
A Discussion of CDF's Recent $B_s \rightarrow \mu^+\mu^-$ Result

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Fermilab

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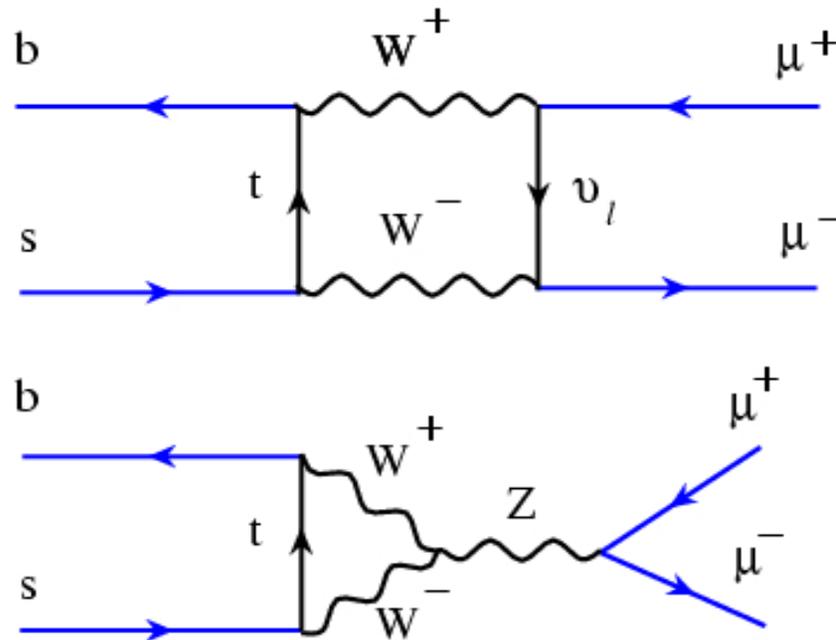
To be clear...

- “Search for $B_s \rightarrow \mu^+ \mu^-$ and $B_d \rightarrow \mu^+ \mu^-$ Decays with CDF II”
 - [arXiv:1107.2304](https://arxiv.org/abs/1107.2304)
 - accepted to PRL
- Public web page
<http://www-cdf.fnal.gov/physics/new/bottom/110707.blessed-Bsd2mumu/>

Why?

Introduction

- In the SM $B_s \rightarrow \mu^+ \mu^-$ is an FCNC... only possible at the loop level

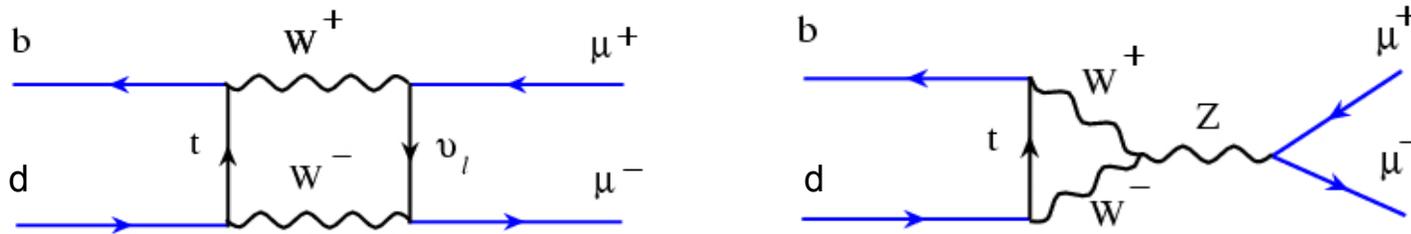


$$BF(B_s \rightarrow \mu^+ \mu^-) = (3.2 \pm 0.2) \times 10^{-9}$$

(E.Gamiz et al. (HPQCD Collaboration), A.J. Buras et al.)

Introduction

- All this also true for $B_d \rightarrow \mu^+ \mu^-$ decays too

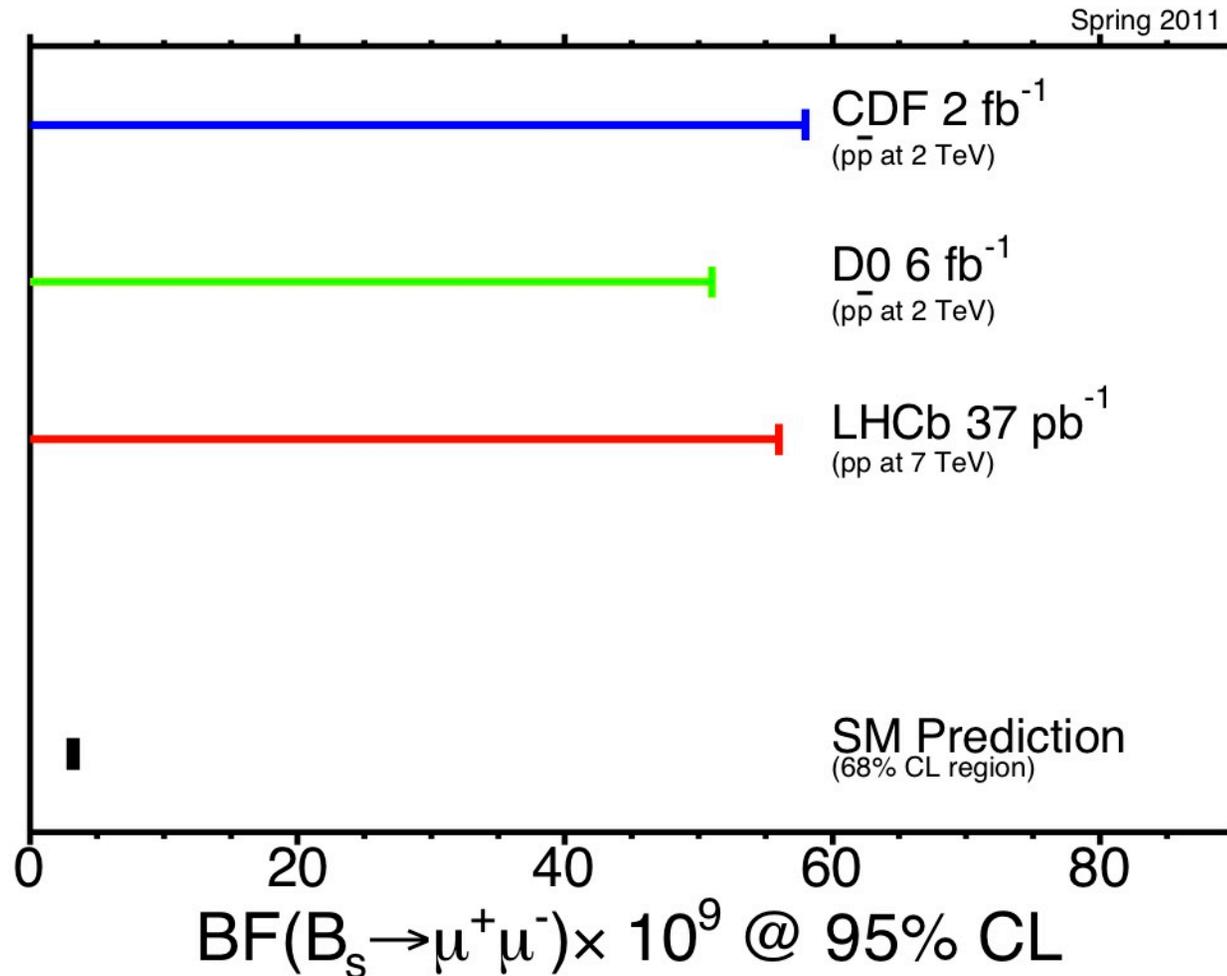


$$BF(B_d \rightarrow \mu^+ \mu^-) = (1.0 \pm 0.1) \times 10^{-10}$$

(E.Gamiz et al. (HPQCD Collaboration), A.J. Buras et al.)

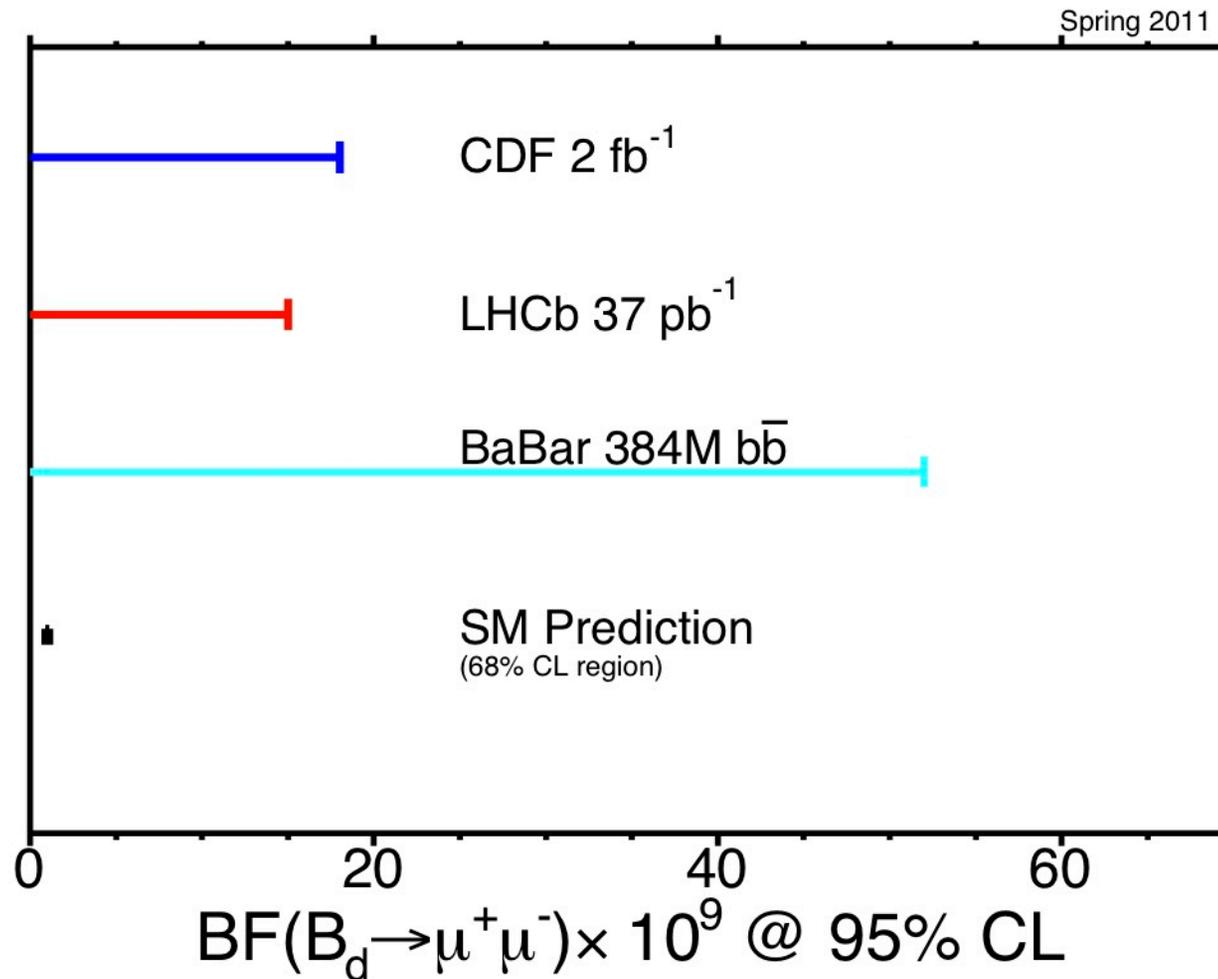
- BF relative to $BF(B_s \rightarrow \mu^+ \mu^-)$ model dependent
 - measurements of both sensitive to flavor structure of underlying physics model
 - In MFV models, $BF(B_d/B_s) \sim |V_{td}|^2/|V_{ts}|^2 \sim 1/20$

Experimental Status: Spring 2011



- Has not yet been experimentally observed

Experimental Status: Spring 2011

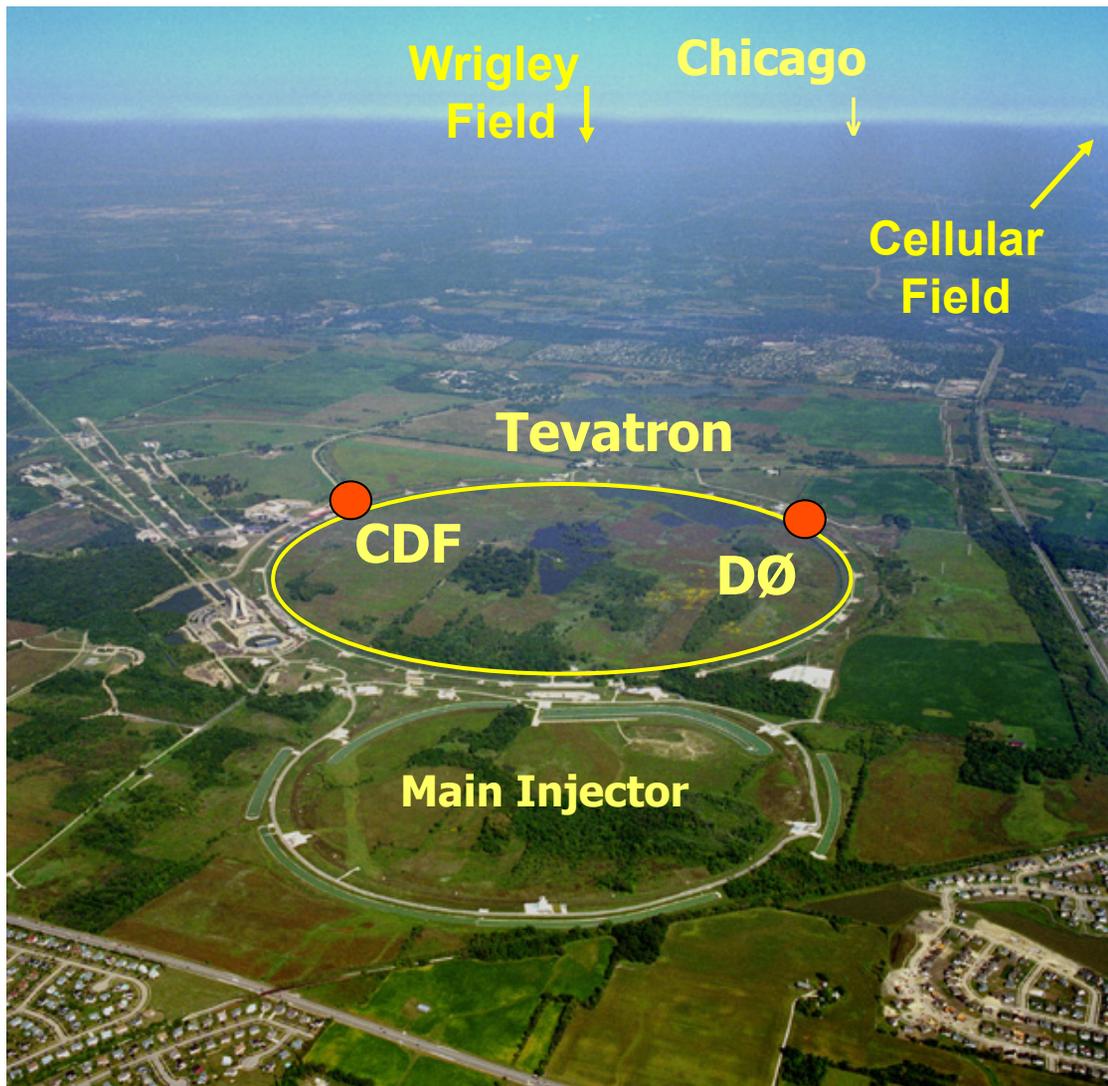


- Has not yet been experimentally observed



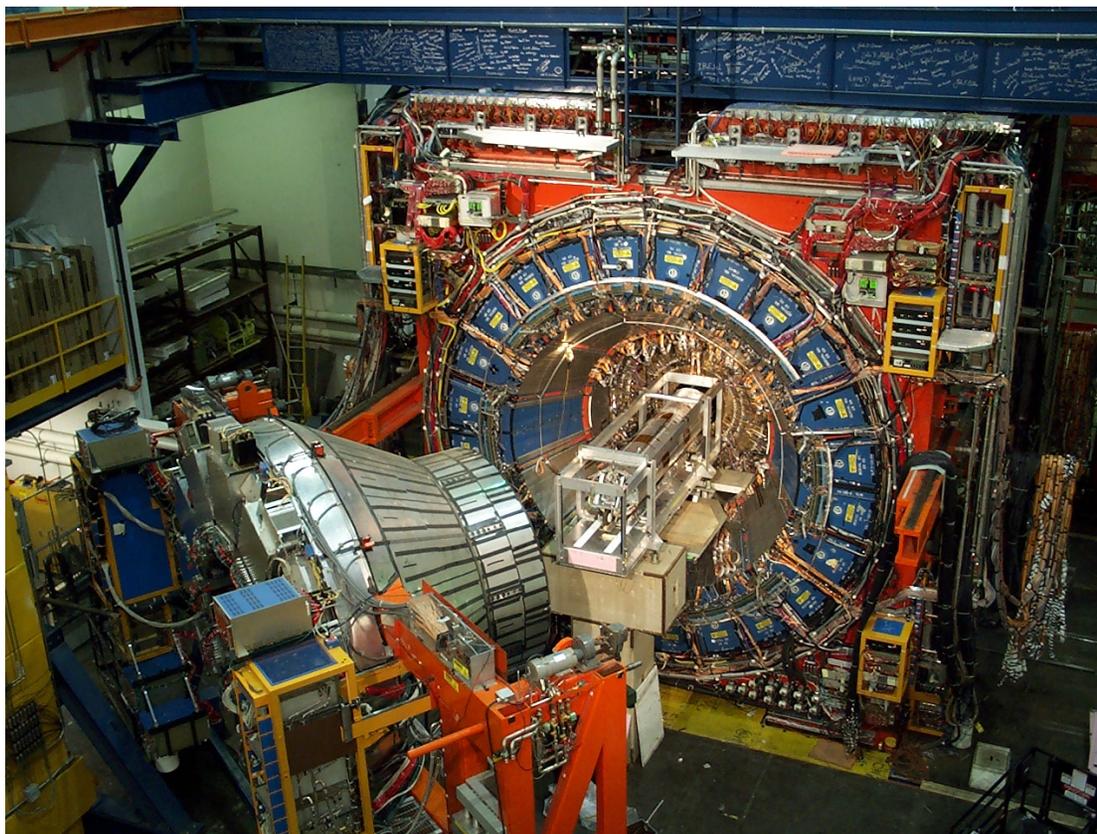
Where?

Fermilab Tevatron



- $p\bar{p}$ collider at
 $E_{\text{cm}} = 2 \text{ TeV}$
- Run-II 2001-2011
(12 fb^{-1} / exp delivered)
- Performing excellently
- All B-hadron species copiously produced

The CDF Experiment

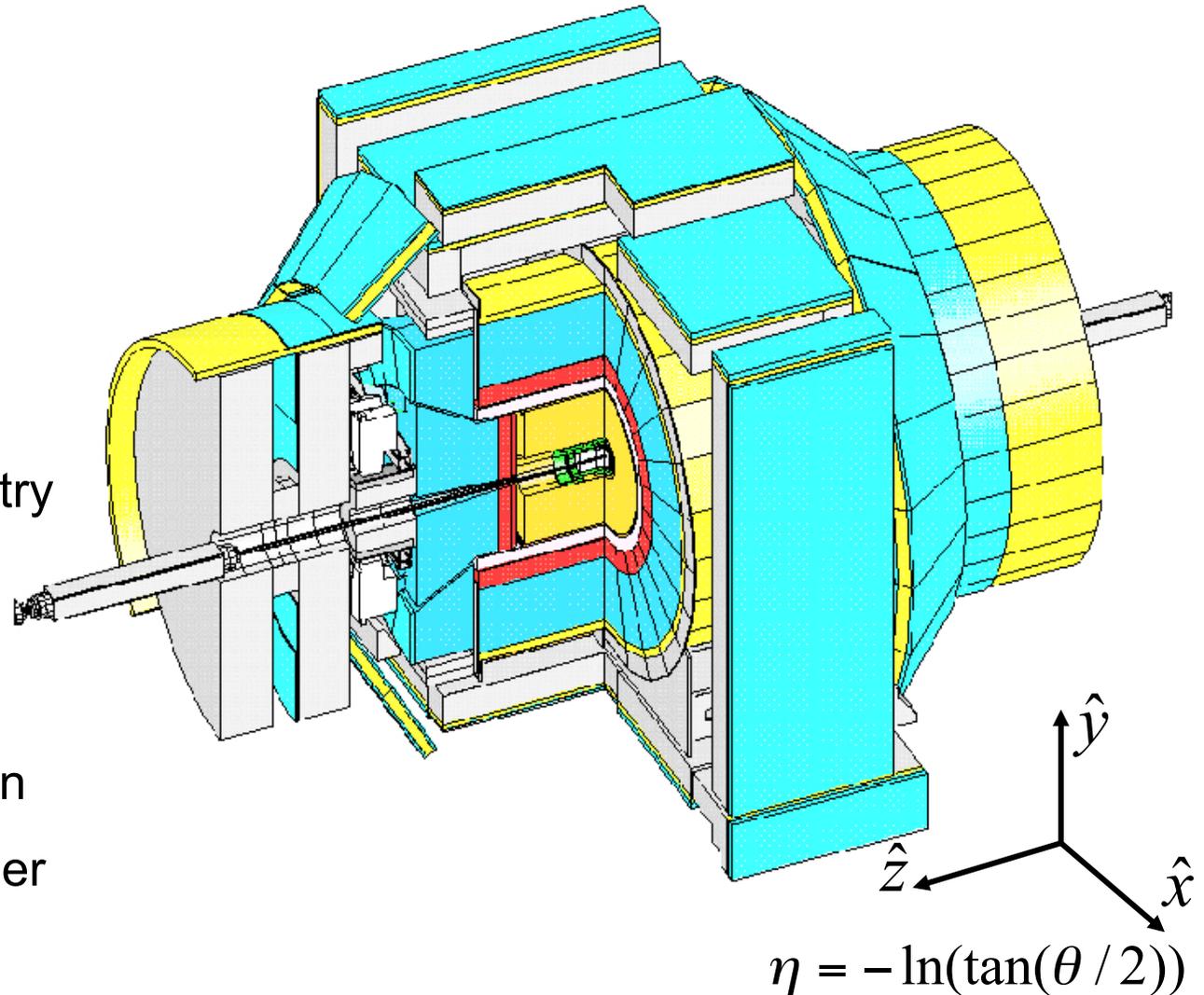


- Multipurpose collider detector
- Pioneered silicon detectors at a hadron collider
- International collaboration, 600+ members

CDF Detector

Features:

- Precision silicon vertexing
- Large radius drift chamber ($r=1.4\text{m}$)
- 1.4 T solenoid
- projective calorimetry ($|\eta| < 3.5$)
- muon chambers ($|\eta| < 1.0$)
- Particle identification
- Silicon Vertex Trigger





How?

Analysis Description

- This is a simple analysis
 - 1) Find events with 2 muons in them
 - 2) Identify means to suppress background while keeping as much signal as possible
 - 3) Look for a bump in the $m_{\mu\mu}$ distribution

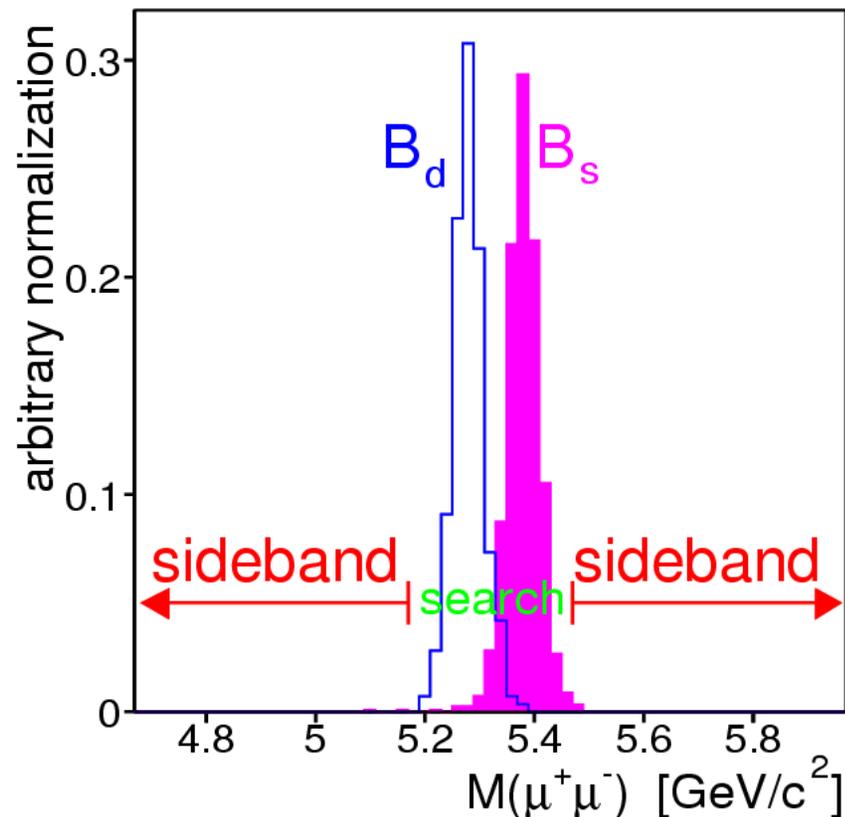
Analysis Strategy

- Our strategy is simple
 - “blind” ourselves to an extended mass signal region
 - Use data in the mass sidebands to estimate the dominant backgnd contribution to the signal region
 - Employ an a priori optimization to choose our final selection criteria
 - Build confidence in background estimate using control regions prior to “opening the box”

Emphasis on being robust and unbiased

Definition of Signal / Sideband Regions

- We “blind” the data in an extended signal region



Search Region:

- $5.169 < m_{\mu\mu} < 5.469 \text{ GeV}/c^2$
- corresponds to $\pm 4\sigma(m_{\mu\mu})$
- final region $\pm 2.5\sigma(m_{\mu\mu})$

Sideband Regions:

- additional 0.5 GeV on either side of search region
- used to understand Bkgd

Some Preliminaries

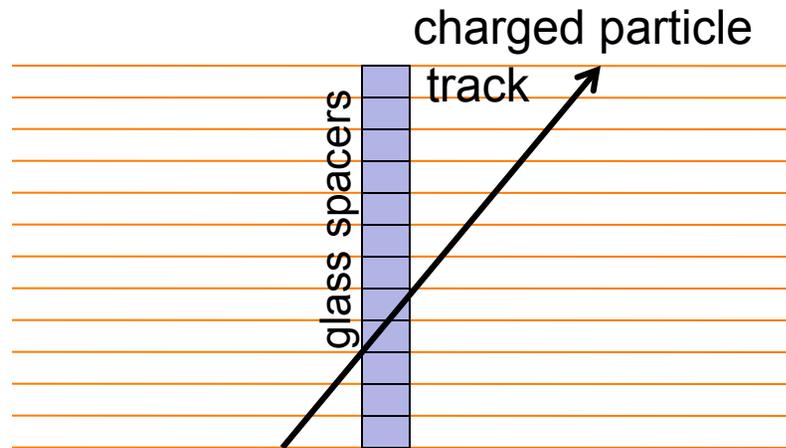
CDF Run-II $B_s \rightarrow \mu^+\mu^-$ Publications

- 1) 170 pb⁻¹ PRL 93, 032001 (2004). 78 citations
 - Sensitivity x3 improvement over Run I
- 2) 350 pb⁻¹ PRL 95, 221805 (2005). 50 citations
 - Sensitivity x4 improvement over 1)
- 3) 2 fb⁻¹ PRL 100, 101802 (2008). 197 citations
 - Sensitivity x4 improvement over 2)
- 4) 7fb⁻¹ accepted PRL, arXiv:1107.2304
 - Sensitivity x3 improvement over 3)

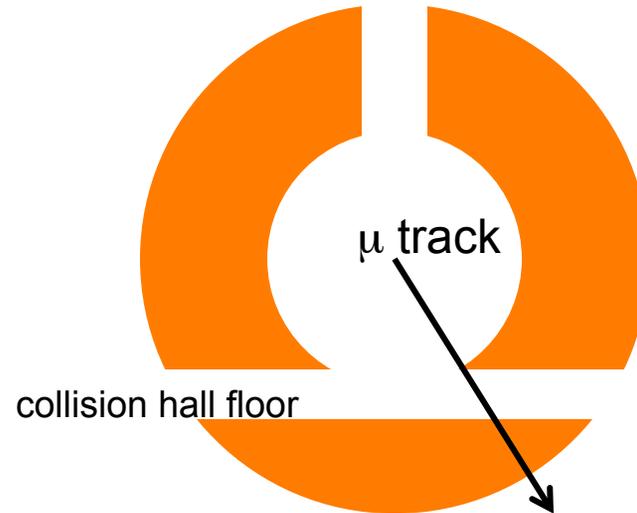
Sensitivity Improvements

- **Added acceptance**
 - Include events that pass near COT “spacer”
 - Include events in CMX “mini-skirts”
- **Improved background discrimination**
 - Improved dE/dX calibrations
 - Improved performance of multi-variate discriminant used in final selection criteria

