. Lipeles, D. Rousseau, and I. Vivarelli,
“Report of the Analysis Model Study Group”, tech. rep. ATL-COM-SOFT-2013-005,
CERN, 2013, URL: <https://cds.cern.ch/record/1543445>.
- [12] T. Sjöstrand, S. Mrenna, and P. Skands, *PYTHIA 6.4 physics and manual*,
Journal of High Energy Physics **2006** (2006) p. 026,
URL: <http://stacks.iop.org/1126-6708/2006/i=05/a=026>.

- [13] T. Cornelissen et al.,
 “Concepts, design and implementation of the ATLAS new tracking (NEWT)”,
 tech. rep. ATL-SOFT-PUB-2007-007. ATL-COM-SOFT-2007-002, CERN, 2007,
 URL: <https://cds.cern.ch/record/1020106>.
- [14] S. Che et al., “Performance of the reconstruction of large impact parameter tracks in the ATLAS inner detector”, tech. rep. ATL-COM-PHYS-2017-253, CERN, 2017,
 URL: <https://cds.cern.ch/record/2255680>.
- [15] ATLAS Collaboration, *Electron and photon reconstruction and identification in ATLAS: expected performance at high energy and results at 900 GeV*, ATLAS-CONF-2010-005, 2010,
 URL: <https://cds.cern.ch/record/1273197>.
- [16] G. Aad et al., *Muon reconstruction performance of the ATLAS detector in proton–proton collision data at $\sqrt{s} = 13$ TeV*, *Eur. Phys. J. C* **76** (2016) p. 292, arXiv: 1603.05598 [hep-ex].
- [17] ATLAS Collaboration, “Mapping the material in the ATLAS Inner Detector using secondary hadronic interactions in 7 TeV collisions”, tech. rep. ATLAS-CONF-2010-058, CERN, 2010,
 URL: <https://cds.cern.ch/record/1281331>.
- [18] *Electron and Photon Selection and Identification for Run2*, URL: <https://twiki.cern.ch/twiki/bin/view/AtlasProtected/EGammaIdentificationRun2>.
- [19] *MCPAnalysisGuidelinesMC15*, URL: <https://twiki.cern.ch/twiki/bin/view/AtlasProtected/MCPAnalysisGuidelinesMC15>.
- [20] M. Backhaus,
 “The upgraded Pixel Detector of the ATLAS Experiment for Run2 at the Large Hadron Collider”,
 tech. rep. ATL-INDET-PROC-2015-015, CERN, 2015,
 URL: <https://cds.cern.ch/record/2110260>.
- [21] M. Aaboud et al., *A measurement of material in the ATLAS tracker using secondary hadronic interactions in 7 TeV pp collisions*, *JINST* **11** (2016) P11020, arXiv: 1609.04305 [hep-ex].
- [22] G. Aad et al., *Kshort and Λ production in pp interactions at $\sqrt{s} = 0.9$ and 7 TeV measured with the ATLAS detector at the LHC*, *Phys. Rev. D* **85** (2012) p. 012001, arXiv: 1111.1297 [hep-ex].

568 **List of contributions**

Siinn Che	Main analyzer
569 K.K. Gan	Advisor
570 Christopher B. Martin	Main analyzer

571 **Appendices**

572 **A Truth-level p_T and η distributions of Signal MC samples**

B K_S yields in data and MC samples used for tracking and vertexing systematic uncertainties

