

Status update o $\Lambda_c \rightarrow p\mu^-\mu^+$

Marcin Chrząszcz¹, Tadeusz Lesiak², Mariusz Witek²,
Borys Nowak²

¹ University of Zurich,

² Institute of nuclear Physics



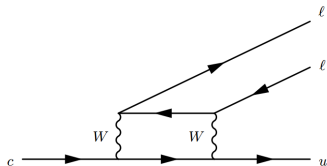
University of
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February 11, 2015

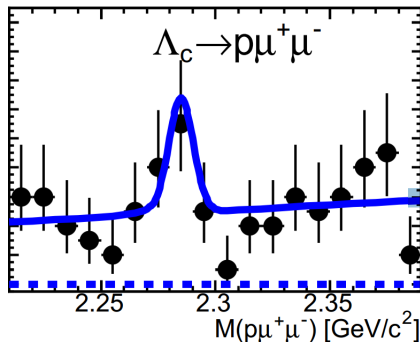
Why to search for $\Lambda_c \rightarrow p\mu^+\mu^-$?

- Decay of $\Lambda_c^+ \rightarrow p\mu^+\mu^-$ is a FCNC.
- Extremely suppressed in SM due to GIM mechanism.
- We will use the experience from $\tau \rightarrow p\mu\mu$.



$$\mathcal{B}(\Lambda_c^+ \rightarrow p\mu^-\mu^+) < 4.4 \times 10^{-5}$$

90% CL arXiv:1107.4465



Yield: $11.1 \pm 5.0 \pm 2.5$

We should easily beat Babar.

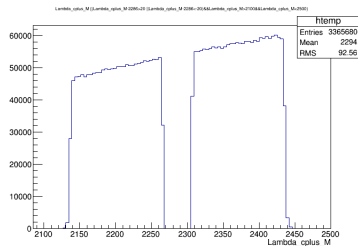
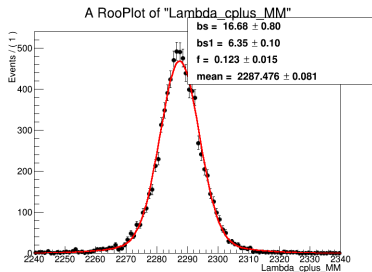
Preliminary selection

Stripping:

- $PID(\mu) > -5$, $PID(p) > 10$
- $IPChi2 > 9$, $PID(\mu - K) > 0$,
 $GHOST < 0.3$, $PID(p) > 10$,
 $P_t > 300$
- $\Delta m < 150 MeV$
- $c\tau > 100 \mu m$
- $IPChi2 < 225$

Additional:

- Blind region
 $|m(p\mu\mu) - 2286.46| < 40 MeV$.
- ϕ , ω veto.



Follow the strategy of τ analysis:

- Take prompt Λ_c , separate approach to SL.
- Loose cut preselection.
- Train MVA on MC prompt signal and recalibrate on data.
- Calibrate on data.
- Normalize to $\Lambda_c^+ \rightarrow pK^-\pi^+$, $\Lambda_c^+ \rightarrow p\pi^-\pi^+$ or $\Lambda_c \rightarrow p\phi(\mu\mu)$.
- Optimise the binning in MVA.
- CLs method for limit.



- We have 3 candidates for normalization channel.

① $\Lambda_c \rightarrow p\phi(\mu\mu)$, $BR = (2.4 \pm 0.8) \times 10^{-7}$

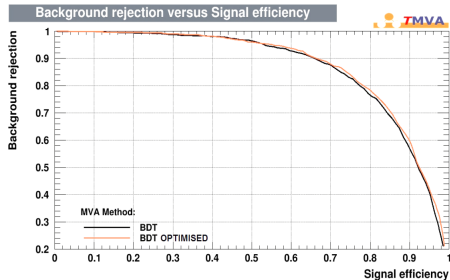
② $\Lambda_c^+ \rightarrow pK^-\pi^+$, $BR = (5.0 \pm 1.3) \times 10^{-2}$

③ $\Lambda_c^+ \rightarrow p\pi^-\pi^+$, $BR = (3.5 \pm 2.0) \times 10^{-3}$

From above list $\Lambda_c \rightarrow p\phi(\mu\mu)$ is a perfect candidate for normalization. However Br is a bit low.