Additional Acceptance



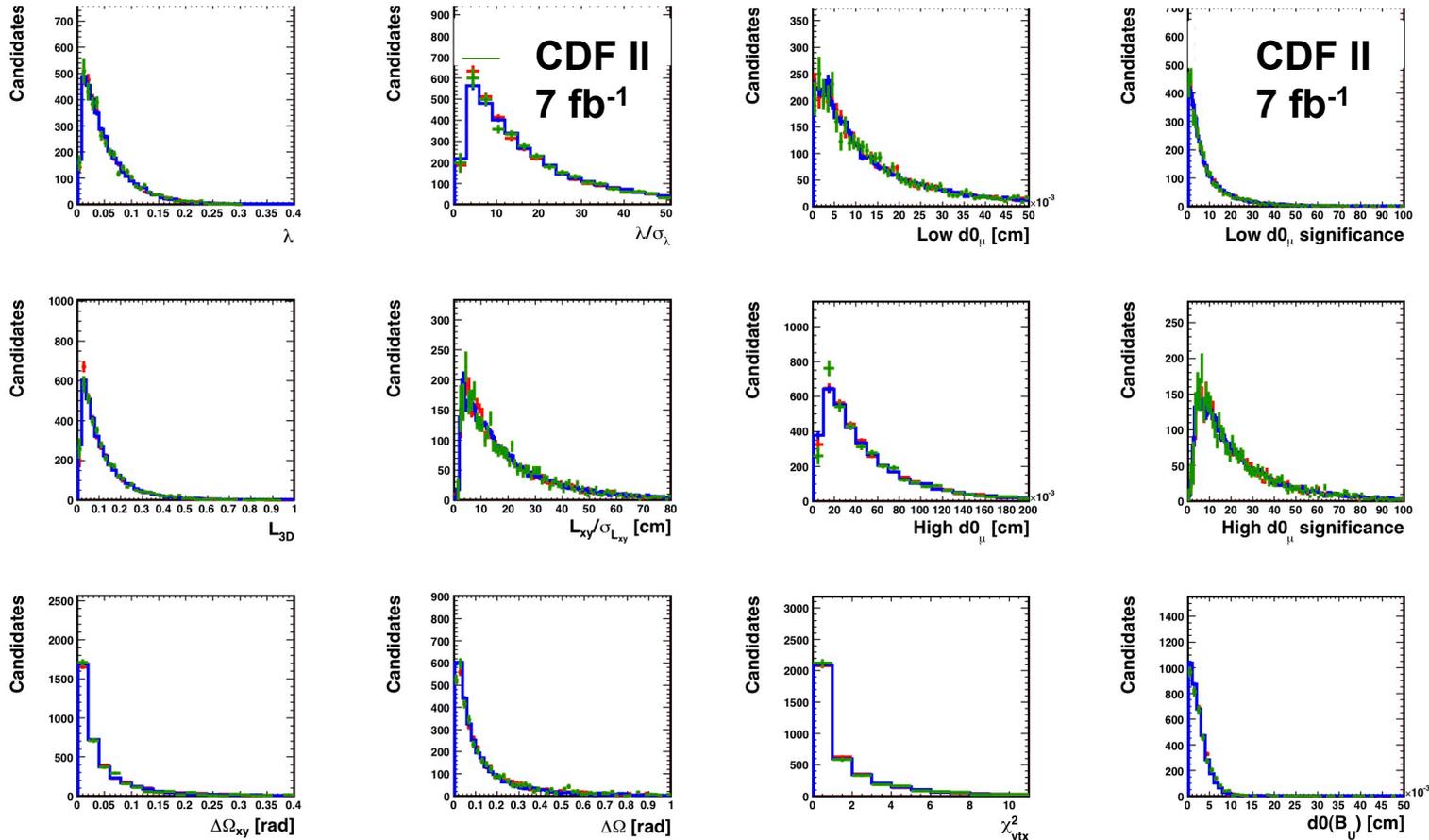
CDF Tracking Drift Chamber - COT
(not to scale)



CDF Central Muon Extension - CMX
(cartoon - not to scale)

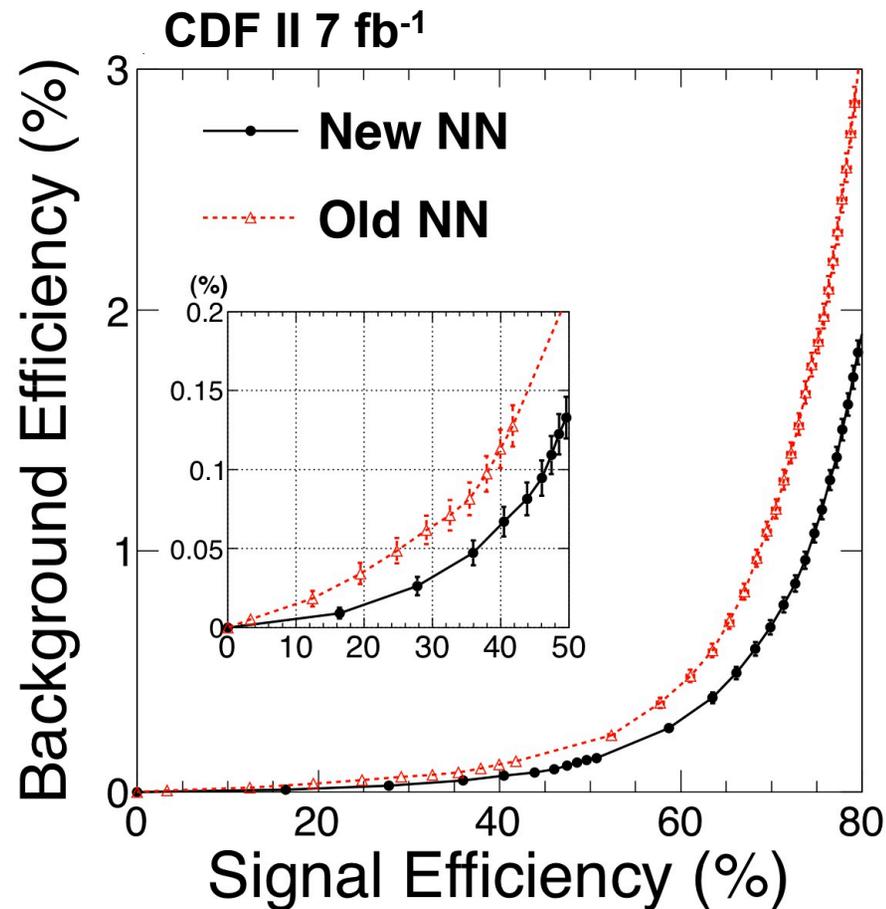
- We've always had these events on tape
- Just needed to understand the trigger efficiencies in these regions

Additional Acceptance



- Kinematics unaffected

Improved Background Discrimination



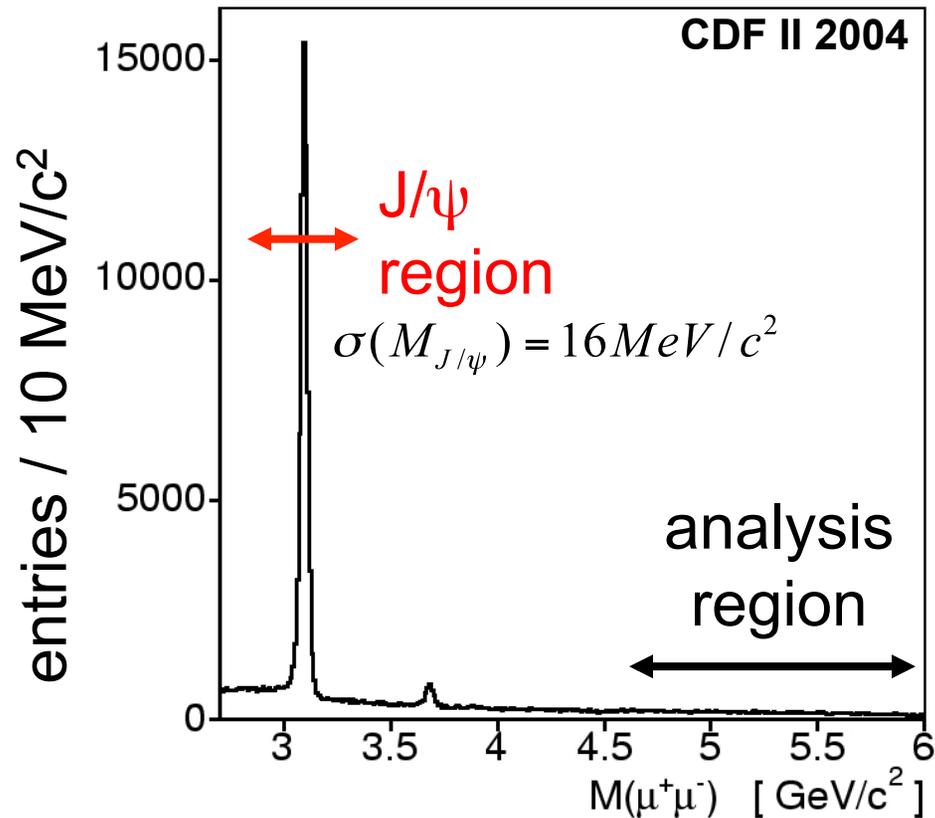
- Improved Neural Net (NN) performance

Normalization

- We employ a relative normalization
 - Using $B^+ \rightarrow J/\psi K^+ \rightarrow \mu^+ \mu^- K^+$ events
 - Collect B^+ and signal events with same trigger
 - Many exp. uncertainties significantly reduced

$$BF(B_{s,d} \rightarrow \mu^+ \mu^-) = \left(\frac{N_{Bs,d}}{N_{B^+}} \right) \left(\frac{\alpha_{B^+} \epsilon_{B^+}}{\alpha_{Bs,d} \epsilon_{Bs,d}} \right) \left(\frac{f_u}{f_s} \right) BF(B^+ \rightarrow J/\psi K^+)$$

Di-muon Mass Distribution from Trigger



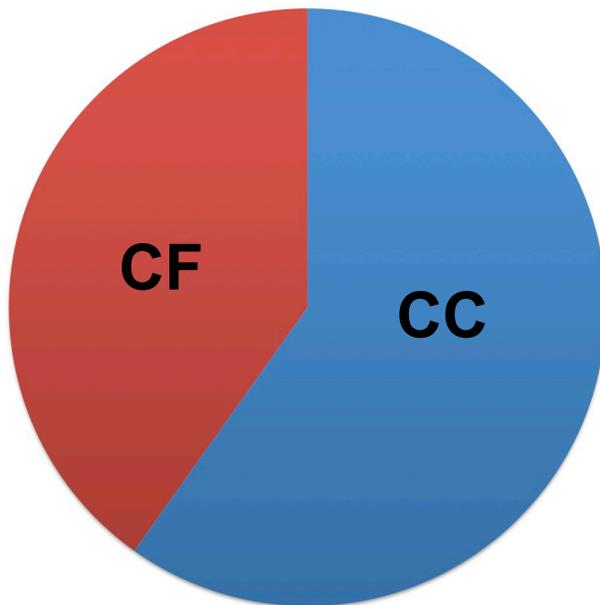
- The trigger paths used for this analysis
 - Collect signal sample: $B \rightarrow \mu^+\mu^-$
 - Collect control sample: $J/\psi \rightarrow \mu^+\mu^-$, $B^+ \rightarrow J/\psi K^+$

Our Trigger Paths

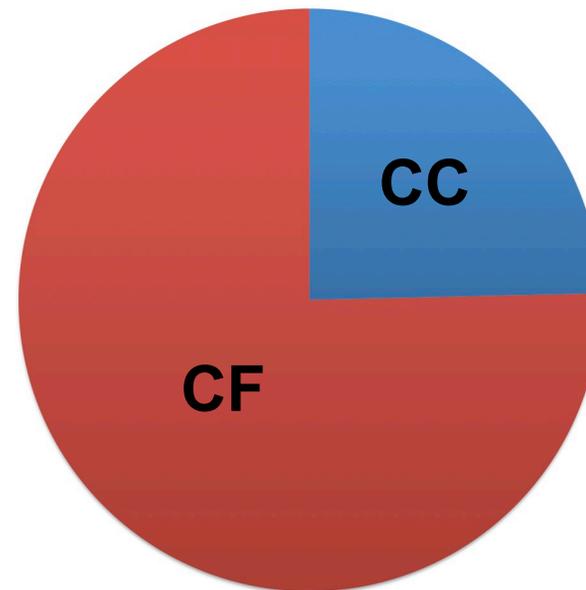
- Collect data using two separate trigger paths corresponding to two separate topologies:
- “Central-Central” (CC)
 - both muons $|\eta| < 0.6$
 - $P_T(\mu) > 1.5 \text{ GeV}/c$
 - $2.7 < m_{\mu\mu} < 6.0 \text{ GeV}/c^2$
 - $\Delta\phi(\mu\mu) < 2.25 \text{ rad}$
 - $P_T(\mu^+) + P_T(\mu^-) > 5 \text{ GeV}/c$
- “Central-Forward” (CF)
 - $|\eta_{\mu 1}| < 0.6, 0.6 < |\eta_{\mu 2}| < 1$
 - $P_T(C) > 1.5 \text{ GeV}/c$
 - $P_T(F) > 2.0 \text{ GeV}/c$
 - $2.7 < m_{\mu\mu} < 6.0 \text{ GeV}/c^2$
 - $\Delta\phi(\mu\mu) < 2.25 \text{ rad}$
 - $P_T(\mu^+) + P_T(\mu^-) > 5 \text{ GeV}/c$

CC vs CF Channels

Signal acceptance



Background Yields



- Treat each channel separately, combine at end

Normalization

- We employ a relative normalization

$$BF(B_{s,d} \rightarrow \mu^+ \mu^-) = \left(\frac{N_{Bs,d}}{N_{B^+}} \right) \left(\frac{\alpha_{B^+} \epsilon_{B^+}}{\alpha_{Bs,d} \epsilon_{Bs,d}} \right) \left(\frac{f_u}{f_s} \right) BF(B^+ \rightarrow J/\psi K^+)$$

● From fits to the data.

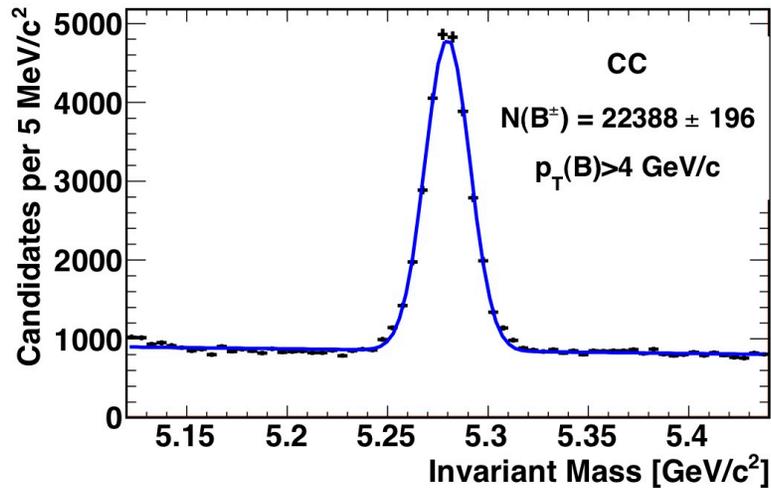
● From the PDG 2010:

$$\frac{f_u}{f_s} = 3.55 \pm 0.47$$

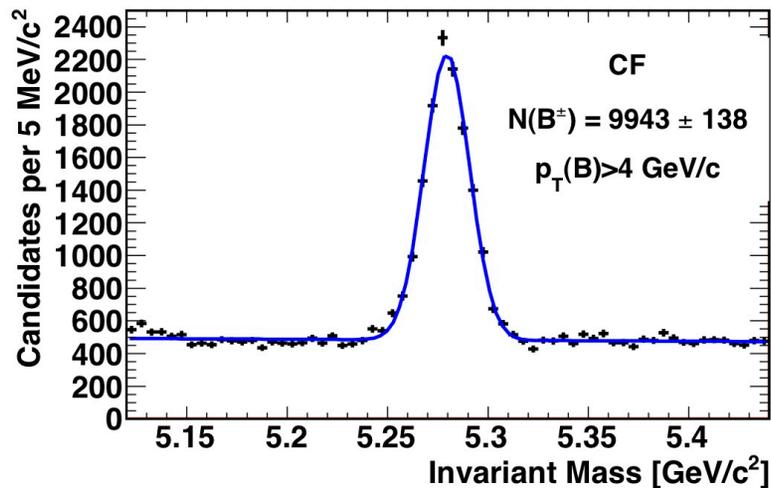
$$BF(B^+ \rightarrow J/\psi K^+) BF(J/\psi \rightarrow \mu^+ \mu^-) = (6.01 \pm 0.21) \times 10^{-5}$$

Normalization: B⁺ Yield

CDF II 7 fb⁻¹



CDF II 7 fb⁻¹



- Use sideband subtracted signal yields
- $B^+ \rightarrow J/\psi\pi^+$ contamination < 1%

Normalization

- We employ a relative normalization

$$BF(B_{s,d} \rightarrow \mu^+ \mu^-) = \left(\frac{N_{Bs,d}}{N_{B^+}} \right) \left(\frac{\alpha_{B^+} \epsilon_{B^+}}{\alpha_{Bs,d} \epsilon_{Bs,d}} \right) \left(\frac{f_u}{f_s} \right) BF(B^+ \rightarrow J/\psi K^+)$$

$\alpha_B \equiv$ geometric and kinematic acceptance of trigger
(from MC simulation)

$$\epsilon_B \equiv \epsilon_{\text{reco}} \cdot \epsilon_{\text{NN}} \cdot \epsilon_{\text{mass}} = \left(\epsilon_{\text{track}} \cdot \epsilon_{\mu\text{-ID}} \cdot \epsilon_{\text{vertex}} \right) \cdot \epsilon_{\text{NN}} \cdot \epsilon_{\text{mass}}$$



From data using “Tag and Probe”



From MC, checked with B^+ and J/ψ data

Normalization

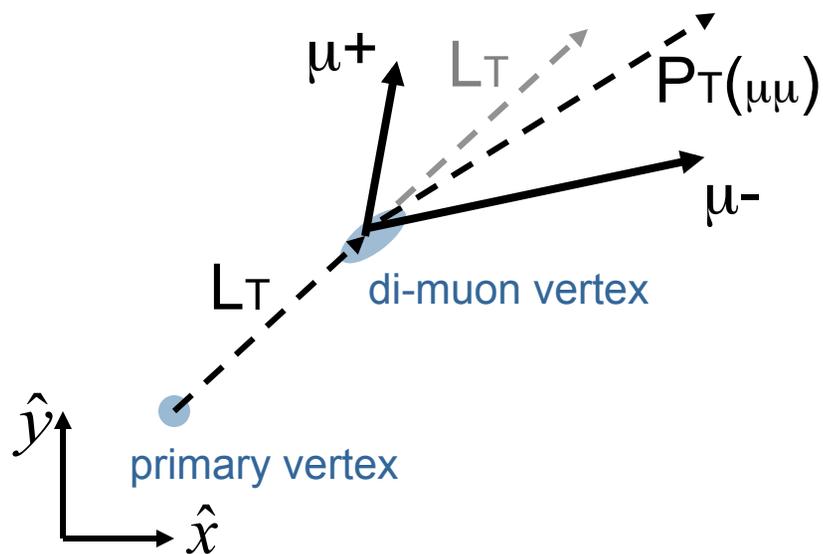
$$BF(B_{s,d} \rightarrow \mu^+ \mu^-) = N_{B_{s,d}} \cdot ses$$

	CC		CF	
$(\alpha_{B^+} / \alpha_{B_s})$	0.307 ± 0.018	($\pm 6\%$)	0.197 ± 0.014	($\pm 7\%$)
$(\epsilon_{B^+}^{trig} / \epsilon_{B_s}^{trig})$	0.99935 ± 0.00012	(< 1%)	0.97974 ± 0.00016	(< 1%)
$(\epsilon_{B^+}^{reco} / \epsilon_{B_s}^{reco})$	0.85 ± 0.06	($\pm 8\%$)	0.84 ± 0.06	($\pm 9\%$)
$\epsilon_{B_s}^{NN} (NN > 0.70)$	0.915 ± 0.042	($\pm 4\%$)	0.864 ± 0.040	($\pm 4\%$)
$\epsilon_{B_s}^{NN} (NN > 0.995)$	0.461 ± 0.021	($\pm 5\%$)	0.468 ± 0.022	($\pm 5\%$)
N_{B^+}	22388 ± 196	($\pm 1\%$)	9943 ± 138	($\pm 1\%$)
f_u / f_s	3.55 ± 0.47	($\pm 13\%$)	3.55 ± 0.47	($\pm 13\%$)
$BR(B^+ \rightarrow J/\psi K^+ \rightarrow \mu^+ \mu^- K^+)$	$(6.01 \pm 0.21) \times 10^{-5}$	($\pm 4\%$)	$(6.01 \pm 0.21) \times 10^{-5}$	($\pm 4\%$)
SES (All bins)	$(2.9 \pm 0.5) \times 10^{-9}$	($\pm 18\%$)	$(4.0 \pm 0.7) \times 10^{-9}$	($\pm 18\%$)

$$ses(CC+CF) = 1.7 \times 10^{-9}$$

- Uncertainty includes: variations in the $p_T(B)$ spectrum, kinematic differences between J/ψ and $B_s \rightarrow \mu\mu$, variations in simulation parameters, differences between B^+ data and MC

Some Definitions



3D and 2D versions of variables
2D denoted with subscript "T"

- $P(B)$ = momentum of B

$$P_B = P_{\mu\mu} = \vec{P}_{\mu^+} + \vec{P}_{\mu^-}$$

- L : decay length

$$L = \vec{L} \cdot \vec{P}_{\mu\mu} / |\vec{P}_{\mu\mu}|$$

- λ : proper decay time

$$\lambda = cLm_{\mu\mu} / P_{\mu\mu}$$

- $\Delta\Omega$ = pointing angle

$$\Delta\Omega = \angle(\vec{L}, \vec{P}_{\mu\mu})$$

The Details...

Analysis Description

- This is a simple analysis
 - ✓) Find events with 2 muons in them
 - 2) Identify means to suppress background while keeping as much signal as possible
 - 3) Look for a bump in the $m_{\mu\mu}$ distribution

Suppress Background, Keep Signal

- We start with some simple “baseline” requirements to ensure two good muons that originate from a common vertex
- Then we exploit features of our signal events to discriminate signal from background

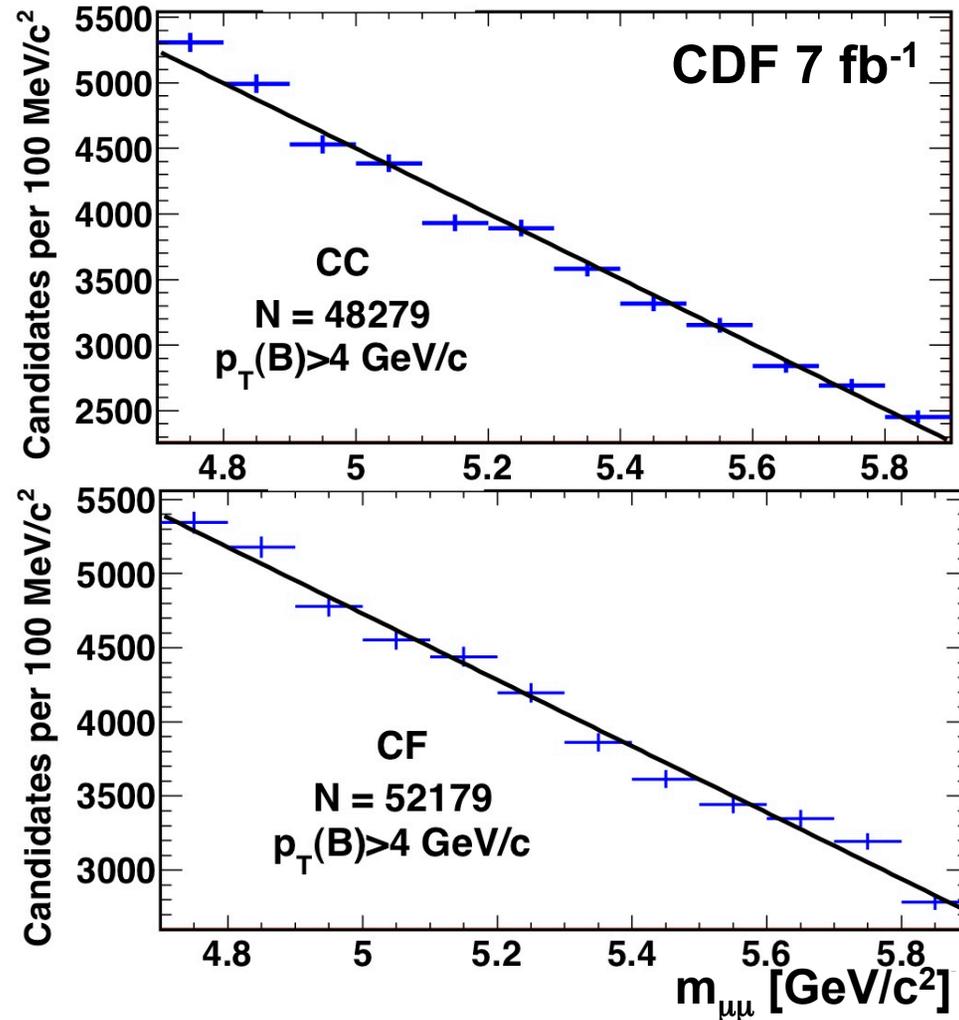
Baseline Requirements

We require:

- “good” COT tracks and μ track-stubs
- ≥ 3 silicon r - ϕ hits
- $4.669 < m_{\mu\mu} < 5.969 \text{ GeV}/c^2$
- “good” vertex
 - $\sigma(L) < 150 \text{ } \mu\text{m}$
 - $\chi^2 < 15$
 - $L_T < 1 \text{ cm}$
- $P_T(C) > 2.0, P_T(F) > 2.2 \text{ GeV}/c$
 - $P_T(\mu\mu) > 4 \text{ GeV}/c$
 - $\lambda < 0.3 \text{ cm}$
 - $\lambda / \sigma_\lambda > 2$
 - $\Delta\Omega < 0.70 \text{ rad}$
 - Isolation > 0.50

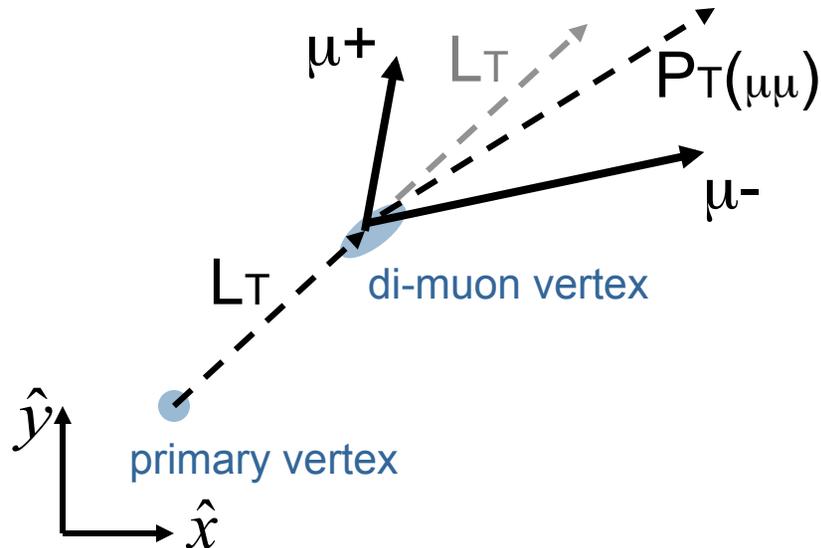
maintain most the signal while significantly reducing bgd

Baseline sample



- Completely background dominated

Discriminate Signal from Background



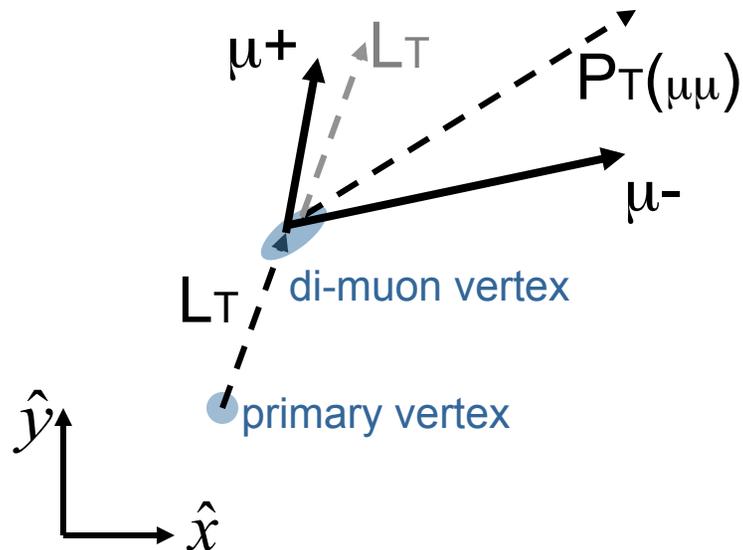
Signal characteristics

- final state is fully reconstructed
- B_s has long lifetime ($c\tau = 440 \mu\text{m}$)
- B fragmentation is hard