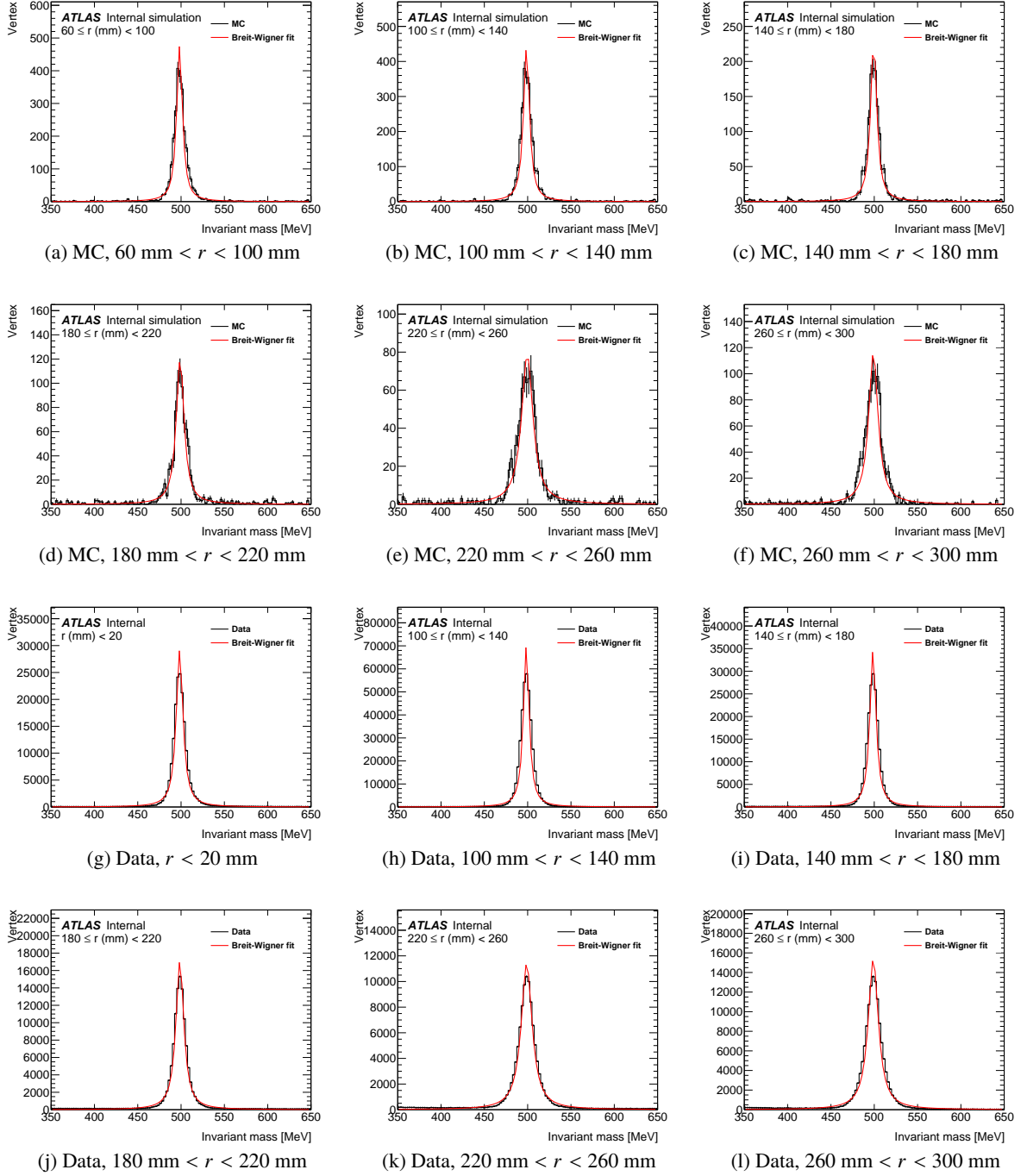


Figure 15: K_S invariant mass for various r . The curve shows a fit to Breit-Wigner.

C K_S and Z' comparison

The ideal K_S sample to estimate the systematic uncertainty in track and vertex reconstruction should have the same kinematic distributions as the Z' MC sample. Obviously, the two samples have different distributions. Figure 16 shows comparisons of the vertex/kinematic distributions. The reconstructed K_S and Z' vertices are required to match to a K_S and Z' vertex produced at truth level with spatial displacement no larger than 0.7 mm. All distributions are normalized to unity. It is fortuitous that their distribution cover the r region of interest, and the two samples have similar z distribution.