Optimising the selection

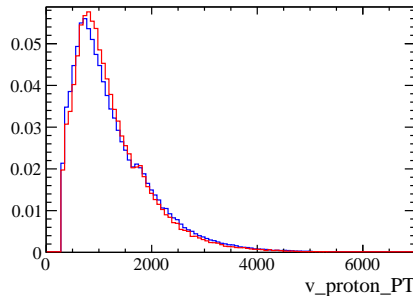
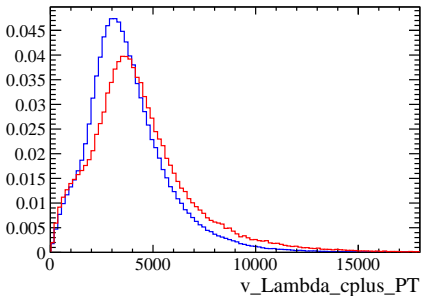
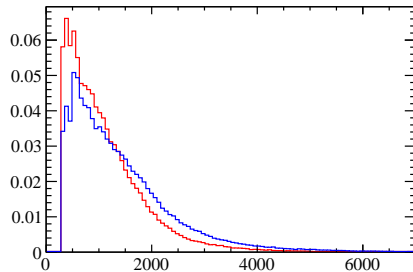
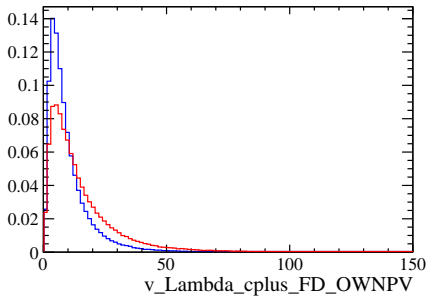
- Last time for studies we used BDT that was trained on the fly.
- Now a student produced a new optimised BDT
- I include his thesis as attachment to this presentation.



Comments about the BDT

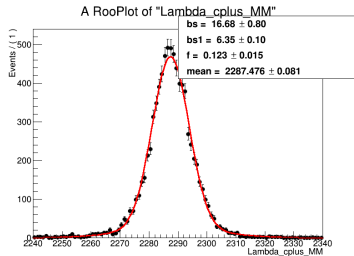
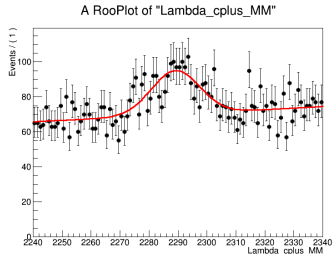
- From historical reasons we are training this classifier on MC vs MC
- Problematic part is that we have limited MC background sample.
- We have however the opposite sign (OS) channel:
 $\Lambda_c^+ \rightarrow \bar{p}\mu^+\mu^+$
- The obvious idea was to use this as an background extrapolation and use it for training and optimisation.

Differences between SS and opposite sign data



Normalization channel

- $\mathcal{O}(500)$ events in our dataset.
- Can be used for normalization!
- With the new BDT we also see small peak of ω
- Will veto that.



Possible background

Resonance	$\mathcal{B}(\Lambda_c \rightarrow pX)$	$\mathcal{B}(X \rightarrow \mu\mu)$
η	UNKNOWN	$(5.8 \pm 0.6) \times 10^{-6}$
ρ^0	UNKNOWN	$(4.55 \pm 0.28) \times 10^{-5}$
ω	UNKNOWN	$(9.1 \pm 3.0) \times 10^{-5}$
$f(980)$	$(2.8 \pm 1.9) \times 10^{-3}$	UNKNOWN
ϕ	$(8.2 \pm 2.7) \times 10^{-4}$	$(2.89 \pm 0.19) \times 10^{-4}$
Resonance	$\mathcal{B}(\Lambda_c \rightarrow pX)$	$\mathcal{B}(X \rightarrow \mu\mu\gamma)$
η	UNKNOWN	$(3.1 \pm 0.4) \times 10^{-4}$



- Looks like we will have limits $\mathcal{O}(10^{-8})$
- We already see a new $\Lambda_c \rightarrow \omega p$ decay, needs separate analysis
- Normalization channel is still open, but we are converging towards $\Lambda_c^+ \rightarrow p\pi^-\pi^+$
- We have one tight cut on the stripping (flight distance), we are considering several solutions.