For real $B_s \rightarrow \mu^+\mu^-$ expect:

- $m_{\mu\mu} = m(B_s)$
- $\lambda = cL_T m_{\mu\mu}/P_T(\mu\mu)$ to be large
- L_T and $P_T(\mu\mu)$ to be co-linear (ie. small $\Delta\Omega$)
- few additional tracks

Discriminate Signal from Background



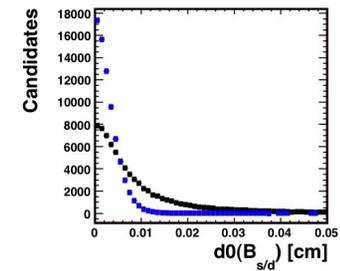
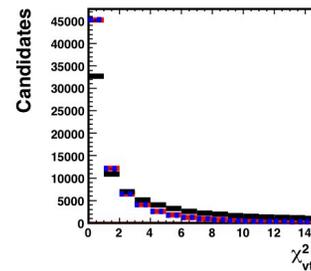
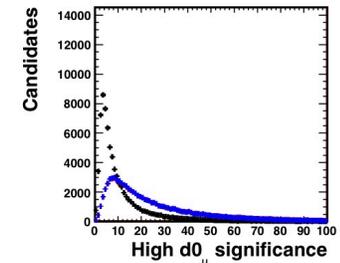
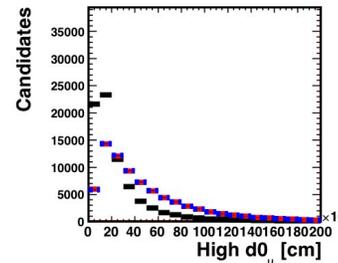
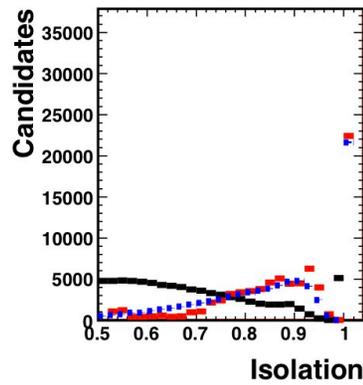
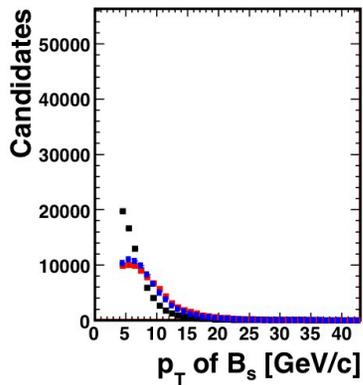
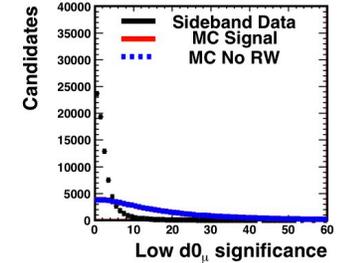
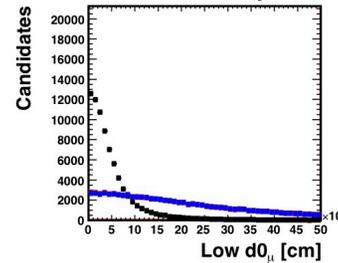
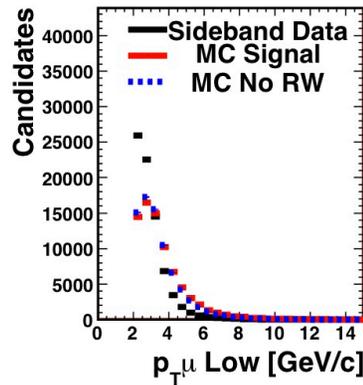
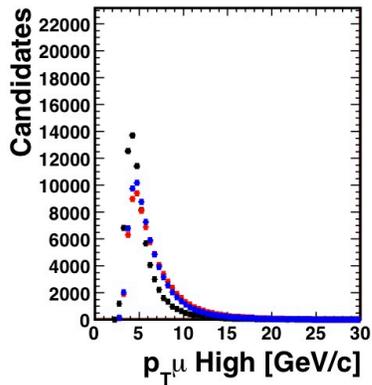
Contributing Backgrounds

- sequential semi-leptonic decay, $b \rightarrow \mu^- c X \rightarrow \mu^+ \mu^- X$
- double semi-leptonic decay, $g \rightarrow b \bar{b} \rightarrow \mu^+ \mu^- X$
- continuum $\mu^+ \mu^-$, μ + fake
fake+fake

In general:

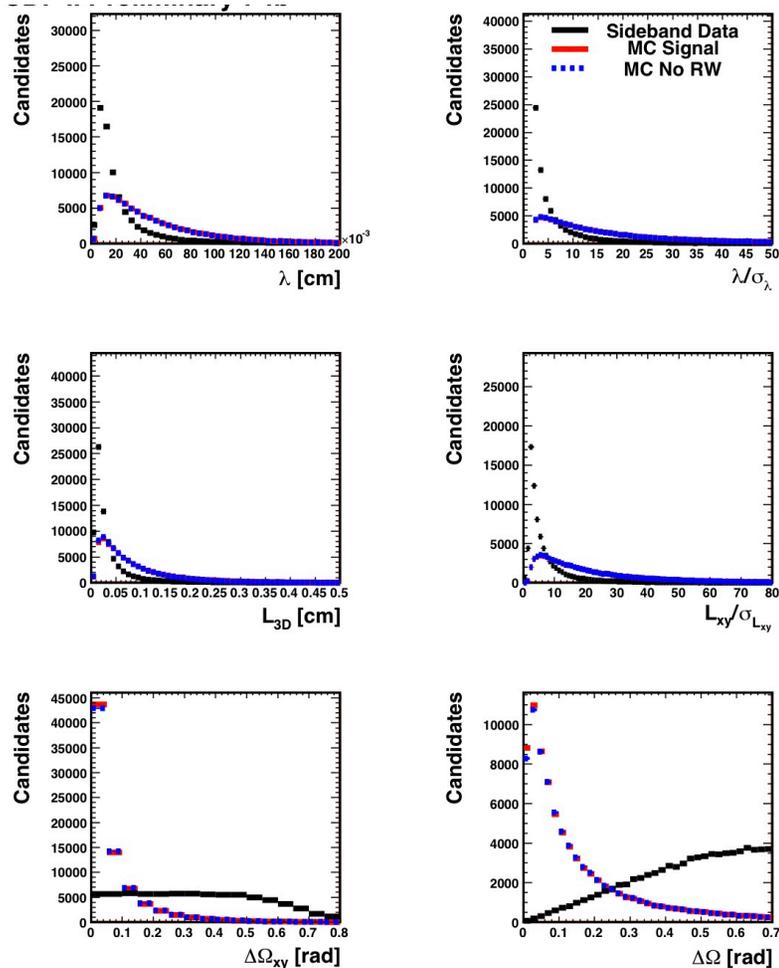
- $m_{\mu\mu} \neq m(B_s)$
- $\lambda = c L_T m_{\mu\mu} / P_T(\mu\mu)$
will be smaller
- L_T and $P_T(\mu\mu)$ will not be co-linear (large $\Delta\Omega$)
- more additional tracks

Discriminating Variables



- Some variables that take advantage of these distinguishing characteristics

Discriminating Variables

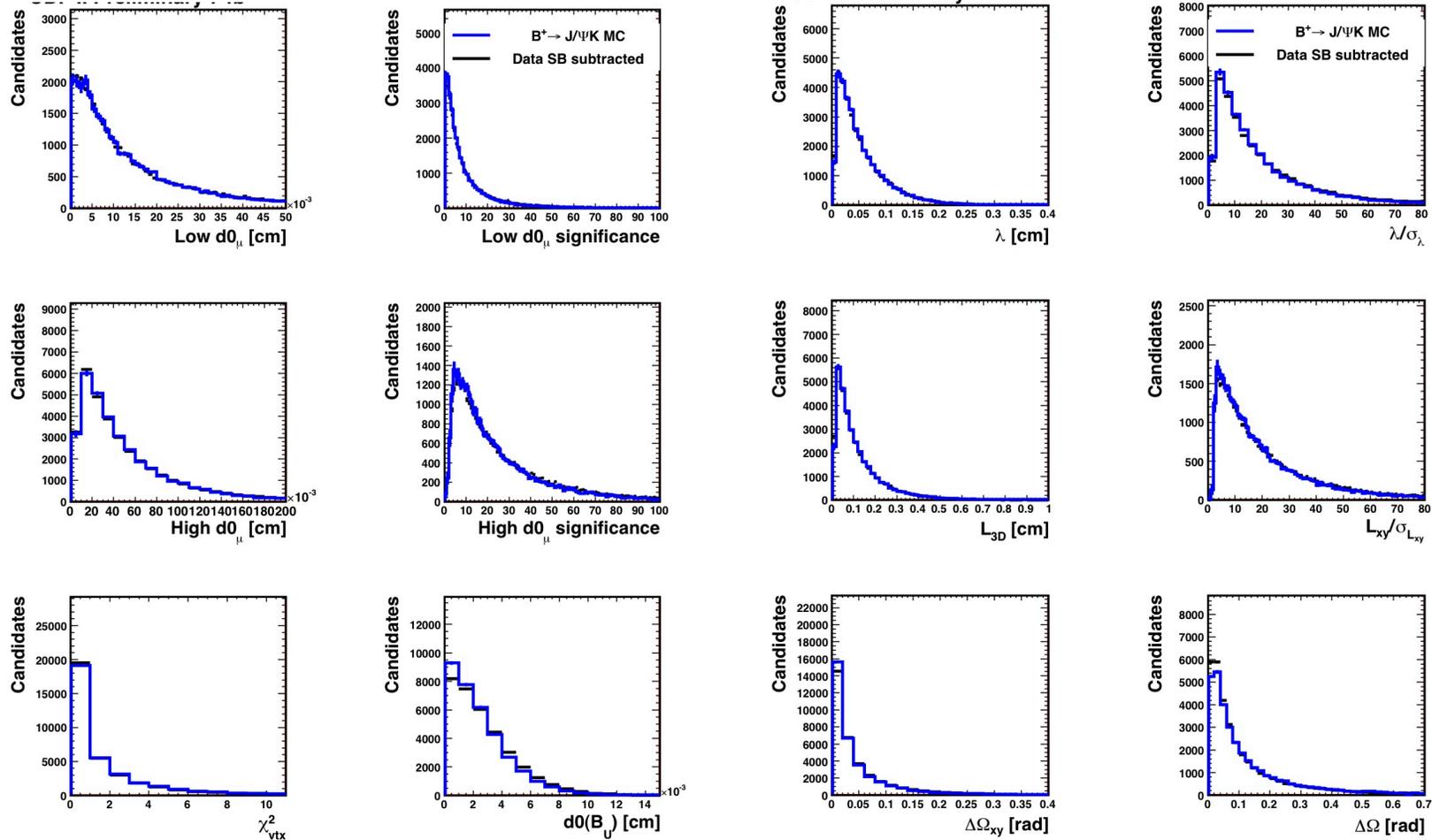


- more variables that take advantage of these distinguishing characteristics

Discriminate Signal from Background

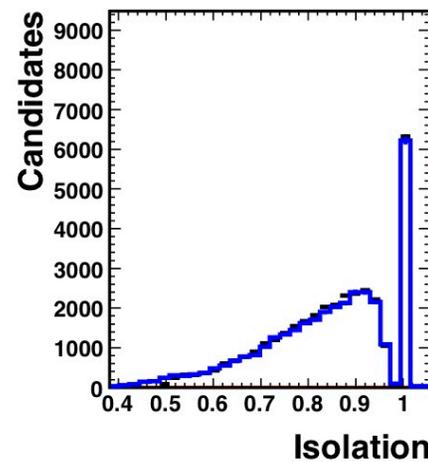
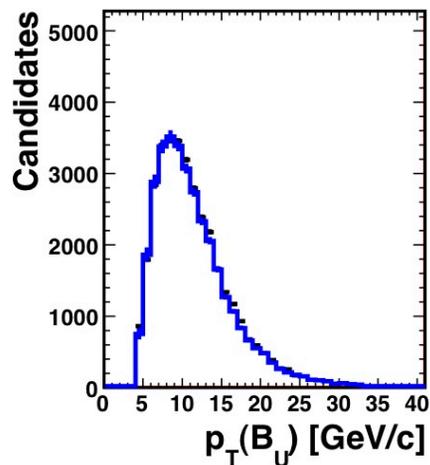
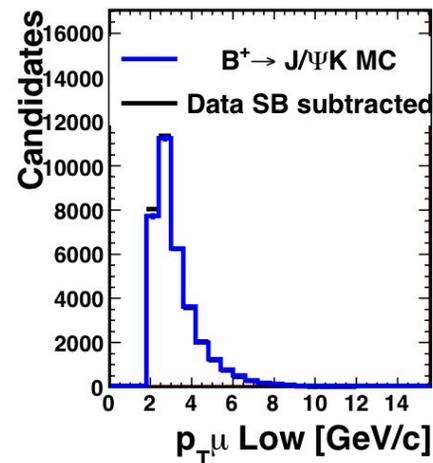
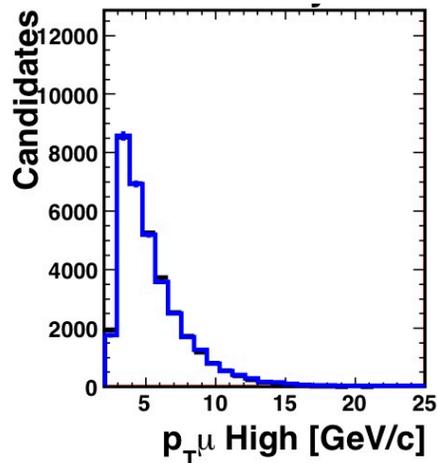
- Employ a Neural Net to optimally combine the information from these variables
 - We **exclude** mass information from the NN
 - M.Feindt and U.Kerzel, NIM A 559, 190 (2006)
- Training
 - Signal: $B_s \rightarrow \mu^+ \mu^-$ MC
 - Background: mass sideband regions
 - Some fraction of each sample set aside to test for bias and overtraining

MC modeling



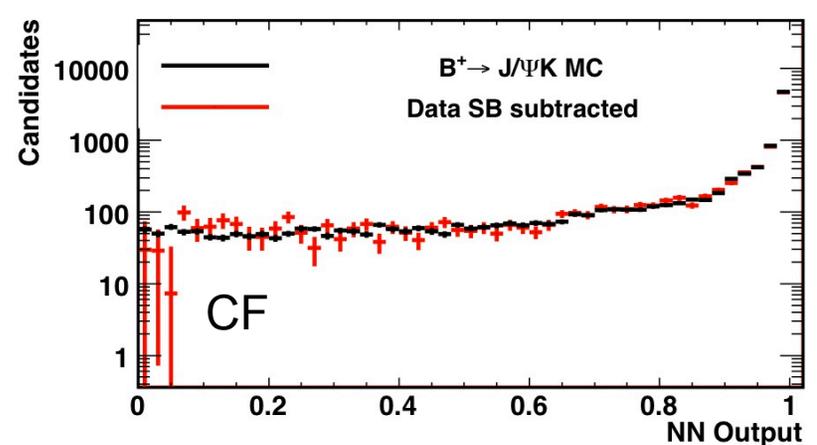
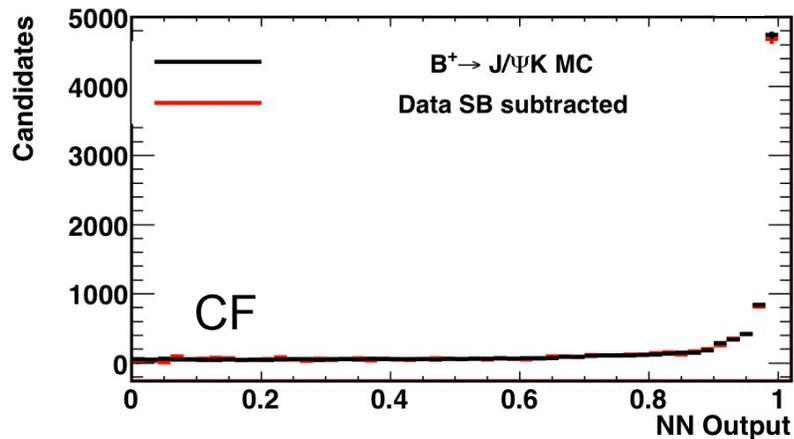
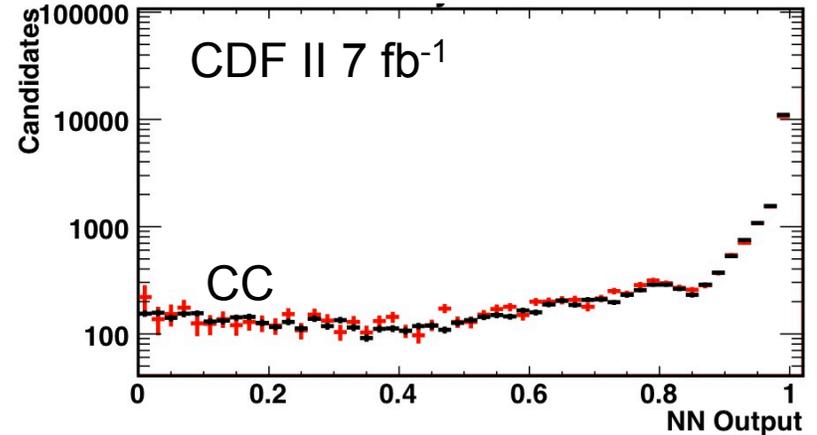
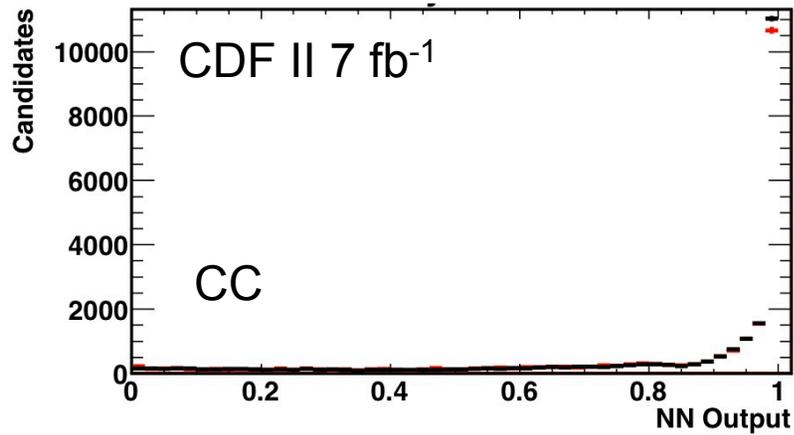
- Verify modeling of signal MC using B^+ events

MC modeling



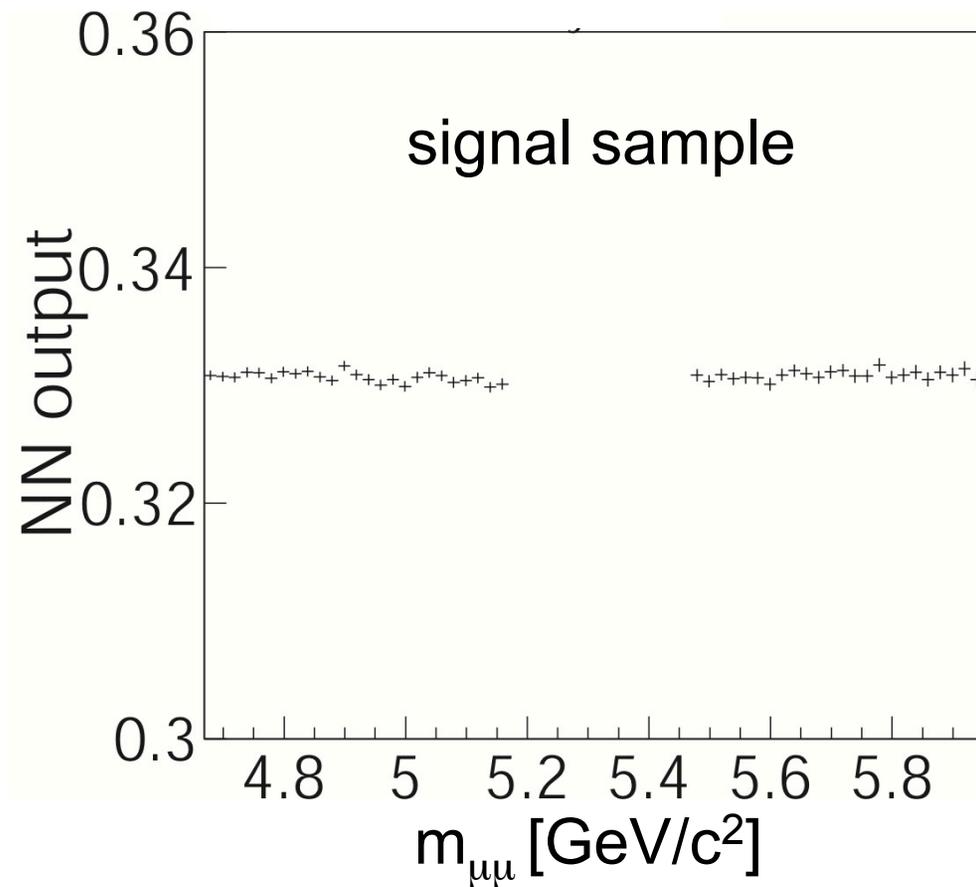
- Verify modeling of signal MC using B⁺ events

MC modeling



- Verify modeling of signal MC using B⁺ events

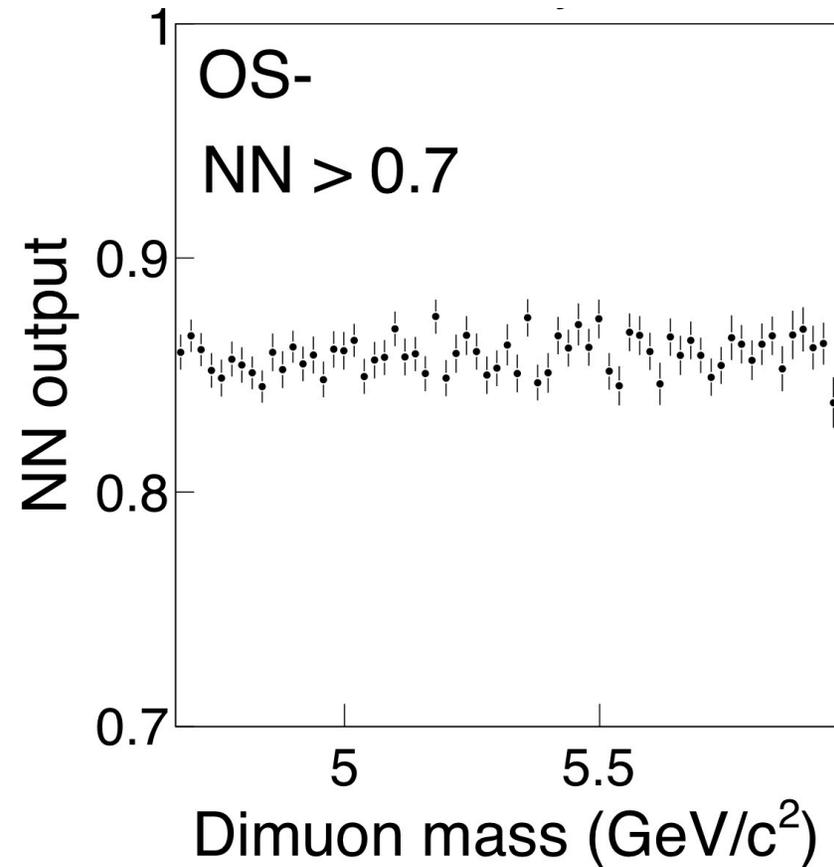
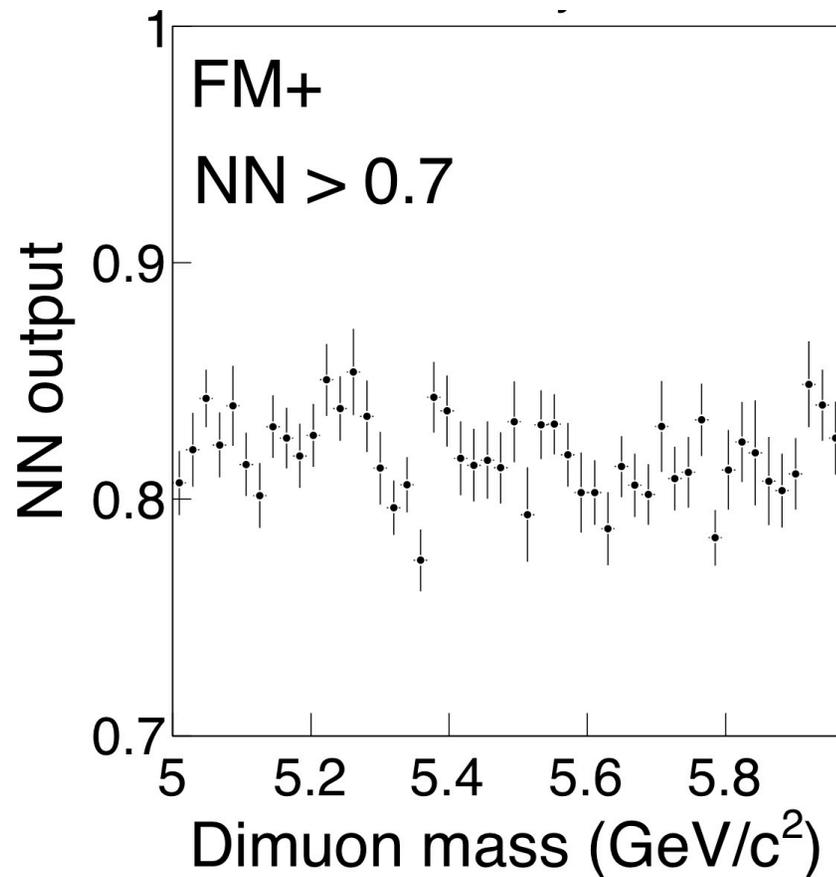
NN correlation with $m_{\mu\mu}$