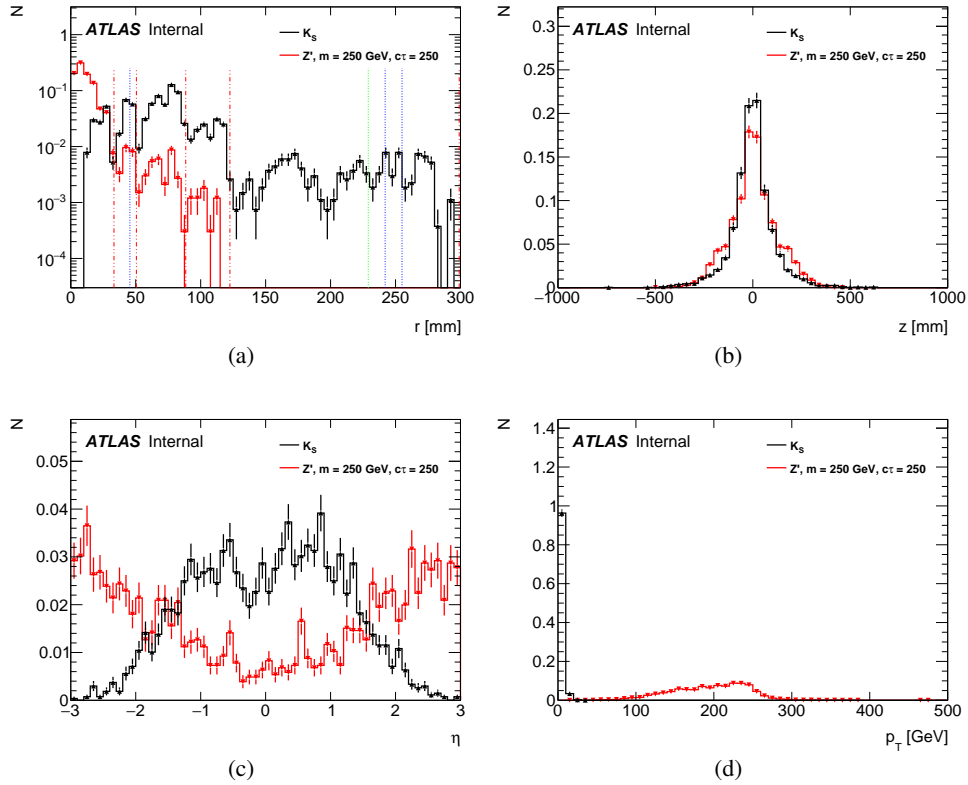


Figure 16: Distribution of transverse (a), longitudinal vertex position (b), η (c), p_T (d), and the opening angle between decay particles of K_S and Z' vertices (e) found in the background and the signal MC sample described in Section ???. Distributions are normalized to unity. Systematic uncertainties are shown.

D Track flipping extrapolation method

D.1 Extrapolation from control region

$$\begin{aligned}
 N_{\mu x}^{est} &\approx 2 \cdot P(\mu) \cdot N_{xx}^{obs}, \\
 N_{ex}^{est} &\approx 2 \cdot P(e) \cdot N_{xx}^{obs}, \\
 N_{\mu\mu}^{est} &\approx P(\mu)^2 \cdot N_{xx}^{obs}, \\
 N_{ee}^{est} &\approx P(e)^2 \cdot N_{xx}^{obs}, \\
 N_{e\mu}^{est} &\approx 2 \cdot P(e) \cdot P(\mu) \cdot N_{xx}^{obs}.
 \end{aligned} \tag{3}$$

where N^{obs} and N^{est} represent track-flipping vertex yield and estimated vertex yield by the extrapolation, respectively.

D.2 Extrapolation from validation region

$$\begin{aligned}
 N_{\mu\mu}^{est} &\approx \frac{1}{2} \cdot P(\mu) \cdot N_{\mu x}^{obs}, \\
 N_{ee}^{est} &\approx \frac{1}{2} \cdot P(e) \cdot N_{ex}^{obs}, \\
 N_{e\mu}^{est} &\approx \frac{1}{2} \cdot (P(e) \cdot N_{\mu x}^{obs} + P(\mu) \cdot N_{ex}^{obs})
 \end{aligned} \tag{4}$$

D.3 Scale factors

$$\begin{aligned}
 S_{xx \rightarrow \mu\mu} &= S_{xx \rightarrow \mu x}^2 \\
 S_{xx \rightarrow ee} &= S_{xx \rightarrow ex}^2 \\
 S_{xx \rightarrow e\mu} &= \left(\frac{1}{2} (S_{xx \rightarrow \mu x} + S_{xx \rightarrow ex}) \right)^2 \\
 S_{\mu x \rightarrow \mu\mu} &= S_{xx \rightarrow \mu x} \\
 S_{ex \rightarrow ee} &= S_{xx \rightarrow ex} \\
 S_{ex, \mu x \rightarrow e\mu} &= \frac{1}{2} (S_{xx \rightarrow \mu x} + S_{xx \rightarrow ex})
 \end{aligned} \tag{5}$$