- Important to verify v_{NN} is independent of $m_{\mu\mu}$

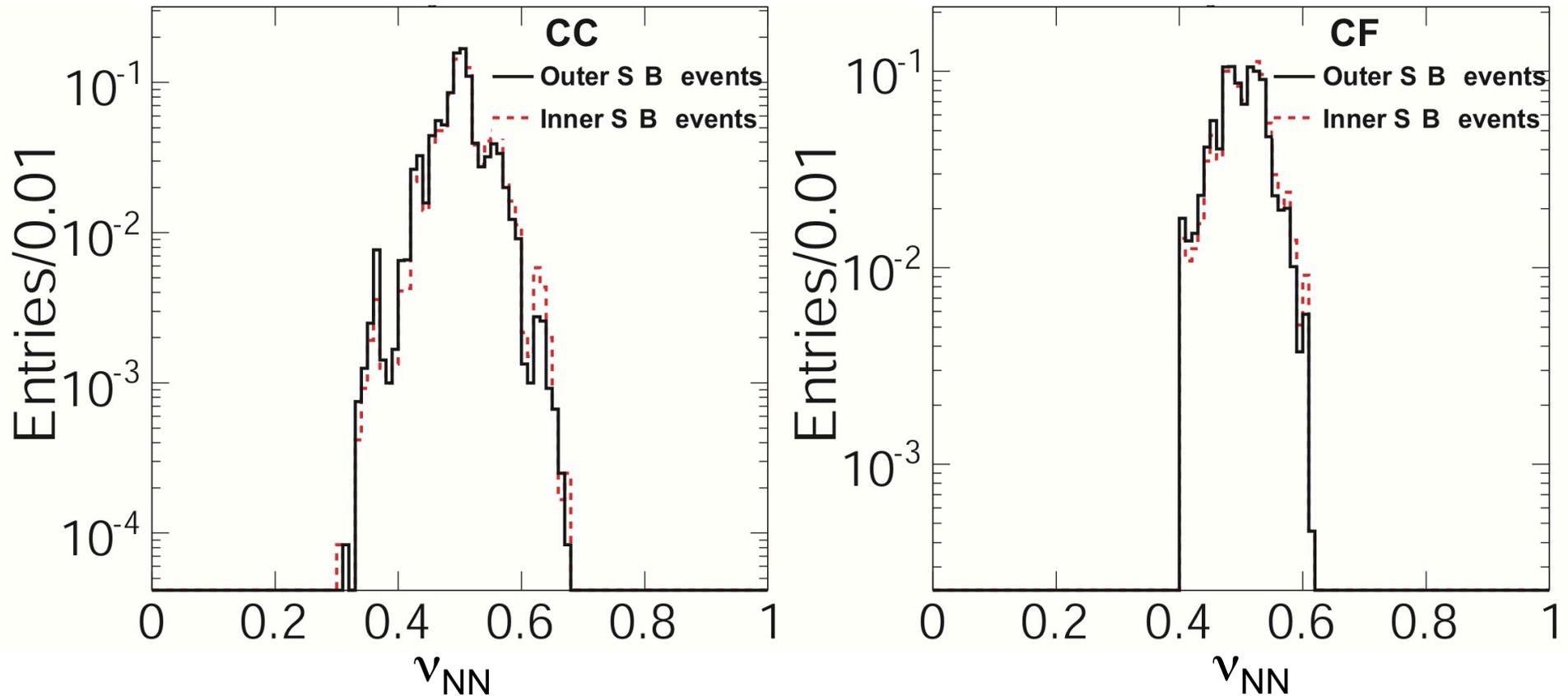
NN correlation with $m_{\mu\mu}$

background dominated control samples



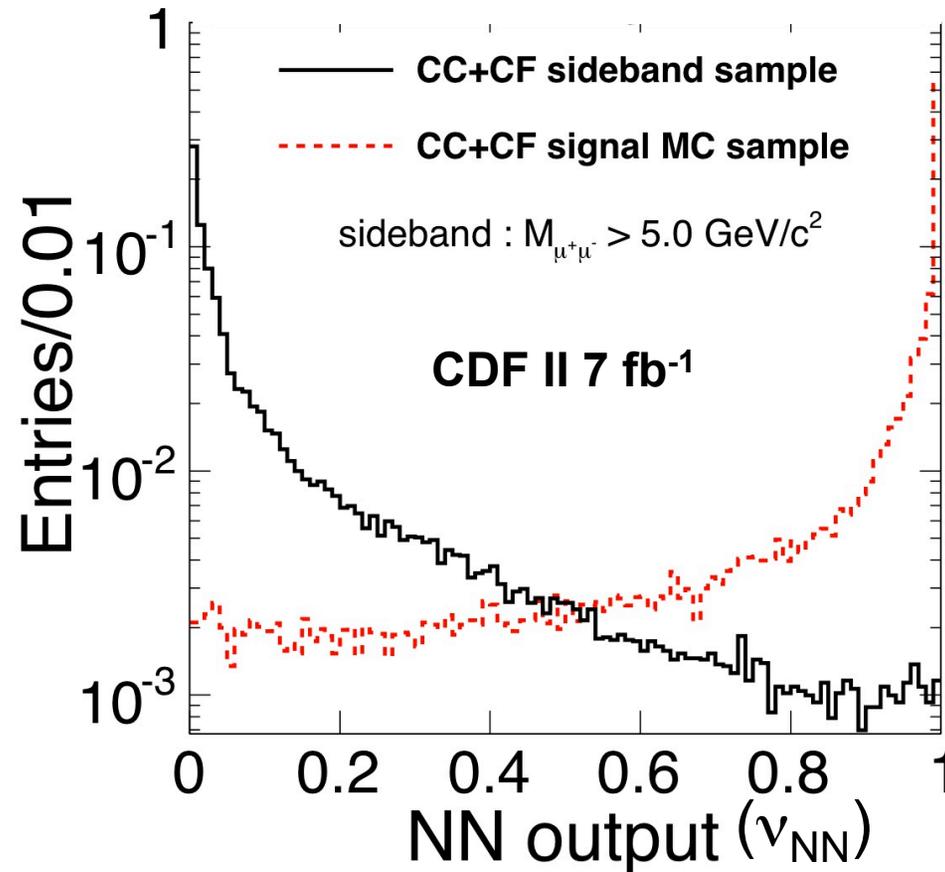
- Important to verify v_{NN} is independent of $m_{\mu\mu}$

NN correlation with $m_{\mu\mu}$



- Important to verify v_{NN} is independent of $m_{\mu\mu}$

NN Separation



- achieves powerful background discrimination

NN Variables

variable	description
$\Delta\Omega$	angle btwn L and p(B) (3D)
Isolation	B candidate isolation
$ d_0(\mu_1) $	muon i.p. where $ d_0(\mu_1) > d_0(\mu_2) $
$ d_0(B) $	B candidate i.p.
L_T/σ_{L_T}	decay length significance in xy plane
$\chi^2(\text{vtx})$	vertex chi-squared vertex
L	decay length (3D)
$\min(p_T(\mu_1), p_T(\mu_2))$	minimum muon p_T
$ d_0(\mu_2) /\sigma_{d_0}$	muon i.p. significance
λ/σ_λ	proper time significance
λ	proper time
$ d_0(\mu_2) $	muon i.p.
$\Delta\Omega_T$	angle btwn L_T and $p_T(B)$ (2D)
$ d_0(\mu_1) /\sigma_{d_0}$	muon i.p. significance

- A ranked list of the 14 variables used in the NN with the most significant variables at the top

Optimization of NN Requirements

- Figure-of-merit: expected limit

$$\langle BF(B_s \rightarrow \mu^+ \mu^-) \rangle = \left(\frac{\langle N_{B_s}^{90\%CL} \rangle}{N_{B^+}} \right) \left(\frac{f_u}{f_s} \right) \left(\frac{\alpha_{B^+} \epsilon_{B^+}}{\alpha_{B_s} \epsilon_{B_s}} \right) BF(B^+ \rightarrow J/\psi K^+)$$

$$\langle N_{B_s}^{90\%CL} \rangle = \sum_{n_{obs}=0}^{\infty} P(n_{obs} | n_{bg}) \cdot N_{B_s}^{90\%CL}(n_{bg}, \Delta_{bg}, \Delta_{\alpha \cdot \epsilon})$$

- Exploit S/B differences in v_{NN} and $m_{\mu\mu}$
 - Bin in $(v_{NN}, m_{\mu\mu})$ and optimize in 2D
- Broad minimum observed
 - Move away from regions with very few SB events
 - Choose something ~middle of minimum

Final NN Requirements

v_{NN} bins		B_s mass bins	B_d mass bins
0.700 – 0.760			
0.760 – 0.850		5310 – 5334	5219 – 5243
0.850 – 0.900	X	5334 – 5358	5243 – 5267
0.900 – 0.940		5358 – 5382	5267 – 5291
0.940 – 0.970		5382 – 5406	5291 – 5315
0.970 – 0.987		5406 – 5430	5315 – 5339
0.987 – 0.995			
0.995 <			

- Require $0.70 < v_{NN}$ ($\epsilon_S \sim 90\%$, $\epsilon_B \sim x\%$)
- Use 40 ($v_{NN}, m_{\mu\mu}$) bins
 - Each for CC and CF channels

Analysis Description

- This is a simple analysis
 - ✓ 1) Find events with 2 muons in them
 - ✓ 2) Identify means to suppress background while keeping as much signal as possible
 - 3) Look for a bump in the $m_{\mu\mu}$ distribution

Analysis Description

- This is a simple analysis
 - ✓ 1) Find events with 2 muons in them
 - ✓ 2) Identify means to suppress background while keeping as much signal as possible
 - 3) Look for a bump in the $m_{\mu\mu}$ distribution
 - Understand signal distributions
 - Understand background yields

Estimating Signal Yield

- Signal yield estimated for each $(v_{\text{NN}}, m_{\mu\mu})$ bin using relative normalization

$$N_{B_{s,d}} = \left(\frac{N_{B^+}}{BF(B^+ \rightarrow J/\psi K^+)} \right) \left(\frac{f_s}{f_u} \right) \left(\frac{\alpha_{B_{s,d}} \epsilon_{B_{s,d}}}{\alpha_{B^+} \epsilon_{B^+}} \right) BF(B_{s,d} \rightarrow \mu^+ \mu^-)$$

$$\epsilon_B \equiv \epsilon_{\text{reco}} \cdot \epsilon_{\text{NN}} \cdot \epsilon_{\text{mass}} = \left(\epsilon_{\text{track}} \cdot \epsilon_{\mu\text{-ID}} \cdot \epsilon_{\text{vertex}} \right) \cdot \epsilon_{\text{NN}} \cdot \epsilon_{\text{mass}}$$

 Varies bin-by-bin

Estimates of SM $B_s \rightarrow \mu^+ \mu^-$ Yields

CC channel:

NN Bin/Mass Bin	5.310-5.334	5.334-5.358	5.358-5.382	5.382-5.406	5.406-5.430
0.700-0.760	0.002 ± 0.000	0.007 ± 0.001	0.011 ± 0.002	0.006 ± 0.001	0.001 ± 0.000
0.760-0.850	0.004 ± 0.001	0.015 ± 0.003	0.020 ± 0.004	0.011 ± 0.002	0.003 ± 0.001
0.850-0.900	0.004 ± 0.001	0.010 ± 0.002	0.014 ± 0.003	0.008 ± 0.001	0.002 ± 0.000
0.900-0.940	0.005 ± 0.001	0.016 ± 0.003	0.023 ± 0.004	0.012 ± 0.002	0.002 ± 0.000
0.940-0.970	0.008 ± 0.001	0.022 ± 0.004	0.032 ± 0.006	0.016 ± 0.003	0.003 ± 0.001
0.970-0.987	0.010 ± 0.002	0.029 ± 0.005	0.041 ± 0.007	0.022 ± 0.004	0.005 ± 0.001
0.987-0.995	0.013 ± 0.002	0.046 ± 0.008	0.062 ± 0.011	0.031 ± 0.006	0.007 ± 0.001
0.995-1.000	0.052 ± 0.009	0.167 ± 0.030	0.227 ± 0.040	0.119 ± 0.021	0.029 ± 0.005

$\Sigma=1.1\text{evt}$

CF channel:

NN Bin/Mass Bin	5.310-5.334	5.334-5.358	5.358-5.382	5.382-5.406	5.406-5.430
0.700-0.760	0.002 ± 0.000	0.006 ± 0.001	0.007 ± 0.001	0.005 ± 0.001	0.001 ± 0.000
0.760-0.850	0.003 ± 0.001	0.012 ± 0.002	0.015 ± 0.003	0.009 ± 0.002	0.002 ± 0.000
0.850-0.900	0.003 ± 0.001	0.009 ± 0.002	0.012 ± 0.002	0.006 ± 0.001	0.001 ± 0.000
0.900-0.940	0.004 ± 0.001	0.012 ± 0.002	0.017 ± 0.003	0.009 ± 0.002	0.002 ± 0.000
0.940-0.970	0.005 ± 0.001	0.015 ± 0.003	0.021 ± 0.004	0.013 ± 0.002	0.003 ± 0.001
0.970-0.987	0.008 ± 0.002	0.026 ± 0.005	0.036 ± 0.007	0.019 ± 0.003	0.005 ± 0.001
0.987-0.995	0.007 ± 0.001	0.021 ± 0.004	0.029 ± 0.005	0.017 ± 0.003	0.004 ± 0.001
0.995-1.000	0.039 ± 0.007	0.116 ± 0.021	0.159 ± 0.029	0.090 ± 0.016	0.023 ± 0.004

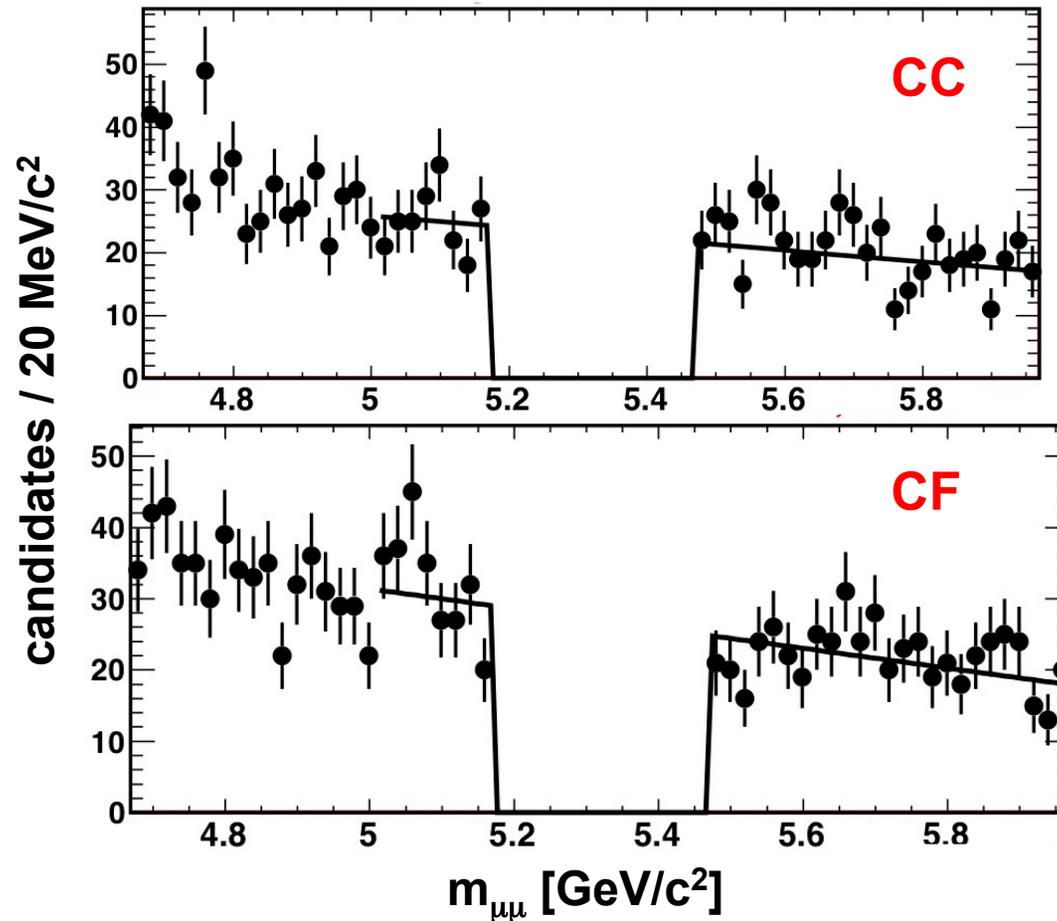
$\Sigma=0.8\text{evt}$

- Number of B_s signal events per bin, BF=SM

Estimating Background Yield

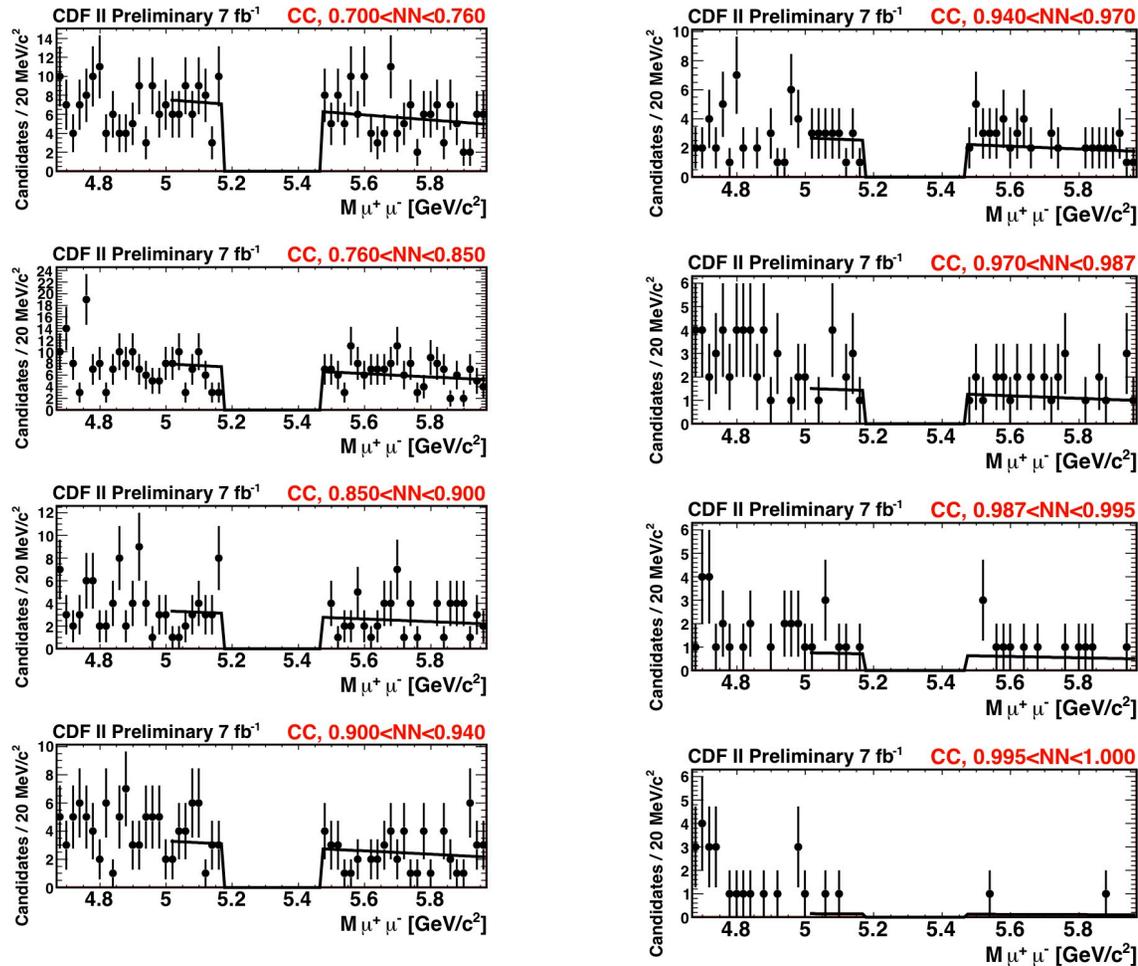
- Only 2 components to the background
 - 1) **Combinatoric**
 - Estimated using mass sidebands
 - 2) **Peaking**
 - Only source from $B \rightarrow h^+ h^-$ ($h = \pi$, or K)
 - Kinematics taken from dedicated MC samples
 - Probability that π, K survive muon ID criteria is taken from D^* tagged $D \rightarrow \pi K$ sample
- Verify accuracy of estimates using background control samples

Estimating Combinatoric Background



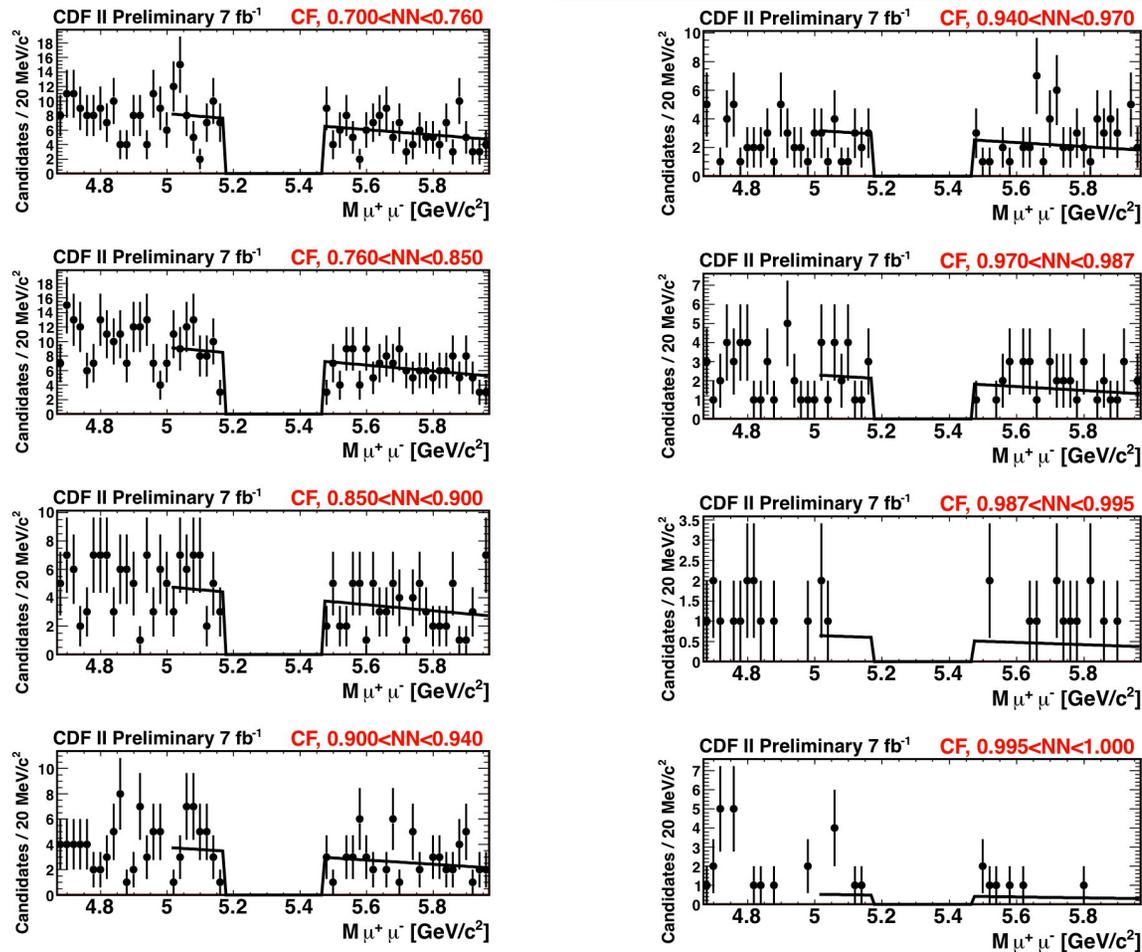
- Slope: from fit to $m_{\mu\mu} > 5 \text{ GeV}/c^2$ for $0.70 < v_{NN}$
 - CC and CF channels separately

Estimating Combinatoric Background



- Normalization determined for each ν_{NN} bin separately (for CC and CF separately)

Estimating Combinatoric Background



- Normalization determined for each ν_{NN} bin separately (for CC and CF separately)

Combinatoric Background Estimate: B_s

Combinatoric background: B_s region

Mass Bin (GeV/c^2) \ NN Bin	5.310—5.334	5.334—5.358	5.358—5.382	5.382—5.406	5.406—5.430
CC					
0.700 < NN < 0.760	8.02 ± 0.62	7.94 ± 0.61	7.87 ± 0.61	7.79 ± 0.60	7.71 ± 0.59
0.760 < NN < 0.850	8.42 ± 0.64	8.34 ± 0.63	8.26 ± 0.62	8.18 ± 0.62	8.10 ± 0.61
0.850 < NN < 0.900	3.55 ± 0.39	3.51 ± 0.39	3.48 ± 0.39	3.44 ± 0.38	3.41 ± 0.38
0.900 < NN < 0.940	3.51 ± 0.39	3.47 ± 0.39	3.44 ± 0.38	3.41 ± 0.38	3.37 ± 0.38
0.940 < NN < 0.970	2.86 ± 0.35	2.83 ± 0.35	2.81 ± 0.34	2.78 ± 0.34	2.75 ± 0.34
0.970 < NN < 0.987	1.61 ± 0.39	1.6 ± 0.39	1.58 ± 0.38	1.57 ± 0.38	1.55 ± 0.37
0.987 < NN < 0.995	0.81 ± 0.23	0.80 ± 0.23	0.79 ± 0.22	0.78 ± 0.22	0.78 ± 0.22
0.995 < NN < 1.000	0.16 ± 0.11	0.16 ± 0.10	0.16 ± 0.10	0.16 ± 0.10	0.16 ± 0.10
CF					
0.700 < NN < 0.760	8.49 ± 0.65	8.39 ± 0.64	8.28 ± 0.63	8.17 ± 0.62	8.07 ± 0.61
0.760 < NN < 0.850	9.45 ± 0.69	9.33 ± 0.68	9.21 ± 0.67	9.1 ± 0.66	8.98 ± 0.65
0.850 < NN < 0.900	4.91 ± 0.48	4.85 ± 0.47	4.79 ± 0.46	4.73 ± 0.46	4.67 ± 0.45
0.900 < NN < 0.940	3.87 ± 0.42	3.82 ± 0.41	3.77 ± 0.41	3.73 ± 0.40	3.68 ± 0.40
0.940 < NN < 0.970	3.29 ± 0.38	3.25 ± 0.38	3.21 ± 0.37	3.17 ± 0.37	3.12 ± 0.36
0.970 < NN < 0.987	2.37 ± 0.53	2.34 ± 0.53	2.31 ± 0.52	2.28 ± 0.52	2.25 ± 0.51
0.987 < NN < 0.995	0.67 ± 0.20	0.66 ± 0.20	0.65 ± 0.20	0.64 ± 0.19	0.63 ± 0.19
0.995 < NN < 1.000	0.54 ± 0.27	0.53 ± 0.27	0.53 ± 0.27	0.52 ± 0.26	0.51 ± 0.26

- uncertainty includes: slope & normalization uncertainties as well as variations in fit function and range

Combinatoric Background Estimate: B_d

Combinatoric background: B_d region

Mass Bin (GeV/c^2) \ NN Bin	5.219–5.243	5.243–5.267	5.267–5.291	5.291–5.315	5.315–5.339
CC					
0.700 < NN < 0.760	8.31 ± 0.64	8.24 ± 0.63	8.16 ± 0.63	8.08 ± 0.62	8.00 ± 0.62
0.760 < NN < 0.850	8.73 ± 0.66	8.65 ± 0.65	8.57 ± 0.65	8.49 ± 0.64	8.41 ± 0.63
0.850 < NN < 0.900	3.68 ± 0.41	3.64 ± 0.40	3.61 ± 0.40	3.57 ± 0.40	3.54 ± 0.39
0.900 < NN < 0.940	3.63 ± 0.40	3.60 ± 0.40	3.57 ± 0.40	3.53 ± 0.39	3.50 ± 0.39
0.940 < NN < 0.970	2.97 ± 0.36	2.94 ± 0.36	2.91 ± 0.36	2.88 ± 0.35	2.86 ± 0.35
0.970 < NN < 0.987	1.67 ± 0.40	1.66 ± 0.40	1.64 ± 0.40	1.62 ± 0.39	1.61 ± 0.39
0.987 < NN < 0.995	0.84 ± 0.24	0.83 ± 0.23	0.82 ± 0.23	0.81 ± 0.23	0.80 ± 0.23
0.995 < NN < 1.000	0.17 ± 0.11	0.17 ± 0.11	0.16 ± 0.11	0.16 ± 0.11	0.16 ± 0.11
CF					
0.700 < NN < 0.760	8.89 ± 0.68	8.78 ± 0.67	8.68 ± 0.66	8.57 ± 0.65	8.47 ± 0.65
0.760 < NN < 0.850	9.89 ± 0.72	9.78 ± 0.71	9.66 ± 0.70	9.54 ± 0.69	9.42 ± 0.69
0.850 < NN < 0.900	5.14 ± 0.50	5.08 ± 0.49	5.02 ± 0.49	4.96 ± 0.48	4.90 ± 0.47
0.900 < NN < 0.940	4.05 ± 0.44	4.00 ± 0.43	3.96 ± 0.43	3.91 ± 0.42	3.86 ± 0.42
0.940 < NN < 0.970	3.44 ± 0.40	3.40 ± 0.40	3.36 ± 0.39	3.32 ± 0.39	3.28 ± 0.38
0.970 < NN < 0.987	2.48 ± 0.56	2.45 ± 0.55	2.43 ± 0.55	2.40 ± 0.54	2.37 ± 0.53
0.987 < NN < 0.995	0.70 ± 0.21	0.69 ± 0.21	0.68 ± 0.21	0.67 ± 0.20	0.66 ± 0.20
0.995 < NN < 1.000	0.57 ± 0.29	0.56 ± 0.28	0.55 ± 0.28	0.55 ± 0.28	0.54 ± 0.27

- uncertainty includes: slope & normalization uncertainties as well as variations in fit function and range

Estimating Peaking Backgrounds

- Backgrounds which peak near the mass signal region will not be included in the combinatoric background estimates
- Only relevant sources of such events:
 - $B_d \rightarrow K^+\pi^-, \pi^+\pi^-, K^+K^-$
 - $B_s \rightarrow K^+K^-, \pi^+K, \pi^+\pi^-$
- These are suppressed because:
 - BF are small (10^{-5} to $<10^{-7}$)
 - $m_{\mu\mu}$ calculated assuming muon mass
 - Probability($\pi/K \rightarrow$ fake μ) is small ($<1 \times 10^{-2}$)

Estimating Peaking Backgrounds

- To estimate yield, solve for $N_{B_{s,d}}$:

$$\frac{BF(B_{s,d} \rightarrow h^+ h'^-)}{BF(B^+ \rightarrow J/\psi K^+)} = \left(\frac{N_{Bhh}}{N_{B^+}} \right) \left(\frac{f_u}{f_{s,d}} \right) \left(\frac{\alpha_{B^+} \epsilon_{B^+}}{\alpha_{B_{s,d}} \epsilon_{B_{s,d}}} \right)$$

- Obtain $\alpha^* \epsilon$:

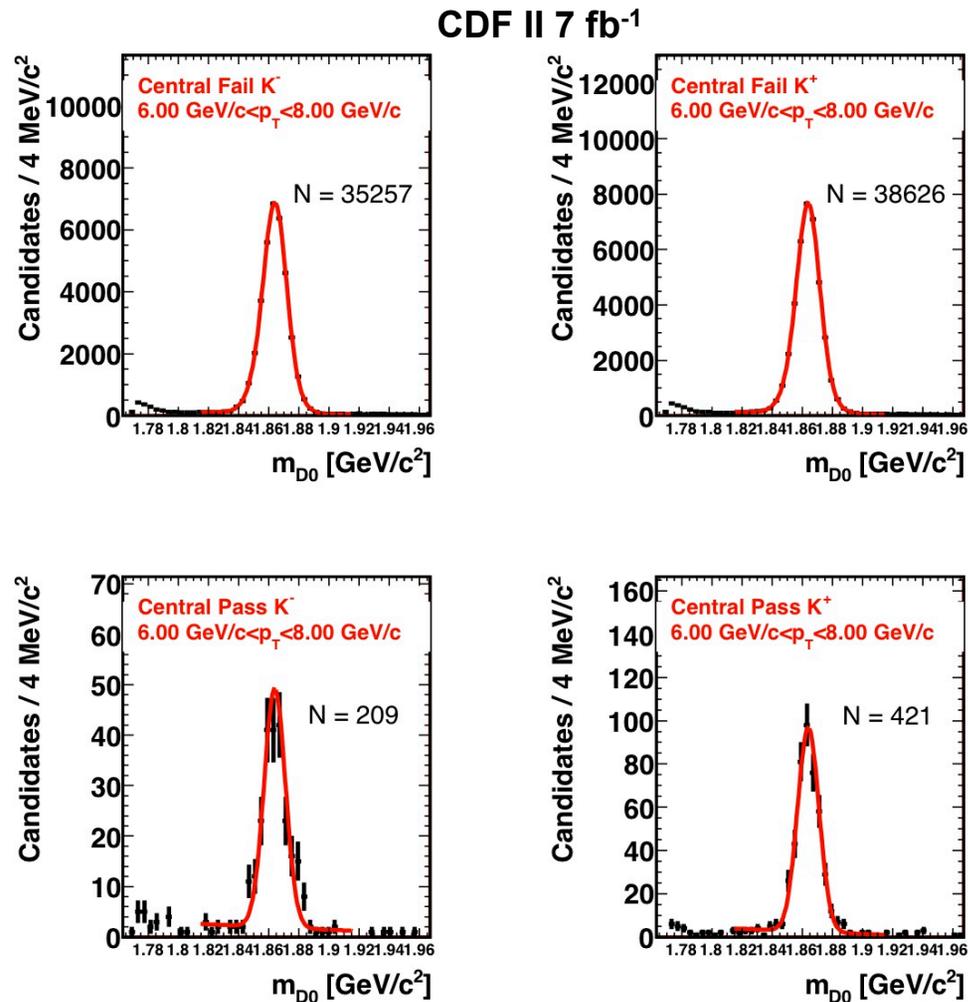
α_B \equiv geometric and kinematic acceptance of trigger

$$\epsilon_B \equiv \epsilon_{\text{reco}} \cdot \epsilon_{\text{NN}} \cdot \epsilon_{\text{mass}} = \left(\epsilon_{\text{track}} \cdot \epsilon_{\mu\text{-ID}} \cdot \epsilon_{\text{vertex}} \right) \cdot \epsilon_{\text{NN}} \cdot \epsilon_{\text{mass}}$$

 same as $B \rightarrow \mu^+ \mu^-$

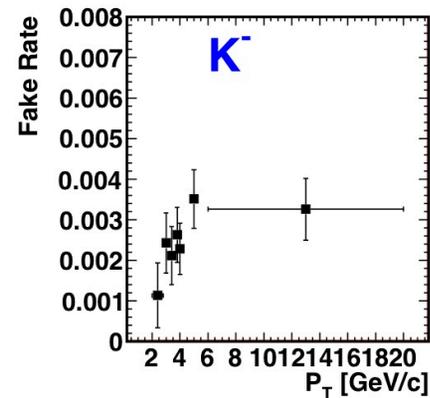
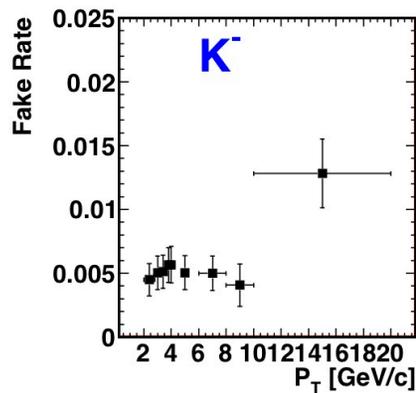
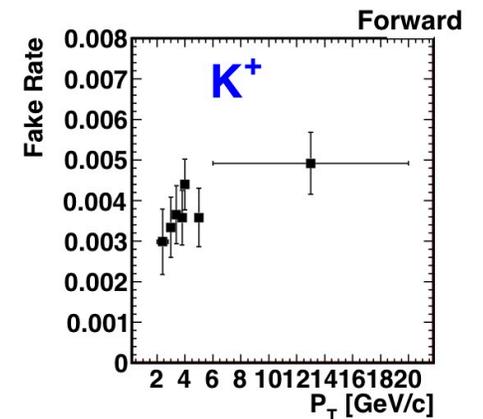
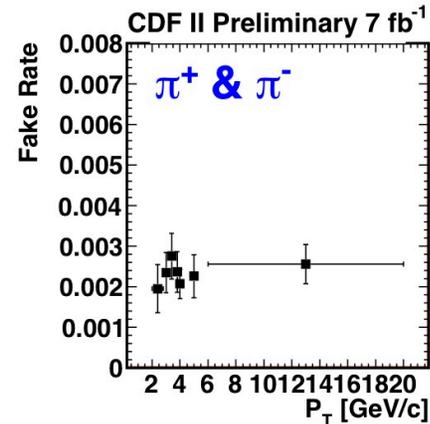
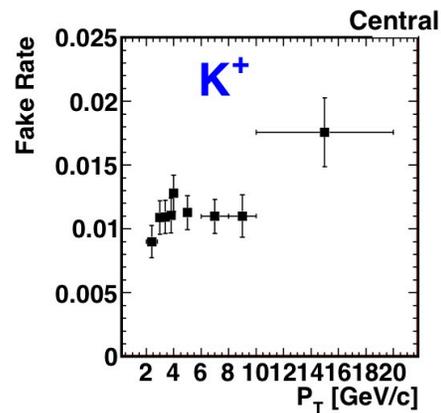
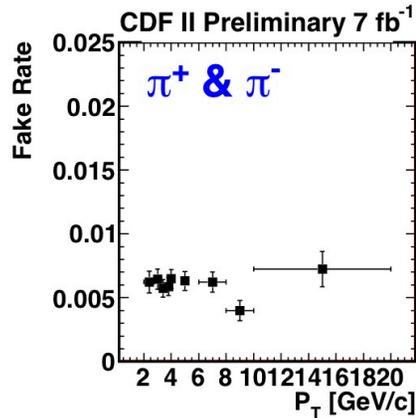
 requires special treatment

Estimating Peaking Backgrounds



- $\epsilon_{\mu\text{-fake}}$ is taken from D^* tagged $D^+ \rightarrow \pi^+ K^-$ data

Estimating Peaking Backgrounds



- $\epsilon_{\mu\text{-fake}}$ parameterized (p_T , $\mathcal{L}_{\text{inst}}$)
 - Separately for $\pi^{+/-}$, K^+ , and K^-

B→hh Background Estimate: B_s

B→hh background: B_s region

Mass Bin (GeV/c ²) \ NN Bin	5.310–5.334	5.334–5.358	5.358–5.382	5.382–5.406	5.406–5.430
CC					
0.700 < NN < 0.760	0.002 ± 0.001	0.001 ± < 0.001	–	–	–
0.760 < NN < 0.850	0.004 ± 0.001	0.002 ± < 0.001	0.001 ± < 0.001	–	–
0.850 < NN < 0.900	0.004 ± 0.001	0.001 ± < 0.001	–	–	–
0.900 < NN < 0.940	0.005 ± 0.001	0.002 ± < 0.001	0.001 ± < 0.001	–	–
0.940 < NN < 0.970	0.008 ± 0.002	0.002 ± 0.001	0.001 ± < 0.001	–	–
0.970 < NN < 0.987	0.010 ± 0.002	0.003 ± 0.001	0.001 ± < 0.001	–	–
0.987 < NN < 0.995	0.013 ± 0.003	0.005 ± 0.001	0.002 ± 0.001	0.001 ± < 0.001	–
0.995 < NN < 1.000	0.052 ± 0.012	0.019 ± 0.005	0.006 ± 0.003	0.002 ± 0.001	0.001 ± < 0.001
CF					
0.700 < NN < 0.760	0.001 ± < 0.001	–	–	–	–
0.760 < NN < 0.850	0.001 ± < 0.001	0.001 ± < 0.001	–	–	–
0.850 < NN < 0.900	0.001 ± < 0.001	–	–	–	–
0.900 < NN < 0.940	0.002 ± < 0.001	0.001 ± < 0.001	–	–	–
0.940 < NN < 0.970	0.002 ± 0.001	0.001 ± < 0.001	–	–	–
0.970 < NN < 0.987	0.003 ± 0.001	0.001 ± < 0.001	0.001 ± < 0.001	–	–
0.987 < NN < 0.995	0.003 ± 0.001	0.001 ± < 0.001	–	–	–
0.995 < NN < 1.000	0.015 ± 0.004	0.006 ± 0.002	0.002 ± 0.001	0.001 ± < 0.001	–

- Uncertainty includes: BF and fake-μ rate uncertainties (statistics of D* subsamples, D⁰ fits, residual luminosity dependence)

B→hh Background Estimate: B_d

B→hh background: B_d region

Mass Bin (GeV/c ²) \ NN Bin	5.219–5.243	5.243–5.267	5.267–5.291	5.291–5.315	5.315–5.339
CC					
0.700 < NN < 0.760	0.011 ± 0.003	0.010 ± 0.002	0.008 ± 0.002	0.004 ± 0.001	0.002 ± < 0.001
0.760 < NN < 0.850	0.019 ± 0.006	0.019 ± 0.004	0.014 ± 0.003	0.008 ± 0.002	0.003 ± 0.001
0.850 < NN < 0.900	0.016 ± 0.005	0.013 ± 0.003	0.010 ± 0.002	0.006 ± 0.001	0.002 ± < 0.001
0.900 < NN < 0.940	0.022 ± 0.006	0.021 ± 0.005	0.016 ± 0.004	0.009 ± 0.002	0.003 ± 0.001
0.940 < NN < 0.970	0.034 ± 0.010	0.028 ± 0.006	0.022 ± 0.005	0.012 ± 0.003	0.004 ± 0.001
0.970 < NN < 0.987	0.042 ± 0.013	0.037 ± 0.009	0.029 ± 0.007	0.016 ± 0.004	0.006 ± 0.001
0.987 < NN < 0.995	0.060 ± 0.018	0.059 ± 0.014	0.043 ± 0.010	0.024 ± 0.006	0.008 ± 0.002
0.995 < NN < 1.000	0.231 ± 0.068	0.211 ± 0.049	0.157 ± 0.036	0.090 ± 0.022	0.035 ± 0.008
CF					
0.700 < NN < 0.760	0.003 ± 0.001	0.003 ± 0.001	0.002 ± 0.001	0.001 ± < 0.001	–
0.760 < NN < 0.850	0.005 ± 0.002	0.006 ± 0.002	0.004 ± 0.001	0.003 ± 0.001	0.001 ± < 0.001
0.850 < NN < 0.900	0.004 ± 0.001	0.004 ± 0.001	0.004 ± 0.001	0.002 ± 0.001	0.001 ± < 0.001
0.900 < NN < 0.940	0.006 ± 0.002	0.006 ± 0.002	0.005 ± 0.001	0.003 ± 0.001	0.001 ± < 0.001
0.940 < NN < 0.970	0.008 ± 0.003	0.008 ± 0.002	0.006 ± 0.002	0.004 ± 0.001	0.001 ± < 0.001
0.970 < NN < 0.987	0.012 ± 0.004	0.013 ± 0.003	0.011 ± 0.003	0.006 ± 0.002	0.002 ± 0.001
0.987 < NN < 0.995	0.010 ± 0.003	0.011 ± 0.003	0.009 ± 0.002	0.005 ± 0.001	0.002 ± < 0.001
0.995 < NN < 1.000	0.057 ± 0.018	0.061 ± 0.015	0.048 ± 0.012	0.028 ± 0.007	0.011 ± 0.003

- Uncertainty includes: BF and fake-μ rate uncertainties (statistics of D* subsamples, D⁰ fits, residual luminosity dependence)

Background Summary: B_s Search

Combinatoric:

NN Bin	CC	CF
$0.700 < NN < 0.970$	129.2 ± 6.5	146.3 ± 7.0
$0.970 < NN < 0.987$	7.9 ± 1.9	11.6 ± 1.8
$0.987 < NN < 0.995$	4.0 ± 1.1	3.3 ± 1.0
$0.995 < NN < 1.000$	0.79 ± 0.52	2.6 ± 1.5

$B \rightarrow hh$:

NN Bin	CC	CF
$0.700 < NN < 0.970$	0.03 ± 0.01	$0.01 \pm < 0.01$
$0.970 < NN < 0.987$	$0.01 \pm < 0.01$	$0.01 \pm < 0.01$
$0.987 < NN < 0.995$	$0.02 \pm < 0.01$	$0.01 \pm < 0.01$
$0.995 < NN < 1.000$	0.08 ± 0.02	0.03 ± 0.01

- Focus on 3 most sensitive v_{NN} bins
 - integrating over $m_{\mu\mu}$ bins, first 5 v_{NN} bins

Background Summary: B_d Search

Combinatoric:

NN Bin	CC	CF
$0.700 < NN < 0.970$	134.0 ± 6.6	153.4 ± 7.3
$0.970 < NN < 0.987$	8.2 ± 2.0	12.1 ± 1.9
$0.987 < NN < 0.995$	4.1 ± 1.2	3.4 ± 1.1
$0.995 < NN < 1.000$	0.8 ± 0.5	2.8 ± 1.6

$B \rightarrow hh$:

NN Bin	CC	CF
$0.700 < NN < 0.970$	0.31 ± 0.08	0.09 ± 0.02
$0.970 < NN < 0.987$	0.13 ± 0.03	0.05 ± 0.01
$0.987 < NN < 0.995$	0.19 ± 0.05	0.04 ± 0.01
$0.995 < NN < 1.000$	0.72 ± 0.20	0.20 ± 0.05

- Focus on 3 most sensitive v_{NN} bins
 - integrating over $m_{\mu\mu}$ bins, first 5 v_{NN} bins

Cross-check Background Methodology

- We employ these data samples:
 - Opposite sign $\mu\mu$
 - $L > 0$ (OS+) this is our signal sample
 - $L < 0$ (OS-) bgd control sample
 - Dominated by combinatoric background
 - Kinematics very similar to signal sample
 - Same sign $\mu\mu$ (SS) bgd control sample
 - Dominated by combinatoric background
 - Different kinematics from signal sample
 - Fake- μ enhanced sample (FM) bgd control sample
(require ≥ 1 muon to fail μ -ID requirements)
 - Large $B \rightarrow hh$ contribution
 - Different kinematics from signal sample

Cross-check Background Methodology

sample	NN cut	CC		
		pred	obsv	prob(%)
OS-	0.700 < NN < 0.760	217.4 ± (12.5)	203	77.7
	0.760 < NN < 0.850	262.0 ± (14.1)	213	99.1
	0.850 < NN < 0.900	117.9 ± (8.6)	120	44.7
	0.900 < NN < 0.940	112.1 ± (8.4)	116	39.4
	0.940 < NN < 0.970	112.7 ± (8.4)	108	64.2
	0.970 < NN < 0.987	80.2 ± (6.9)	75	68.3
	0.987 < NN < 0.995	67.6 ± (6.3)	41	99.8
	0.995 < NN < 1.000	32.5 ± (4.2)	35	37.5
SS+	0.700 < NN < 0.760	3.0 ± (0.9)	3	55.0
	0.760 < NN < 0.850	3.3 ± (1.0)	5	25.4
	0.850 < NN < 0.900	1.5 ± (0.7)	2	43.2
	0.900 < NN < 0.940	0.9 ± (0.5)	1	56.8
	0.940 < NN < 0.970	1.2 ± (0.6)	1	65.9
	0.970 < NN < 0.987	1.5 ± (0.7)	2	43.2
	0.987 < NN < 0.995	0.3 ± (0.3)	0	74.1
	0.995 < NN < 1.000	0.3 ± (0.3)	0	74.1
SS-	0.700 < NN < 0.760	5.7 ± (1.3)	8	23.7
	0.760 < NN < 0.850	8.4 ± (1.6)	7	69.8
	0.850 < NN < 0.900	3.3 ± (1.0)	6	14.3
	0.900 < NN < 0.940	2.4 ± (0.8)	4	24.0
	0.940 < NN < 0.970	2.4 ± (0.8)	4	24.0
	0.970 < NN < 0.987	2.1 ± (0.8)	0	12.2
	0.987 < NN < 0.995	1.5 ± (0.7)	0	22.3
	0.995 < NN < 1.000	0.3 ± (0.3)	1	30.0
FM+	0.700 < NN < 0.760	118.3 ± (8.6)	136	11.1
	0.760 < NN < 0.850	110.5 ± (8.3)	121	22.3
	0.850 < NN < 0.900	52.0 ± (5.4)	37	96.3
	0.900 < NN < 0.940	37.3 ± (4.5)	37	53.0
	0.940 < NN < 0.970	20.1 ± (3.3)	20	52.3
	0.970 < NN < 0.987	8.3 ± (2.0)	6	77.1
	0.987 < NN < 0.995	8.7 ± (2.0)	3	97.5
	0.995 < NN < 1.000	20.8 ± (3.5)	24	30.7

- Compare #observed to #predicted in all 80 (v_{NN} , $m_{\mu\mu}$)bins across all background dominated control samples

Cross-check Background Methodology

sample	NN cut	CF		
		pred	obsv	prob(%)
OS-	0.700 < NN < 0.760	209.3 ± (12.0)	187	88.8
	0.760 < NN < 0.850	332.3 ± (16.3)	325	62.0
	0.850 < NN < 0.900	146.7 ± (9.7)	144	57.7
	0.900 < NN < 0.940	144.2 ± (9.6)	139	63.9
	0.940 < NN < 0.970	128.6 ± (8.9)	112	88.4
	0.970 < NN < 0.987	92.8 ± (7.4)	89	63.0
	0.987 < NN < 0.995	45.4 ± (5.0)	55	14.0
	0.995 < NN < 1.000	38.3 ± (4.5)	37	58.2
SS+	0.700 < NN < 0.760	0.3 ± (0.3)	1	30.0
	0.760 < NN < 0.850	4.2 ± (1.1)	4	57.8
	0.850 < NN < 0.900	0.3 ± (0.3)	3	1.3
	0.900 < NN < 0.940	0.6 ± (0.4)	1	45.4
	0.940 < NN < 0.970	0.9 ± (0.5)	1	56.8
	0.970 < NN < 0.987	0.6 ± (0.4)	0	54.9
	0.987 < NN < 0.995	0.5 ± (0.4)	0	60.1
	0.995 < NN < 1.000	0.3 ± (0.3)	1	30.0
SS-	0.700 < NN < 0.760	4.2 ± (1.1)	4	57.8
	0.760 < NN < 0.850	5.1 ± (1.2)	7	27.1
	0.850 < NN < 0.900	2.7 ± (0.9)	2	71.0
	0.900 < NN < 0.940	0.9 ± (0.5)	4	2.8
	0.940 < NN < 0.970	3.0 ± (0.9)	1	92.3
	0.970 < NN < 0.987	2.4 ± (0.8)	5	12.2
	0.987 < NN < 0.995	0.6 ± (0.4)	0	54.9
	0.995 < NN < 1.000	1.8 ± (0.7)	0	16.5
FM+	0.700 < NN < 0.760	54.8 ± (5.6)	66	12.7
	0.760 < NN < 0.850	66.3 ± (6.2)	57	83.1
	0.850 < NN < 0.900	33.7 ± (4.3)	25	90.3
	0.900 < NN < 0.940	17.4 ± (3.1)	26	6.6
	0.940 < NN < 0.970	9.5 ± (2.2)	15	10.2
	0.970 < NN < 0.987	5.3 ± (1.7)	9	13.4
	0.987 < NN < 0.995	2.7 ± (1.2)	3	49.3
	0.995 < NN < 1.000	2.1 ± (1.0)	8	0.7

- Compare #observed to #predicted in all 80 (ν_{NN} , $m_{\mu\mu}$)bins across all background dominated control samples

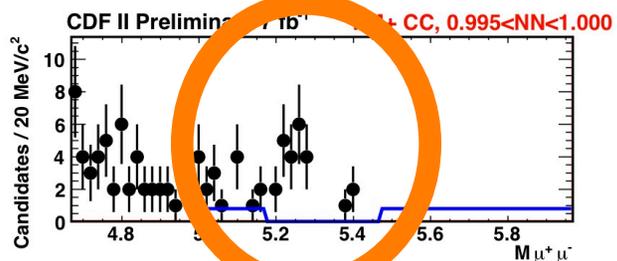
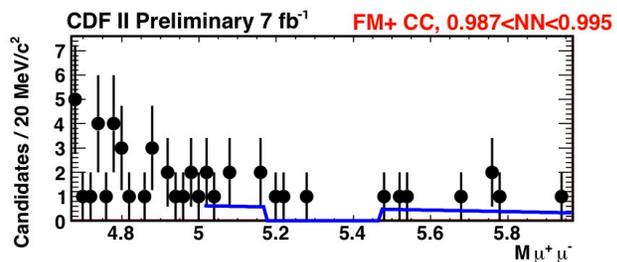
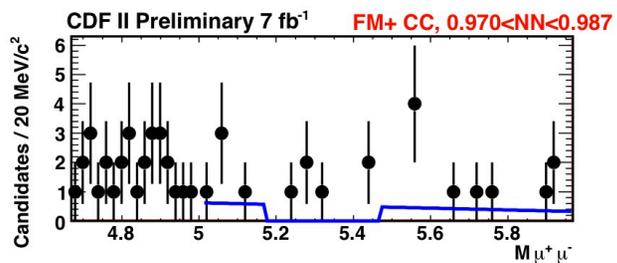
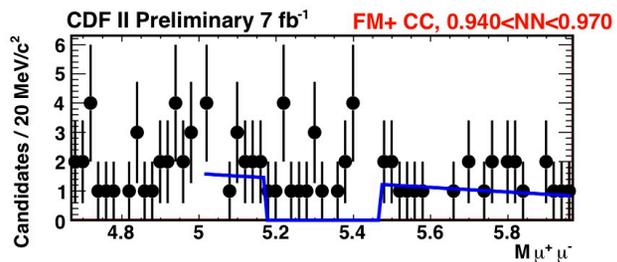
Cross-check Background Methodology

Control Sample	Prediction	Nobs	Prob($N \geq Nobs$)
OS-	2140.0 ± 53.9	1999	98%
SS+	19.7 ± 3.4	25	19%
SS-	46.8 ± 5.3	53	25%
FM+	567.8 ± 25.4	593	24%
Sum	2774.3 ± 59.9	2670	91%

Table: A comparison of the predicted and observed number of events in an extended signal mass region for all NN cuts for all the control samples. This is used as a cross check of the background estimates.

- Integrating over all bins in each sample

Cross-check Background Methodology



- Observe $B \rightarrow hh$ in predicted place at predicted rate



What?

Sensitivity

- CDF expected sensitivity $BF(B \rightarrow \mu\mu)$:

$$B_d : 4.6 \text{ E-9 @ 95\% CL}$$

$$3.6 \text{ E-9 @ 90\% CL}$$

$$B_s : 1.5 \text{ E-8 @ 95\% CL}$$

$$1.1 \text{ E-8 @ 90\% CL}$$

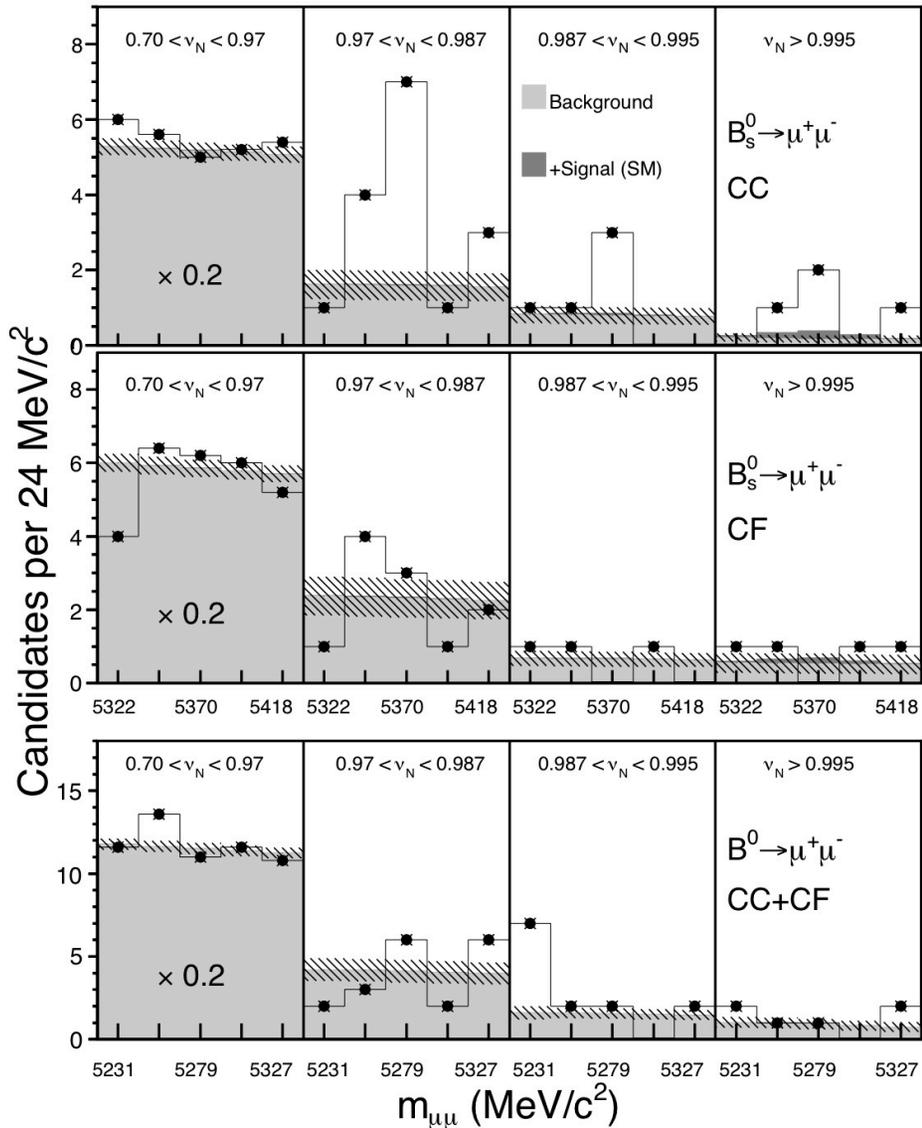
- Among world's best in both channels

- CMS: $B_d = 4.8\text{E-9}$ $B_s = 1.8\text{E-8}$ @ 95% CL

- LHCb: $B_d = 3.1\text{E-9}$ $B_s = 1.0\text{E-8}$

(all of these derived assuming background-only)

Result



- Comparison of data to background prediction in the $(v_{NN}, m_{\mu\mu})$ bins from the optimization
- Only showing systematic uncertainties

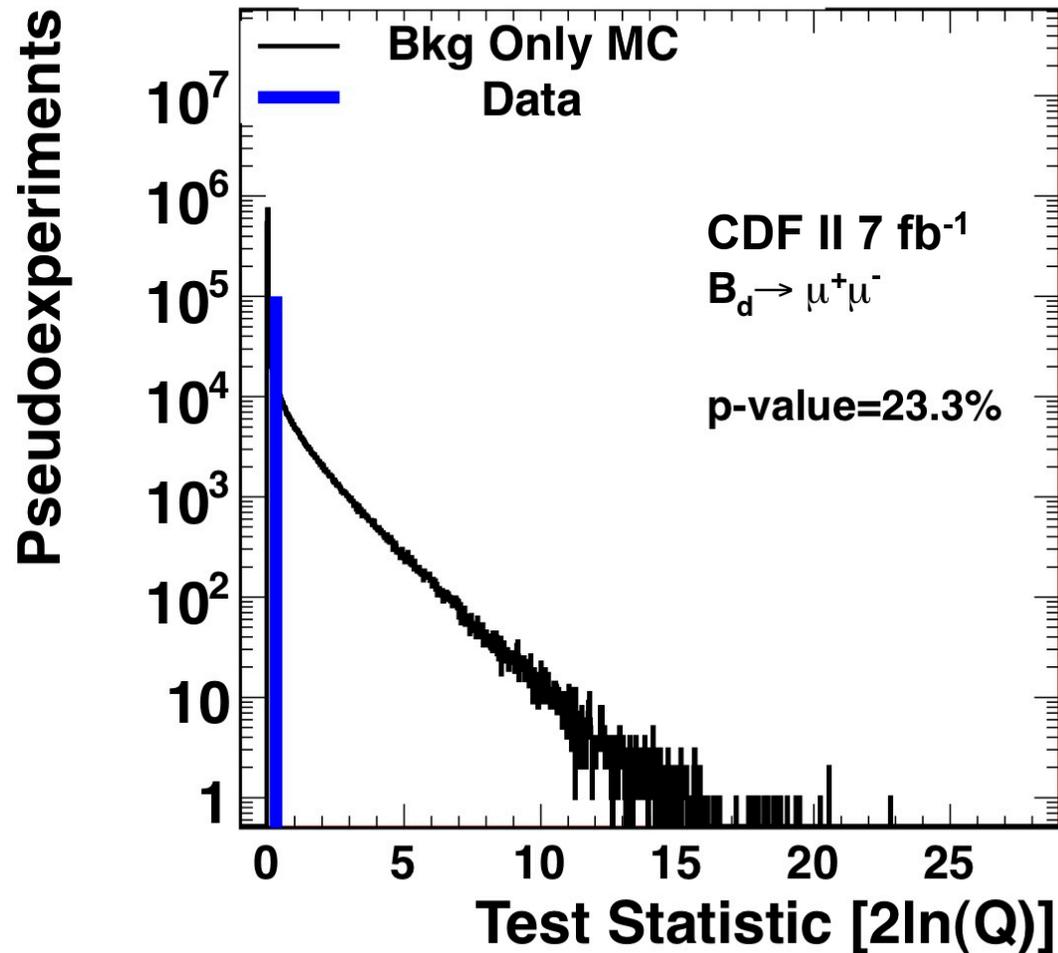
Likelihood ratio

- We fit the data twice
 - 1) Assuming signal = 0
 - 2) Leave signal BF floating
 - Then take ratio: $Q = L(s+b) / L(b)$
- The likelihood:

$$L = \left[\prod_{i=1}^{\text{Nbins}} P(n_{obs}^i | s_i + b_i) \right] \prod_{j=1}^{\text{Nsyst}} G(x_j | \sigma_j)$$

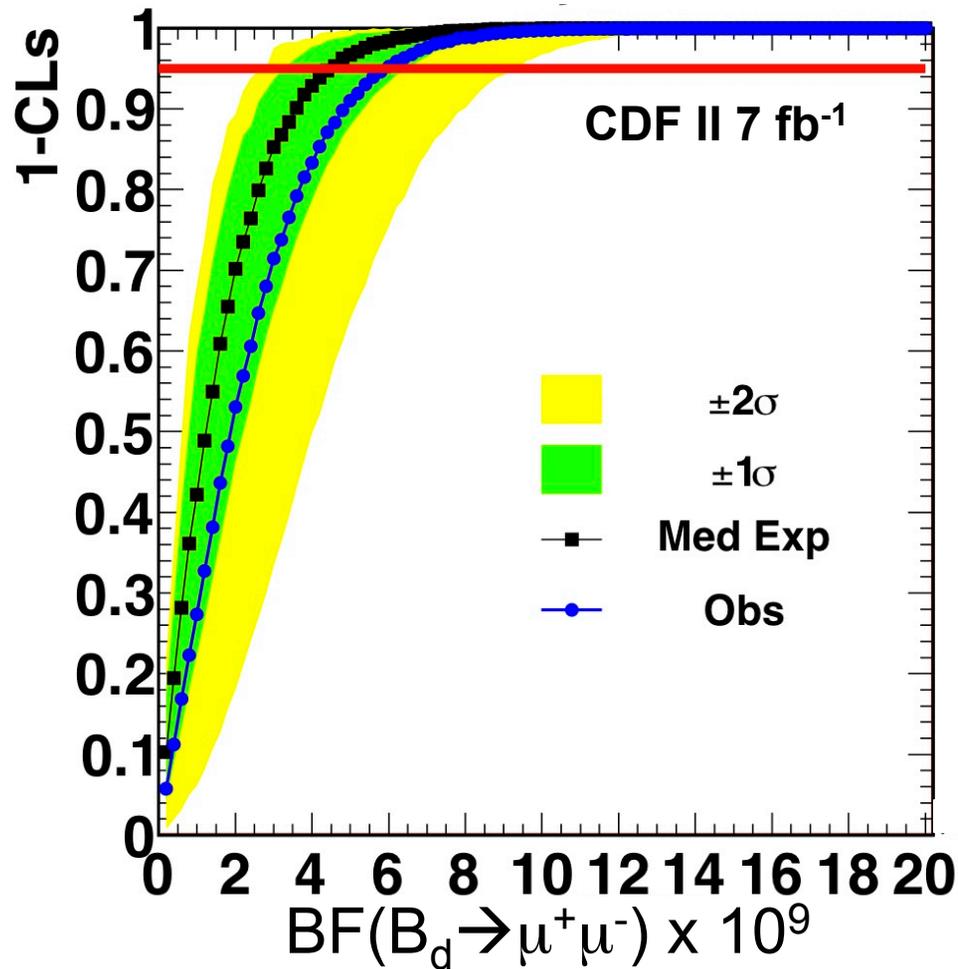
$$s_i = F(BF(B_s \rightarrow \mu^+ \mu^-), x_j), \quad b_i = F(x_j)$$

Result: B_d



- p-value using background-only pseudo-exp.

Result: B_d



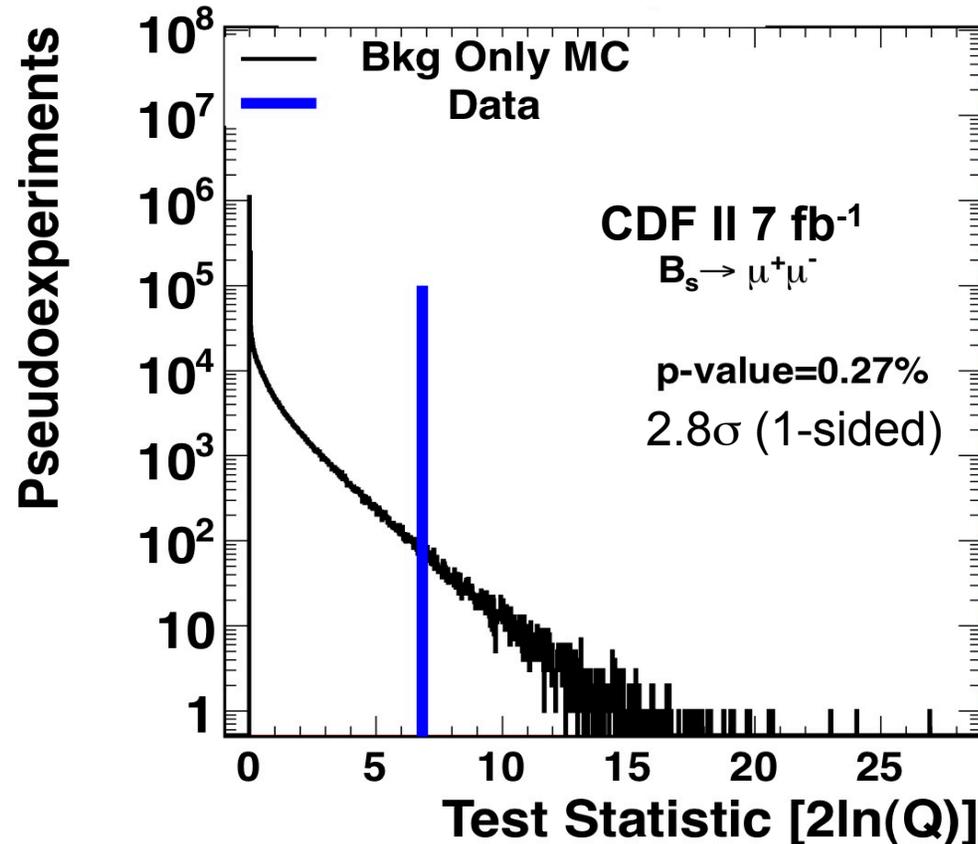
- Expected:

$$BF(B_d \rightarrow \mu^+ \mu^-) < 4.6 \times 10^{-9} \text{ (95\% CL)}$$
$$BF(B_d \rightarrow \mu^+ \mu^-) < 3.6 \times 10^{-9} \text{ (90\% CL)}$$

- Observed:

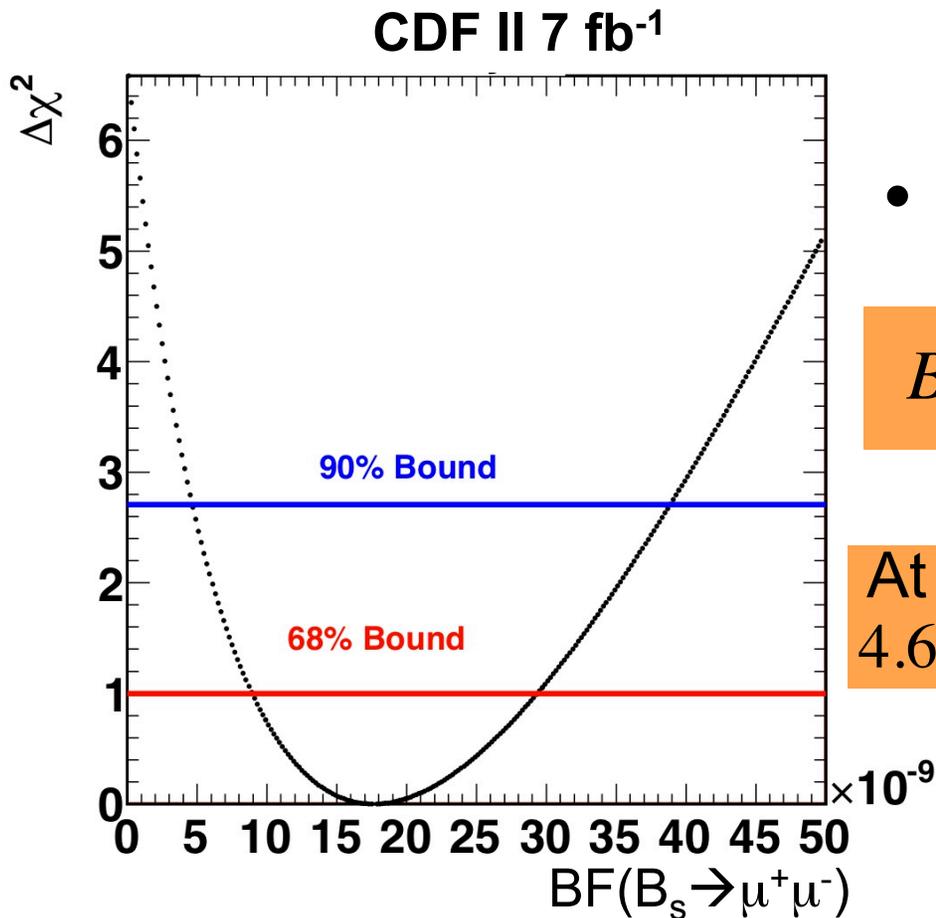
$$BF(B_d \rightarrow \mu^+ \mu^-) < 6.0 \times 10^{-9} \text{ (95\% CL)}$$
$$BF(B_d \rightarrow \mu^+ \mu^-) < 5.0 \times 10^{-9} \text{ (90\% CL)}$$

Result: B_s



- p-value using background-only pseudo-exp.
 - If we include SM signal, p-value $\rightarrow 1.9\%$ (2.1σ)

Result: B_s



- Fit to determine BF

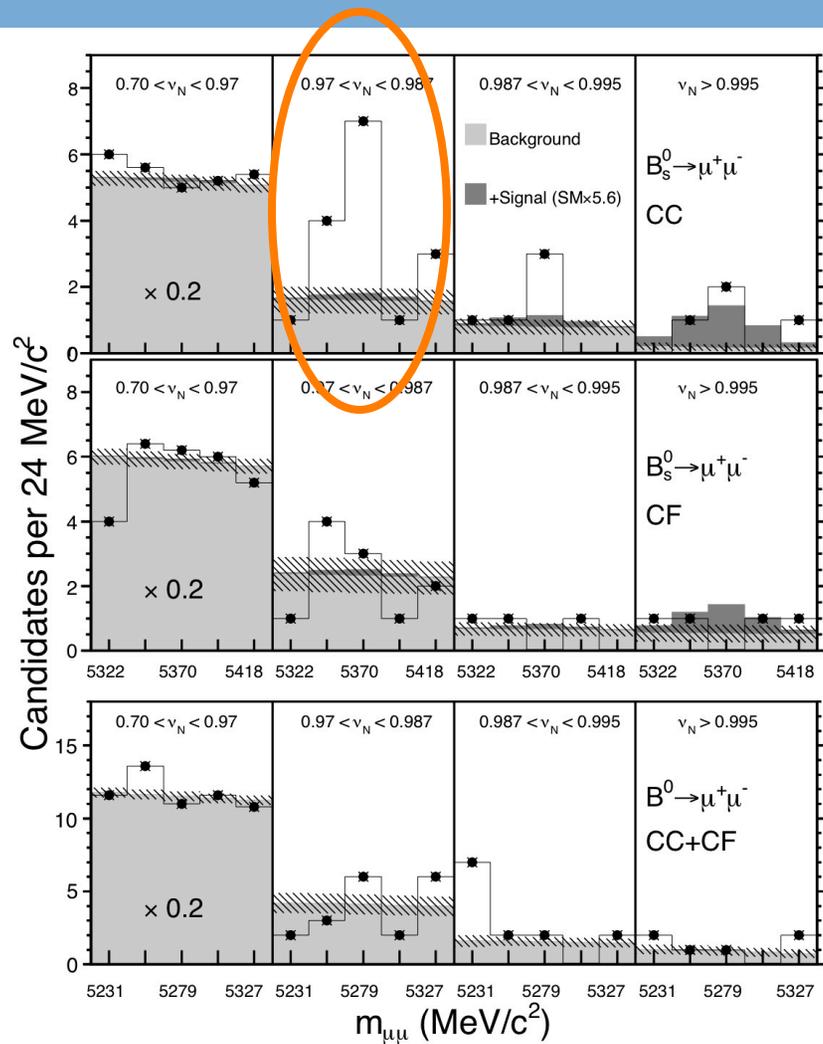
$$BF(B_s \rightarrow \mu^+ \mu^-) = (1.8^{+1.1}_{-0.9}) \times 10^{-8}$$

At 90% CL:

$$4.6 \times 10^{-9} < BF(B_s \rightarrow \mu^+ \mu^-) < 3.9 \times 10^{-8}$$

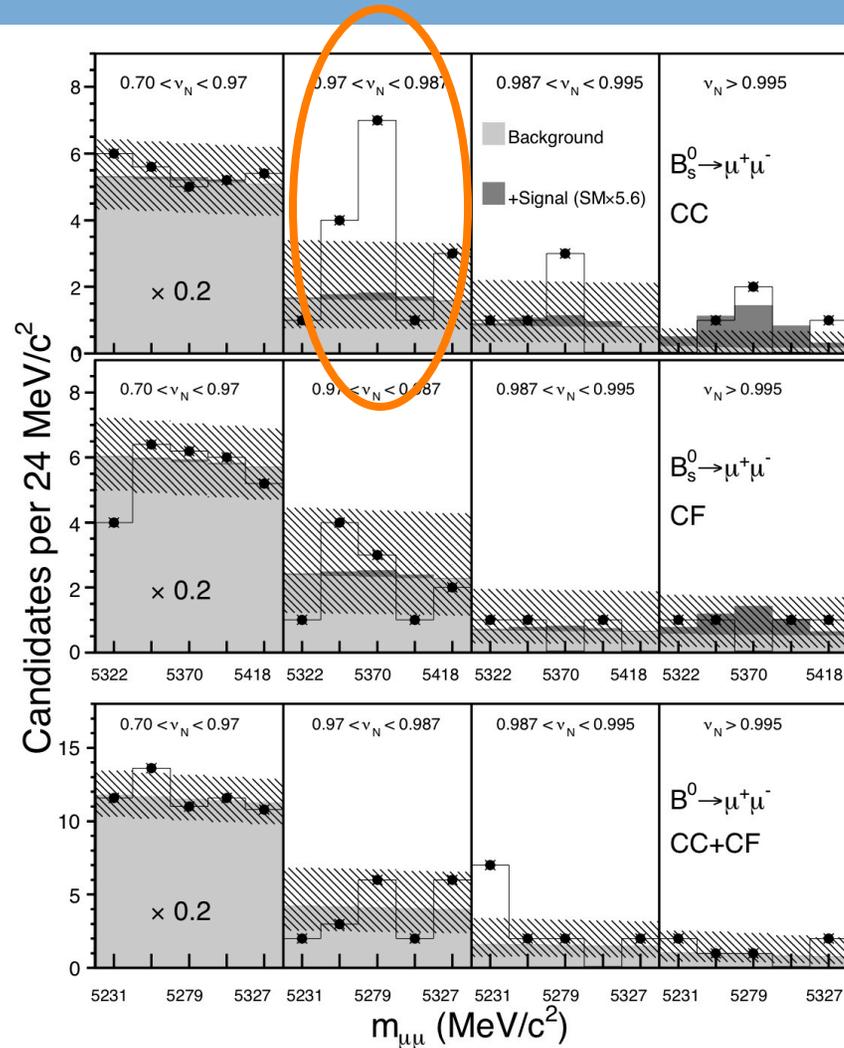
But...

Not so fast...



- What is this?

Not so fast...



Uncertainty:
syst \oplus Poiss.

- What is this?

Possibilities

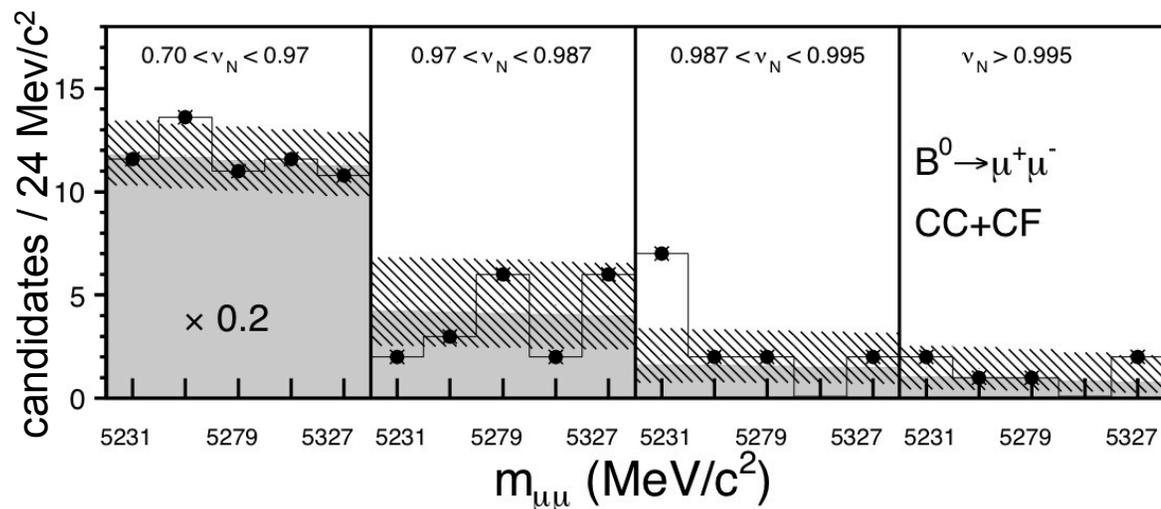
- Only two possible problems to consider (this is a simple analysis):
 - 1) Problem with background estimate
 - e.g. your π/K fake rates are wrong
 - 2) Problem with NN
 - e.g. NN is over trained or mis-modeled

Recall

- Background estimate in B_d region
 - Uses exact same sideband events
 - Uses exact same sideband fits for slope and normalization
 - Uses exact same $\pi/K \rightarrow \mu$ fake rates

Recall

- Background estimate in B_d region
 - Uses exact same sideband events
 - Uses exact same sideband fits for slope and normalization
 - Uses exact same $\pi/K \rightarrow \mu$ fake rates



- Accurately predicts data in signal region

Recall

- The yield of $B \rightarrow hh$ events in B_d region is about a factor of 10 larger than in the B_s region
 - If there were a problem with the π/K fake rates, it would show-up much more significantly in B_d
- In order to account for the observed excess, fake rates would have to be off by x10
 - They have a systematic uncertainty of 20%
 - Would generate much larger excesses in other bins

Possibilities

- Only two possible problems to consider (this is a simple analysis):
 - ✓ 1) Problem with background estimate
 - e.g. your π/K fake rates are wrong
 - 2) Problem with NN
 - e.g. NN is over trained or mis-modeled

Possibilities

- Only two possible problems to consider (this is a simple analysis):

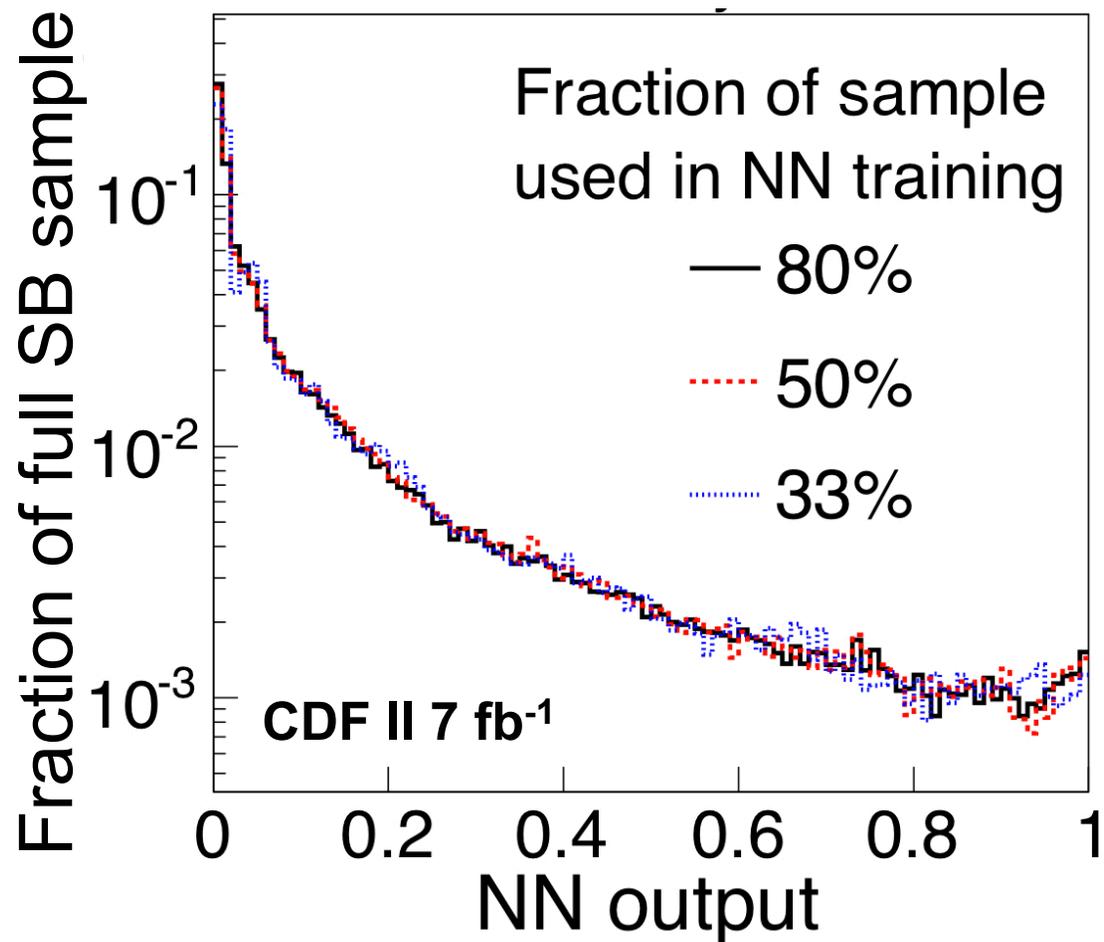
- ✓ 1) Problem with background estimate

- e.g. your π/K fake rates are wrong

- 2) Problem with NN

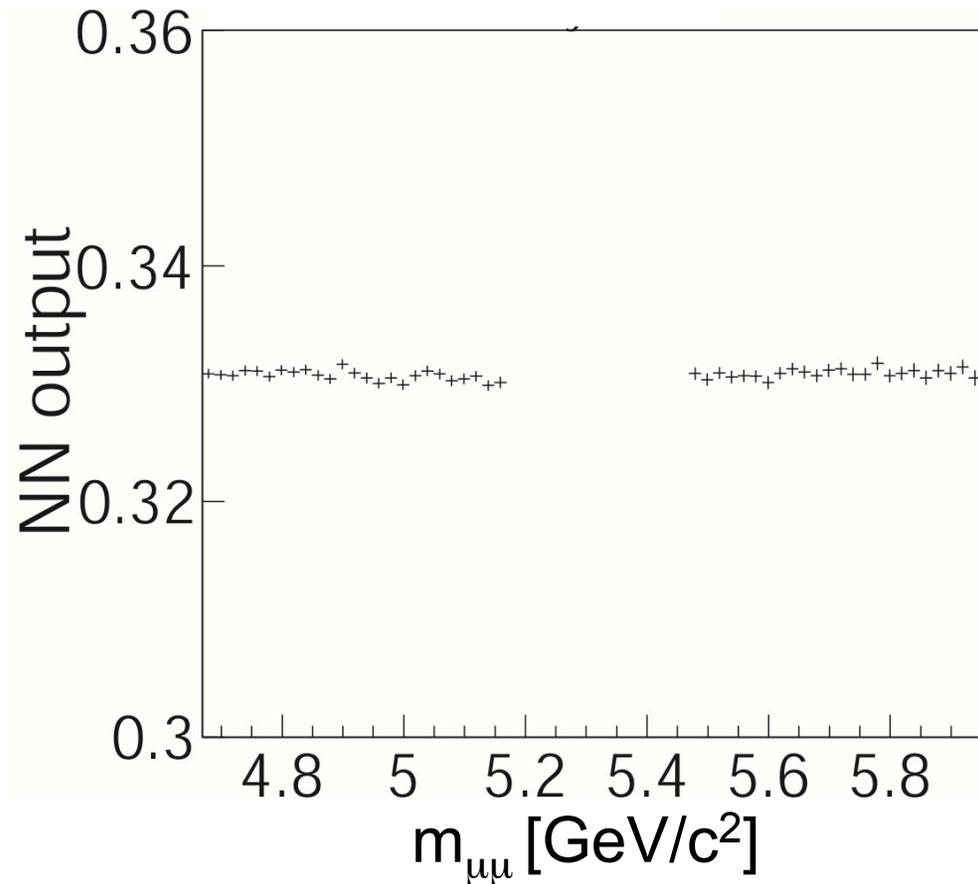
- NN over trained and biases comb. bgd. low
- NN has mass bias suppressing B_d events
- Shape of v_{NN} distribution poorly modeled

Problems with NN: Overtraining?



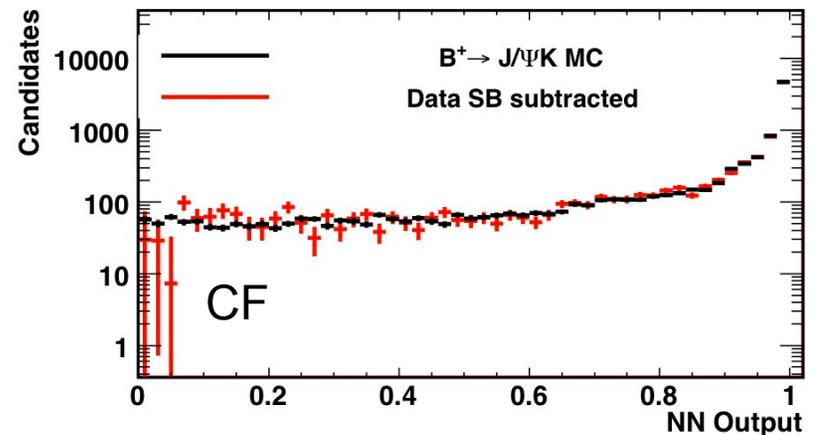
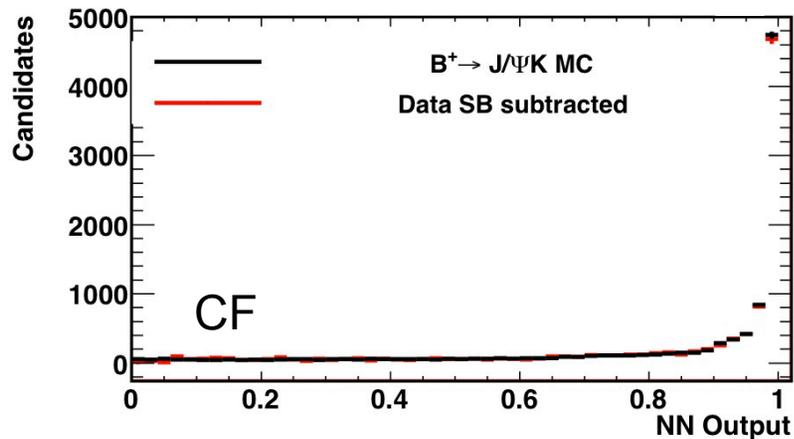
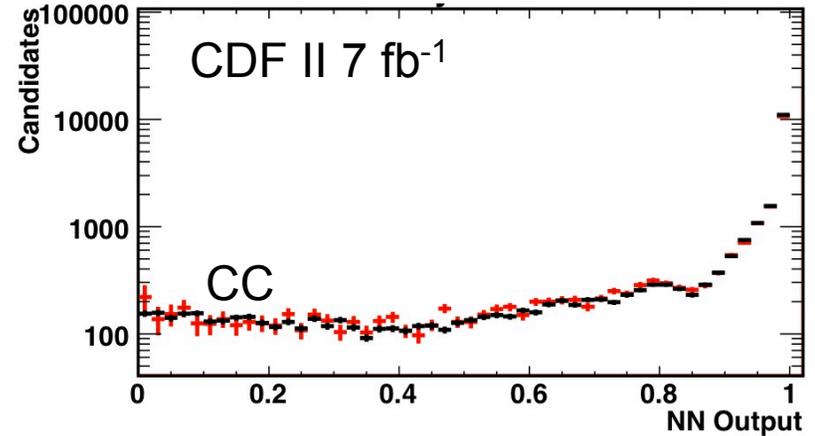
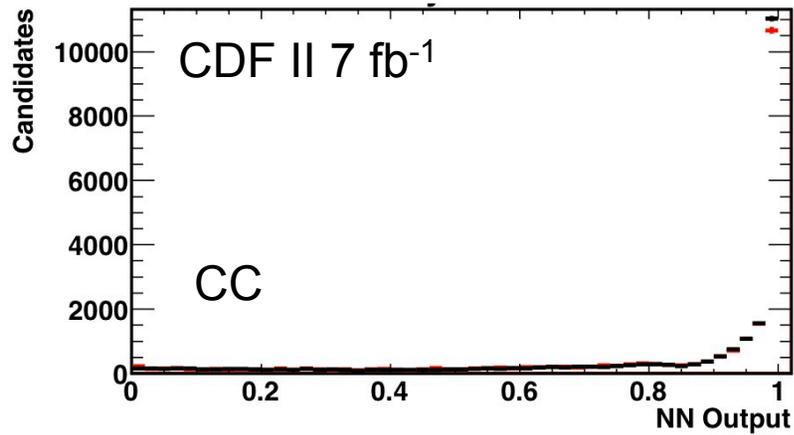
- No evidence of overtraining or bias

Recall



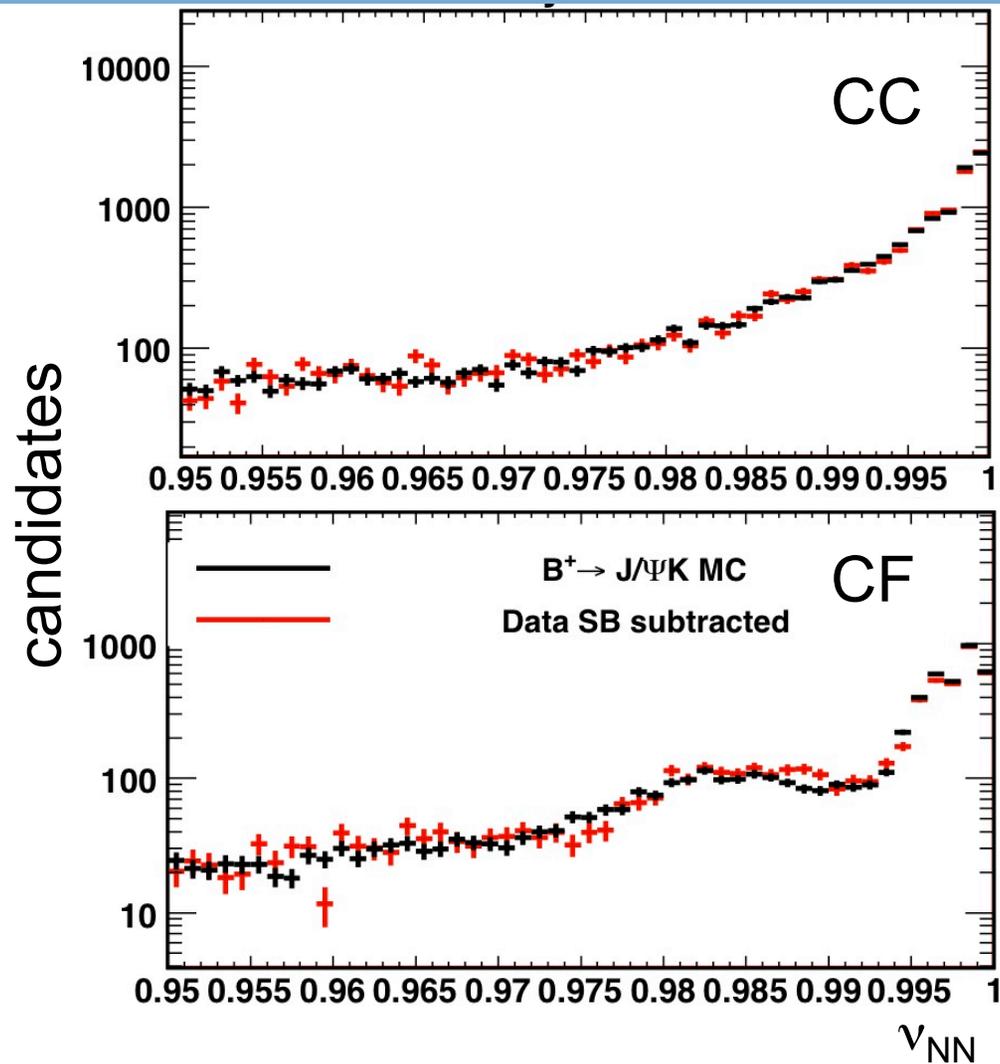
- No evidence that v_{NN} is correlated with $m_{\mu\mu}$
(cf. pages 36-38)

Recall



- No evidence of a significant MC mis-modeling of v_{NN} distribution for real B-decays

In addition



- Even in the steeply falling region above 0.99

Possibilities

- Only two possible problems to consider (this is a simple analysis):
 - ✓ 1) Problem with background estimate
 - e.g. your π/K fake rates are wrong
 - ✓ 2) Problem with NN
 - ✓ NN over trained and biases comb. bgd. Low
 - ✓ NN has mass bias suppressing B_d events
 - ✓ Shape of v_{NN} distribution poorly modeled

So?

Our conclusion

From the PRL:

“The source of the data excess in the $0.970 < \nu_{\text{NN}} < 0.987$ bin of the B_s signal region is investigated. ... Because the data in the B_d search region shows no excess, problems with the background estimates are ruled out. ... Problems with the NN are ruled out ... [since] studies find no evidence of a $\nu_{\text{NN}} - m_{\mu\mu}$ correlation, no evidence of overtraining, and no evidence of a significant mis-modeling of the ν_{NN} shape.... **In short, there is no evidence that the excess in this bin is caused by a mistake or systematic error in our background estimates or our modeling of the ν_{NN} performance and distribution. The most plausible remaining explanation is that this is a statistical fluctuation.**”

Our conclusion

“For our central result we use the full set of bins that had been established a priori since this represents an unbiased choice.”

$$p - \text{value}(\text{b - only}) = 0.27\%$$

$$p - \text{value}(\text{b + SM}) = 1.9\%$$

$$BF(B_s \rightarrow \mu^+ \mu^-) = \left(1.8_{-0.9}^{+1.1}\right) \times 10^{-8}$$

$$4.6 \times 10^{-9} < BF(B_s \rightarrow \mu^+ \mu^-) < 3.9 \times 10^{-8} @ 90\% CL$$

FYI

“...if we remove the $0.970 < v_{NN} < 0.987$ bin the results are not significantly affected.”

All bins ($0.70 < v_{NN}$)

$$p - \text{value}(\text{b - only}) = 0.27\%$$

$$p - \text{value}(\text{b + SM}) = 1.9\%$$

$$BF(B_s \rightarrow \mu^+ \mu^-) = (1.8_{-0.9}^{+1.1}) \times 10^{-8}$$

$$(4.6 - 39) \times 10^{-9} @ 90\%CL$$

2 Highest Bins ($0.987 < v_{NN}$)

$$p - \text{value}(\text{b - only}) = 0.66\%$$

$$p - \text{value}(\text{b + SM}) = 4.1\%$$

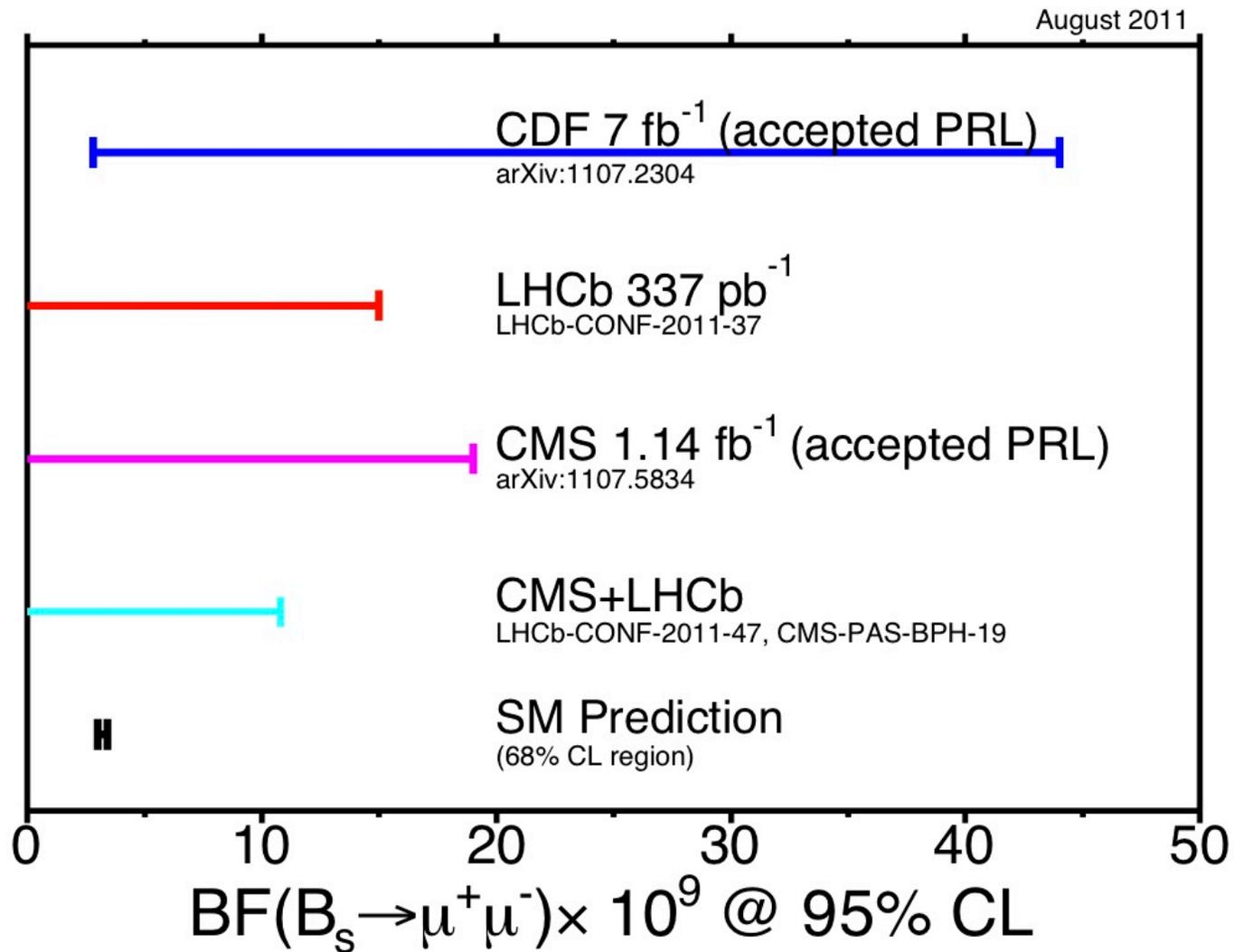
$$BF(B_s \rightarrow \mu^+ \mu^-) = (1.4_{-0.8}^{+1.0}) \times 10^{-8}$$

$$(3.3 - 33) \times 10^{-9} @ 90\%CL$$

Closing Remarks

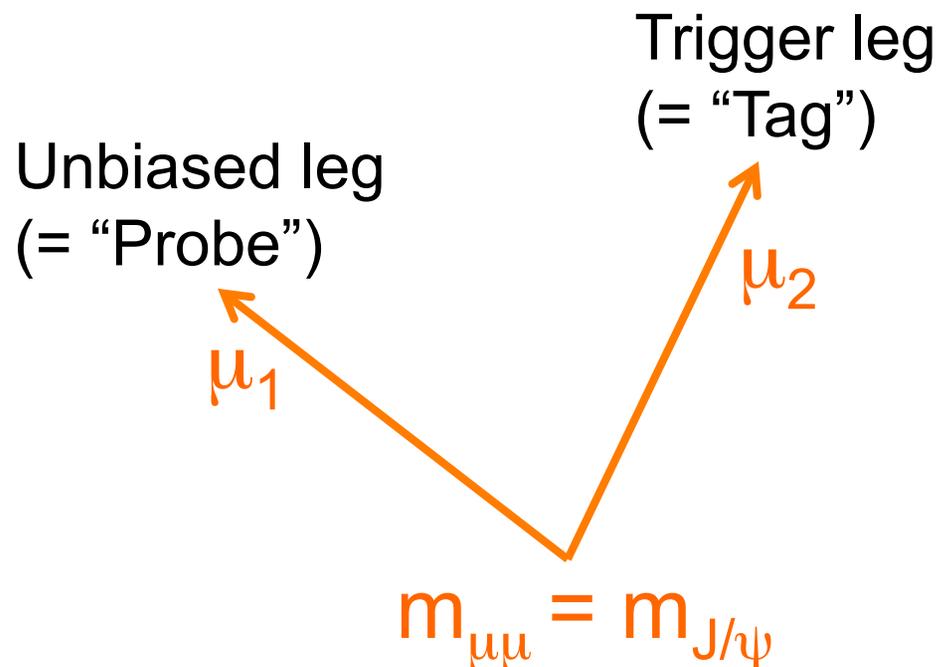
- CDF has an excess of $B_s \rightarrow \mu^+ \mu^-$ events at the level of $>2.7\sigma$ relative to bgd-only
- The fitted BF is compatible with the results from other experiments and the SM
- CDF will increase the data set by another 40% and publish a PRD

Closing Remarks



Backup Slides

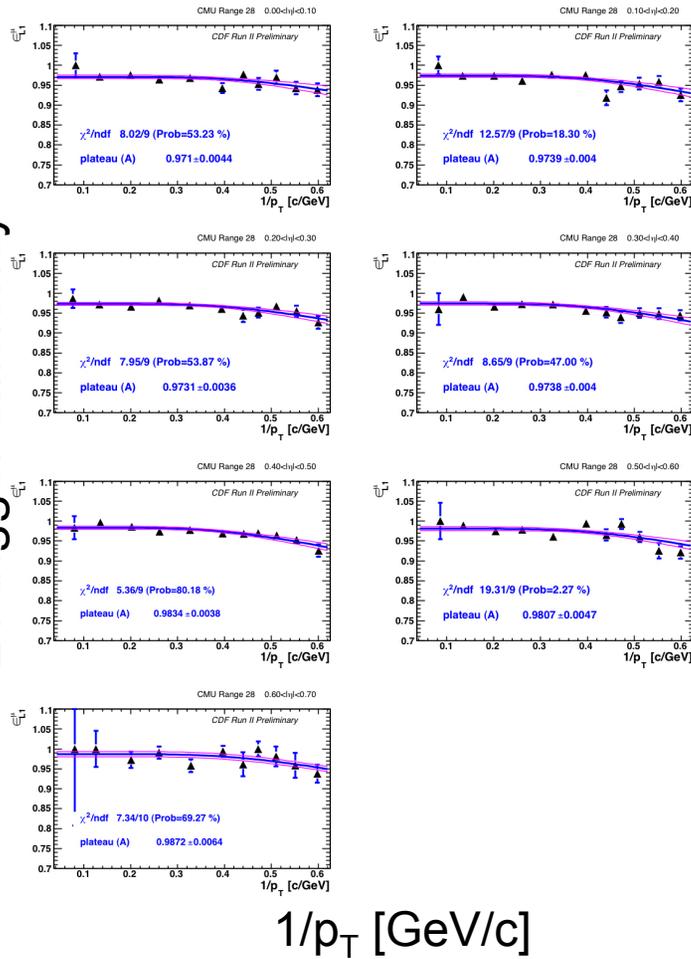
Trigger efficiency



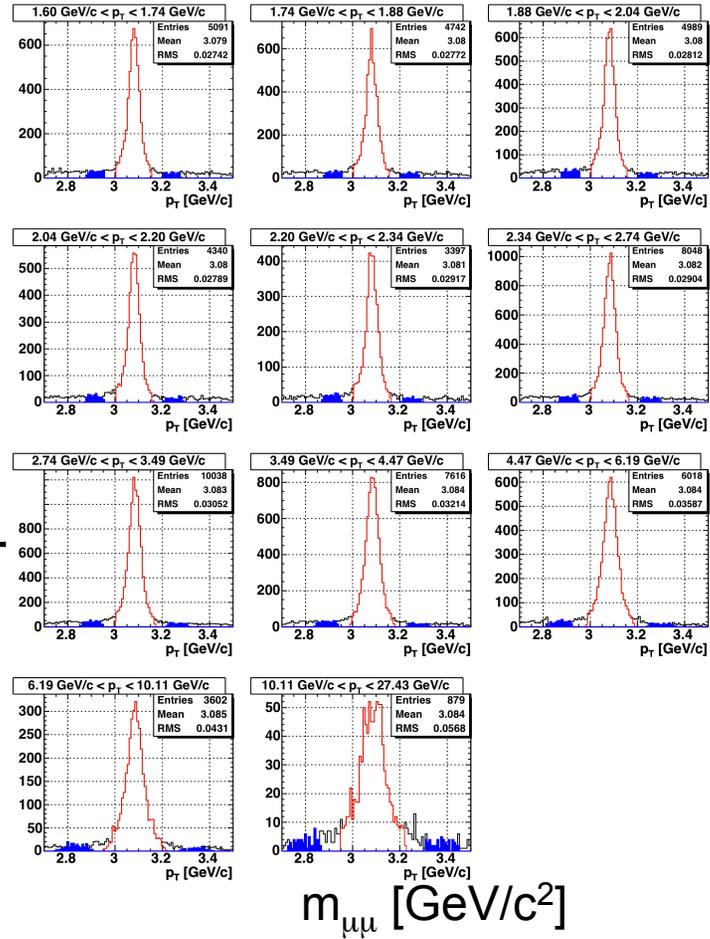
- “Tag-and-Probe” method using $J/\psi \rightarrow \mu^+\mu^-$ events collected with a single-leg μ trigger

Trigger Efficiency

L1 trigger efficiency



J/ψ candidates



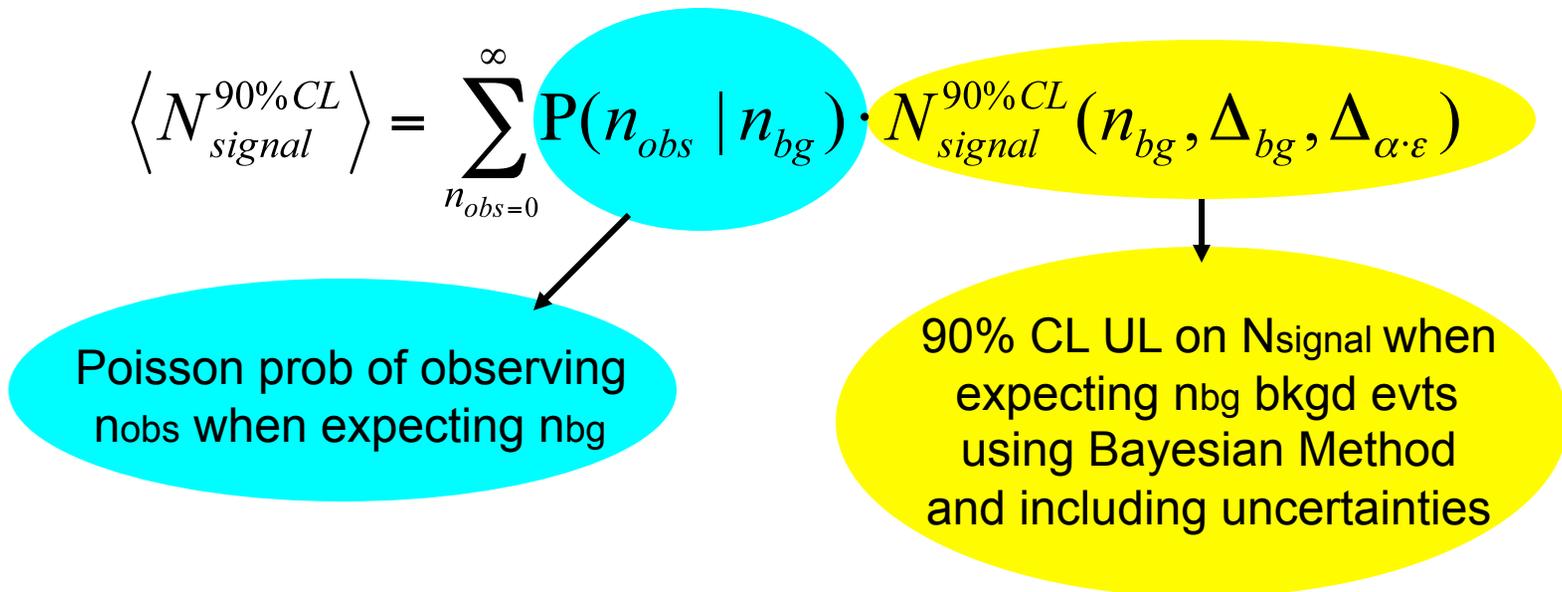
- $\epsilon_{\text{trig}} = \mathcal{F}(p_T, \phi, \eta, \text{run\#}, |z|_{\text{min}}^{\text{COT}})$

Expected Limit

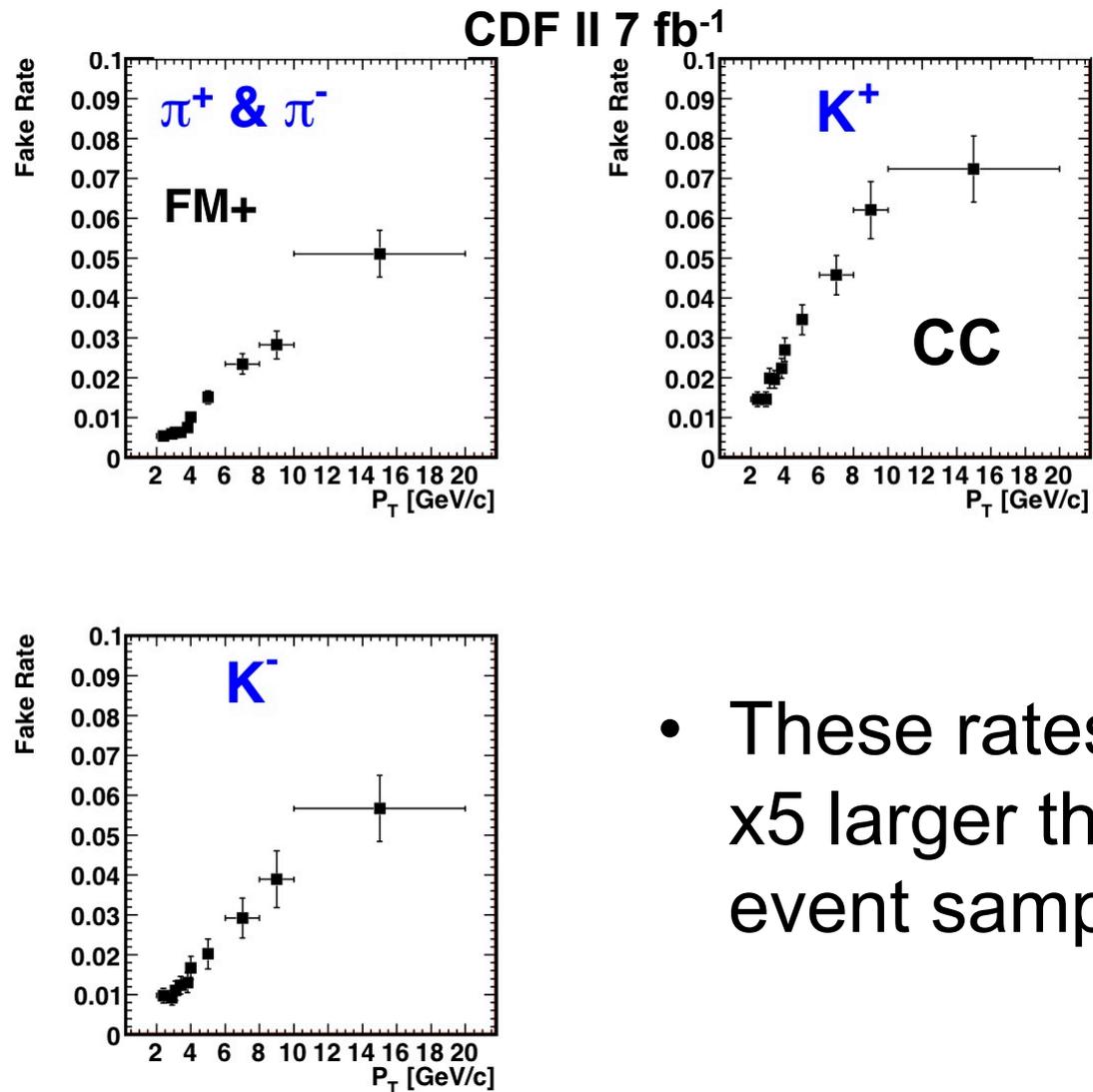
We used the set of requirements which yielded the minimum *a priori* expected BR Limit:

$$\langle BR(B_s \rightarrow \mu^+ \mu^-) \rangle = \frac{\langle N_{signal}^{90\%CL} \rangle}{\alpha \cdot \epsilon_{total} \cdot \sigma_{B_s} \int L dt}$$

where we've summed over all possible n_{obs} :



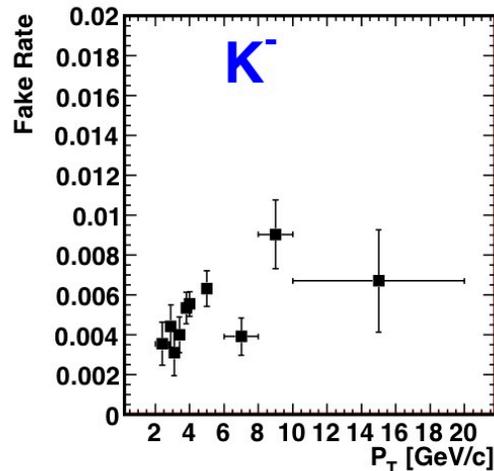
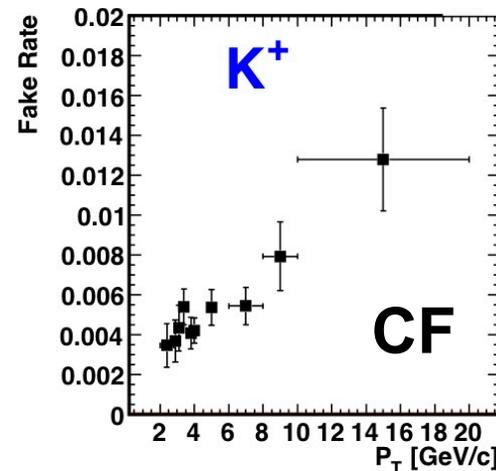
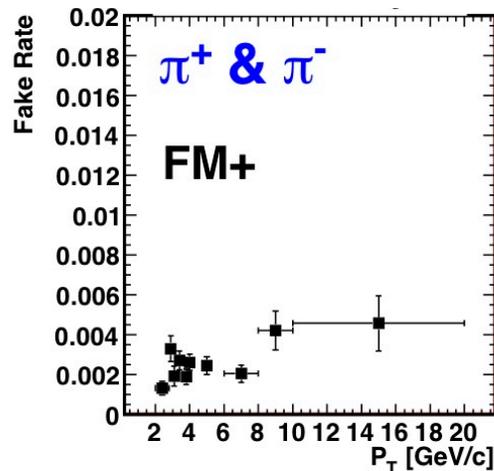
Hadron to muon Fake Rates: FM+ Sample



- These rates are about x5 larger than signal event sample

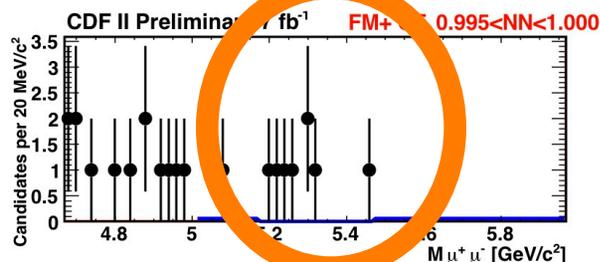
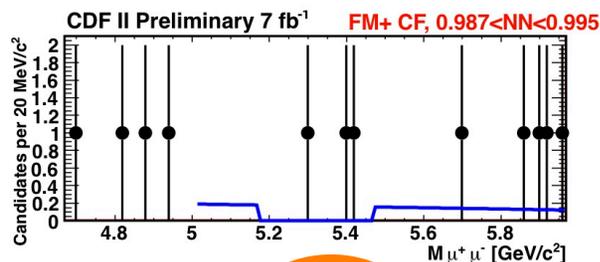
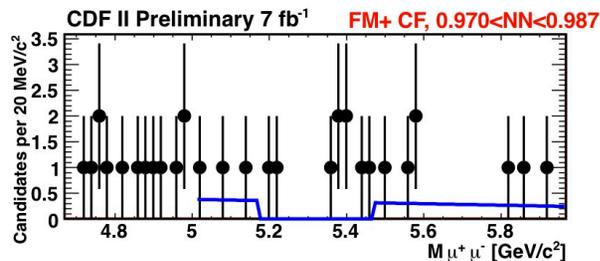
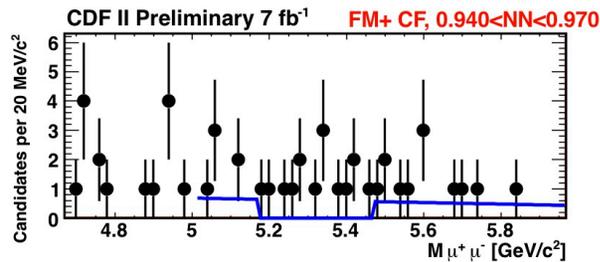
Hadron to muon Fake Rates: FM+ Sample

CDF II 7 fb⁻¹



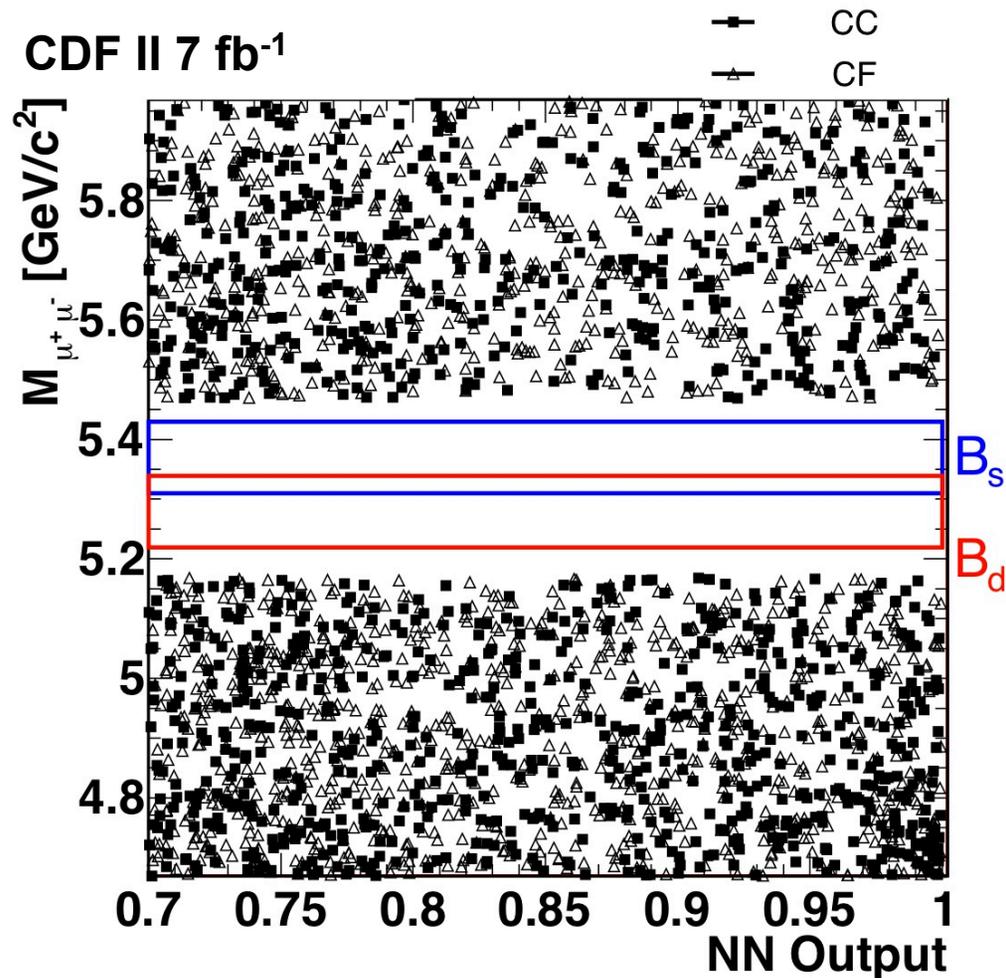
- These rates are about x2 larger than signal event sample

Cross-check Background Methodology



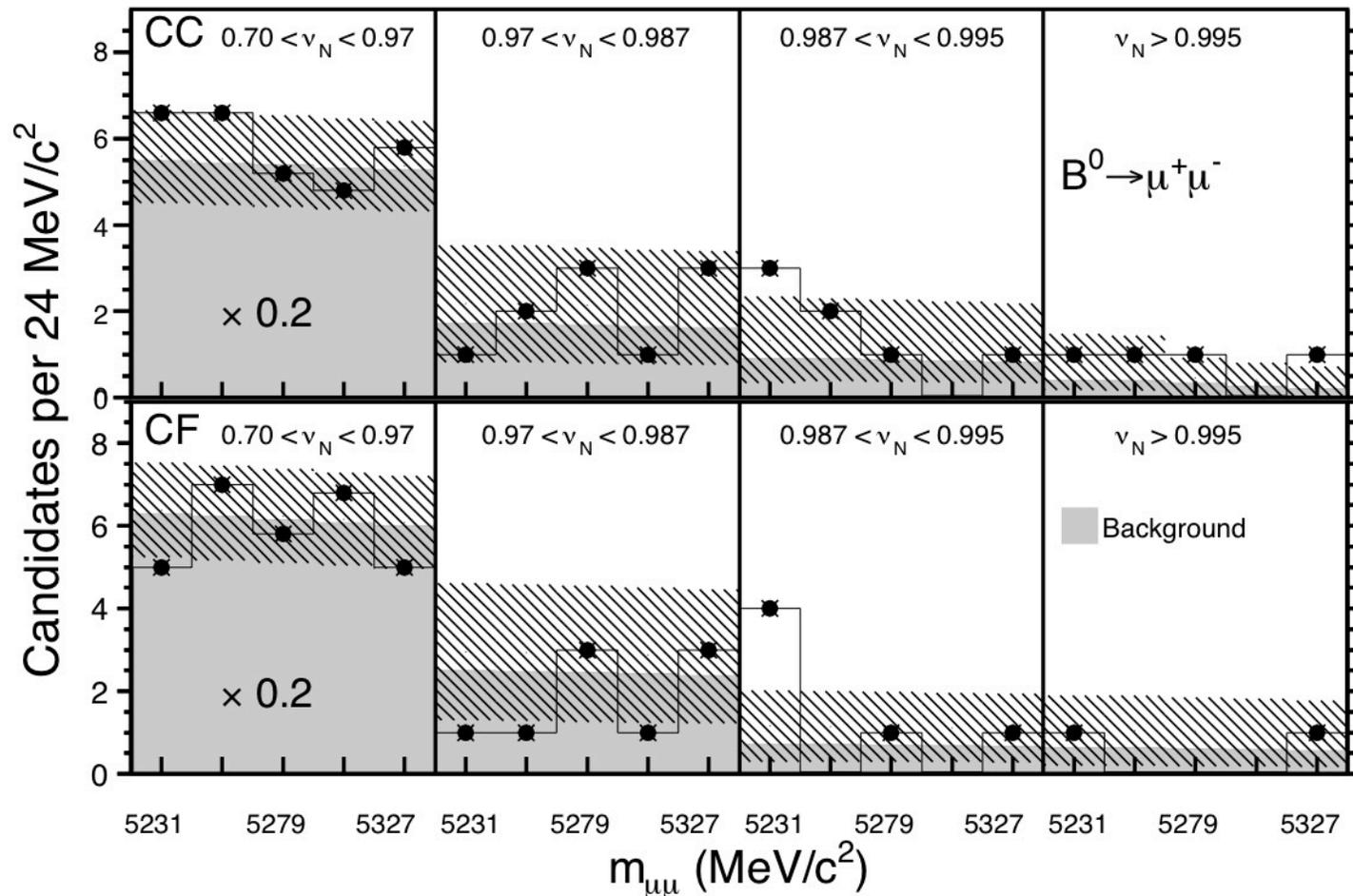
- Excess in this bin looks more consistent with combinatoric than $B \rightarrow hh$

Search sample: ν_{NN} vs $m_{\mu\mu}$ distribution



- Extended signal region blinded

Results in B_d Region



- B_d results for CC and CF separately

Results in B_d Region

Mass Bin (GeV/c^2)		5.219-5.243	5.243-5.267	5.267-5.291	5.291-5.315	5.315-5.339	Total
CC NN bin	Exp Bkg	8.32 ± 0.64	8.25 ± 0.63	8.17 ± 0.63	8.09 ± 0.62	8.01 ± 0.62	40.83
0.7-0.76	Obs	11	10	6	5	7	39
CC NN bin	Exp Bkg	8.75 ± 0.66	8.67 ± 0.65	8.58 ± 0.65	8.5 ± 0.64	8.41 ± 0.63	42.91
0.76-0.85	Obs	8	10	5	6	9	38
CC NN bin	Exp Bkg	3.69 ± 0.41	3.66 ± 0.4	3.62 ± 0.4	3.58 ± 0.4	3.54 ± 0.39	18.09
0.85-0.9	Obs	7	2	6	5	4	24
CC NN bin	Exp Bkg	3.66 ± 0.4	3.62 ± 0.4	3.58 ± 0.4	3.54 ± 0.39	3.5 ± 0.39	17.9
0.9-0.94	Obs	5	8	5	5	5	28
CC NN bin	Exp Bkg	3.0 ± 0.36	2.97 ± 0.36	2.93 ± 0.36	2.9 ± 0.35	2.86 ± 0.35	14.65
0.94-0.97	Obs	2	3	4	3	4	16
CC NN bin	Exp Bkg	1.71 ± 0.50	1.69 ± 0.50	1.67 ± 0.50	1.64 ± 0.49	1.62 ± 0.49	8.33
0.97-0.987	Obs	1	2	3	1	3	10
CC NN bin	Exp Bkg	0.90 ± 0.28	0.89 ± 0.28	0.86 ± 0.27	0.84 ± 0.27	0.81 ± 0.27	4.29
0.987-0.995	Obs	3	2	1	0	1	7
CC NN bin	Exp Bkg	0.40 ± 0.21	0.38 ± 0.20	0.32 ± 0.17	0.25 ± 0.15	0.20 ± 0.14	1.54
0.995-1	Obs	1	1	1	0	1	4
CF NN bin	Exp Bkg	8.89 ± 0.68	8.79 ± 0.67	8.68 ± 0.66	8.58 ± 0.65	8.47 ± 0.65	43.41
0.7-0.76	Obs	7	10	10	12	9	48
CF NN bin	Exp Bkg	9.9 ± 0.72	9.78 ± 0.71	9.66 ± 0.7	9.54 ± 0.69	9.42 ± 0.69	48.31
0.76-0.85	Obs	7	10	11	13	10	51
CF NN bin	Exp Bkg	5.15 ± 0.5	5.09 ± 0.49	5.02 ± 0.49	4.96 ± 0.48	4.9 ± 0.47	25.12
0.85-0.9	Obs	3	4	1	2	1	11
CF NN bin	Exp Bkg	4.06 ± 0.44	4.01 ± 0.43	3.96 ± 0.43	3.91 ± 0.42	3.86 ± 0.42	19.8
0.9-0.94	Obs	3	5	5	6	4	23
CF NN bin	Exp Bkg	3.45 ± 0.4	3.41 ± 0.4	3.37 ± 0.39	3.32 ± 0.39	3.28 ± 0.38	16.83
0.94-0.97	Obs	5	6	2	1	1	15
CF NN bin	Exp Bkg	2.50 ± 0.59	2.47 ± 0.58	2.44 ± 0.58	2.40 ± 0.57	2.37 ± 0.56	12.17
0.97-0.987	Obs	1	1	3	1	3	9
CF NN bin	Exp Bkg	0.71 ± 0.25	0.70 ± 0.25	0.69 ± 0.25	0.68 ± 0.24	0.67 ± 0.24	3.44
0.987-0.995	Obs	4	0	1	0	1	6
CF NN bin	Exp Bkg	0.62 ± 0.42	0.62 ± 0.42	0.60 ± 0.41	0.57 ± 0.40	0.55 ± 0.39	2.97
0.995-1	Obs	1	0	0	0	1	2

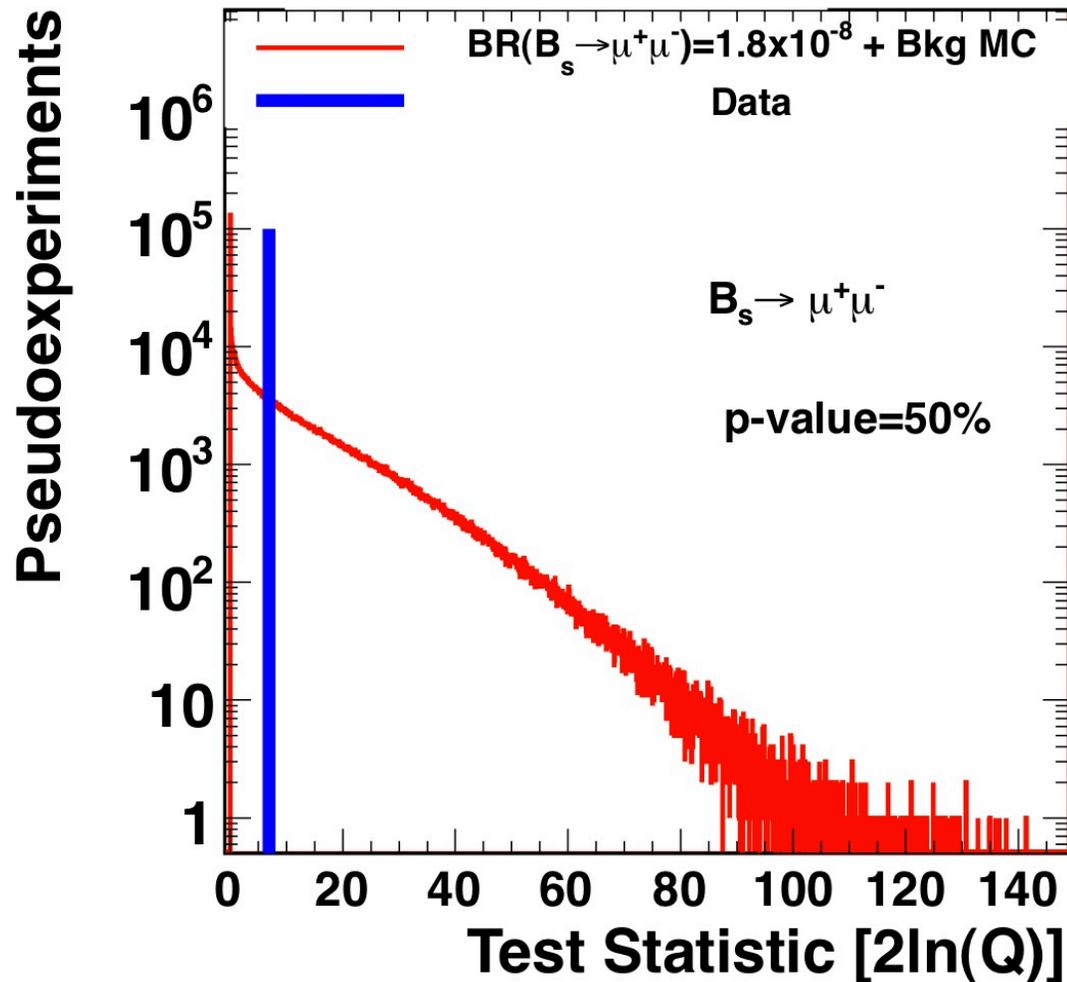
Table: B_d signal window for CC(top) and CF(bottom): Expected backgrounds, including $B \rightarrow hh$, and number of observed events.

Results in B_s Region

Mass Bin (GeV/c^2)		5.31-5.334	5.334-5.358	5.358-5.382	5.382-5.406	5.406-5.43	Total
CC NN bin	Exp Bkg	8.02 ± 0.62	7.94 ± 0.61	7.87 ± 0.61	7.79 ± 0.6	7.71 ± 0.59	39.34
0.7-0.76	Obs	9	6	6	2	5	28
CC NN bin	Exp Bkg	8.43 ± 0.64	8.34 ± 0.63	8.26 ± 0.62	8.18 ± 0.62	8.1 ± 0.61	41.32
0.76-0.85	Obs	8	6	11	11	7	43
CC NN bin	Exp Bkg	3.55 ± 0.39	3.51 ± 0.39	3.48 ± 0.39	3.44 ± 0.38	3.41 ± 0.38	17.4
0.85-0.9	Obs	5	6	2	5	4	22
CC NN bin	Exp Bkg	3.51 ± 0.39	3.47 ± 0.39	3.44 ± 0.38	3.41 ± 0.38	3.37 ± 0.38	17.2
0.9-0.94	Obs	4	5	4	5	7	25
CC NN bin	Exp Bkg	2.87 ± 0.35	2.84 ± 0.35	2.81 ± 0.34	2.78 ± 0.34	2.75 ± 0.34	14.04
0.94-0.97	Obs	4	5	2	3	4	18
CC NN bin	Exp Bkg	1.62 ± 0.49	1.60 ± 0.48	1.58 ± 0.47	1.57 ± 0.47	1.55 ± 0.46	7.92
0.97-0.987	Obs	1	4	7	1	3	16
CC NN bin	Exp Bkg	0.82 ± 0.27	0.80 ± 0.27	0.79 ± 0.26	0.78 ± 0.26	0.78 ± 0.26	3.97
0.987-0.995	Obs	1	1	3	0	0	5
CC NN bin	Exp Bkg	0.21 ± 0.14	0.18 ± 0.13	0.16 ± 0.12	0.16 ± 0.12	0.16 ± 0.12	0.87
0.995-1	Obs	0	1	2	0	1	4
CF NN bin	Exp Bkg	8.49 ± 0.65	8.39 ± 0.64	8.28 ± 0.63	8.17 ± 0.62	8.07 ± 0.61	41.4
0.7-0.76	Obs	8	13	9	9	9	48
CF NN bin	Exp Bkg	9.45 ± 0.69	9.33 ± 0.68	9.21 ± 0.67	9.1 ± 0.66	8.98 ± 0.65	46.07
0.76-0.85	Obs	7	8	7	11	4	37
CF NN bin	Exp Bkg	4.91 ± 0.48	4.85 ± 0.47	4.79 ± 0.46	4.73 ± 0.46	4.67 ± 0.45	23.95
0.85-0.9	Obs	1	5	6	3	5	20
CF NN bin	Exp Bkg	3.87 ± 0.42	3.82 ± 0.41	3.77 ± 0.41	3.73 ± 0.4	3.68 ± 0.4	18.88
0.9-0.94	Obs	4	1	6	3	3	17
CF NN bin	Exp Bkg	3.29 ± 0.38	3.25 ± 0.38	3.21 ± 0.37	3.17 ± 0.37	3.12 ± 0.36	16.04
0.94-0.97	Obs	0	5	3	4	5	17
CF NN bin	Exp Bkg	2.38 ± 0.56	2.34 ± 0.55	2.31 ± 0.54	2.28 ± 0.54	2.25 ± 0.53	11.57
0.97-0.987	Obs	1	4	3	1	2	11
CF NN bin	Exp Bkg	0.67 ± 0.24	0.66 ± 0.24	0.65 ± 0.24	0.64 ± 0.23	0.63 ± 0.22	3.25
0.987-0.995	Obs	1	1	0	1	0	3
CF NN bin	Exp Bkg	0.56 ± 0.39	0.54 ± 0.38	0.53 ± 0.38	0.52 ± 0.37	0.51 ± 0.36	2.66
0.995-1	Obs	1	1	0	1	1	4

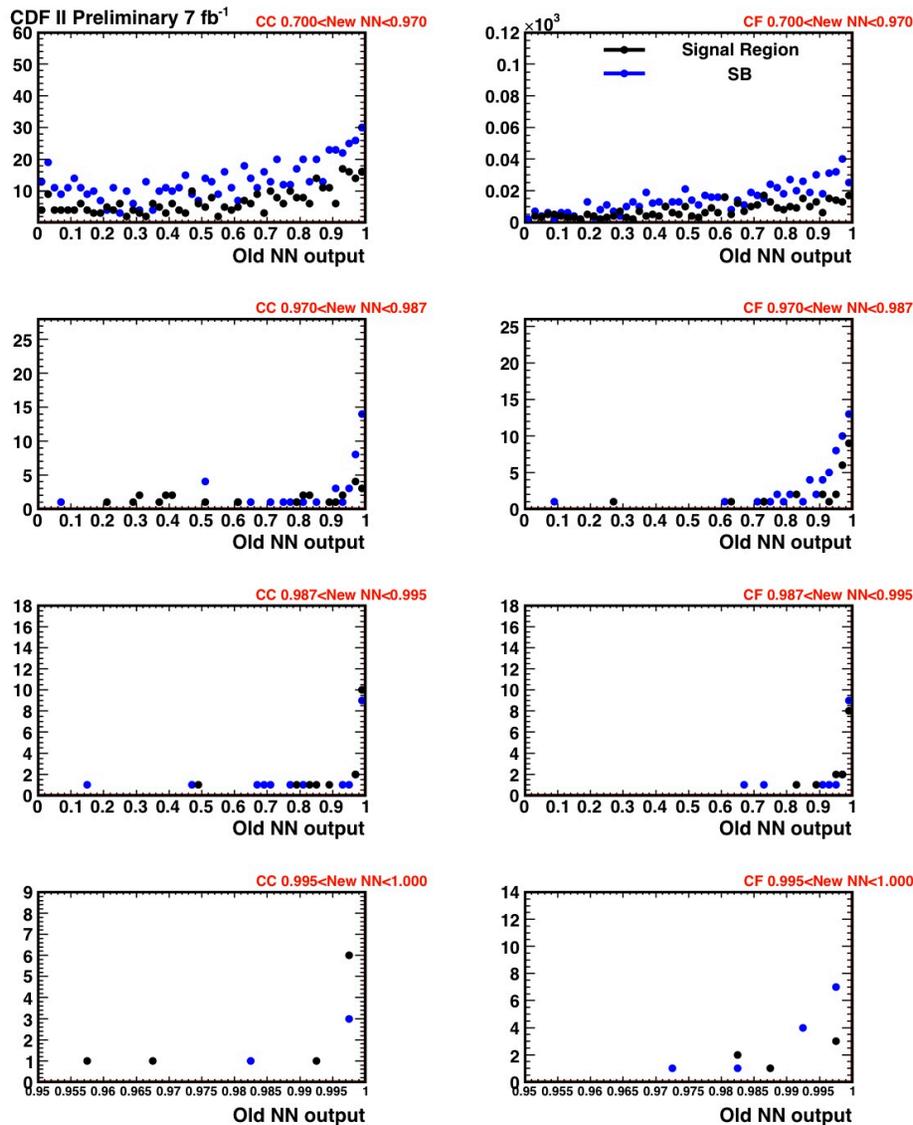
Table: B_s signal window for CC(top) and CF(bottom): Expected backgrounds, including $B \rightarrow hh$, and number of observed events.

p-value using best fit BF



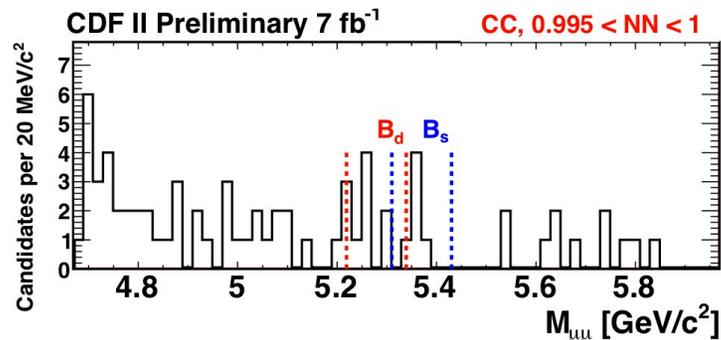
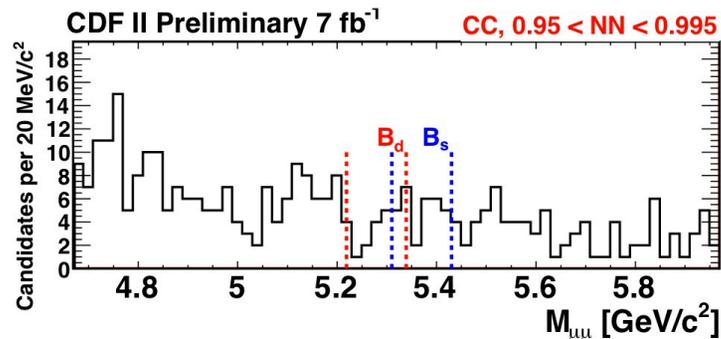
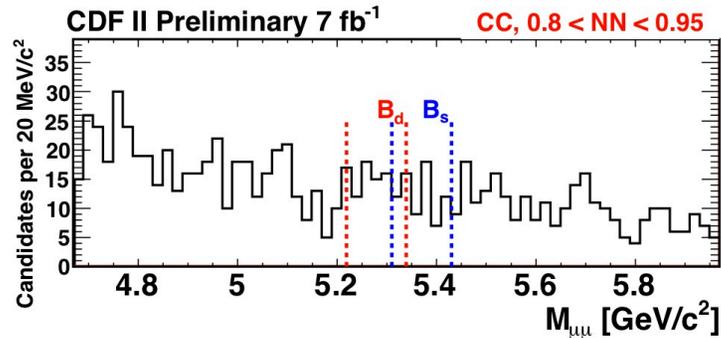
- pseudo-experiments used BF=best fit BF=5.6*SM

Comparisons with Old NN



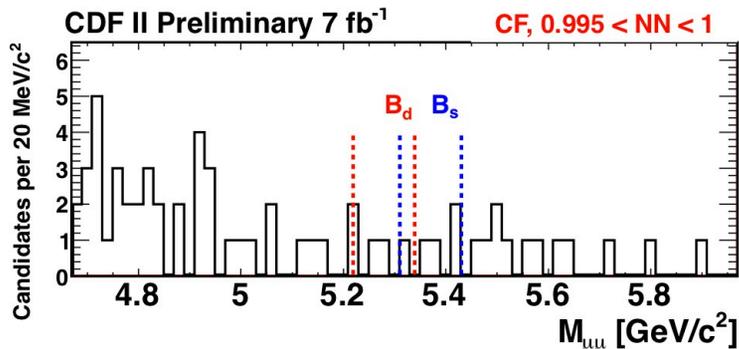
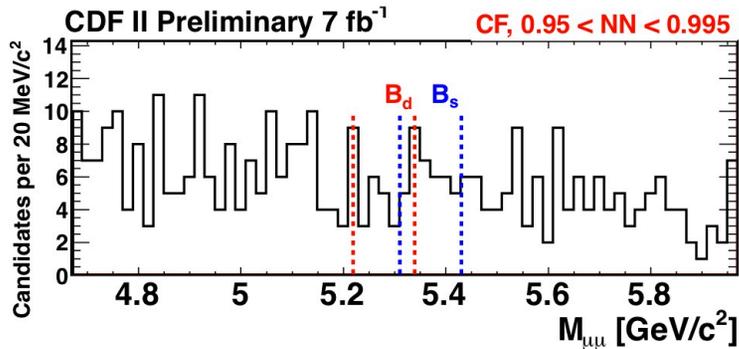
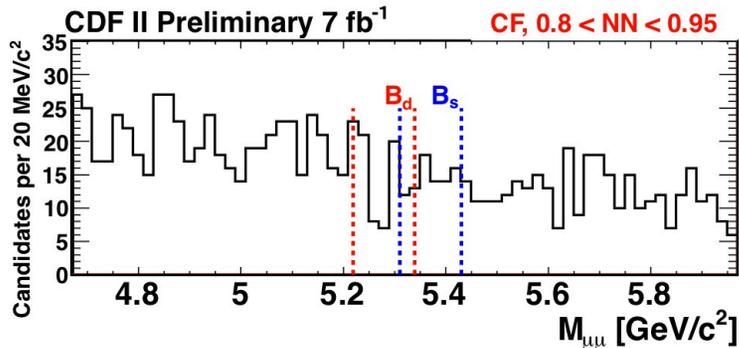
- The high score newNN events also high score in the oldNN

Comparisons with Old NN



- $m_{\mu\mu}$ distributions using **oldNN** and binning optimized for oldNN in 2 fb⁻¹ PRL

Comparisons with Old NN



- $m_{\mu\mu}$ distributions using **oldNN** and binning optimized for oldNN in 2 fb⁻¹ PRL