A machine learning procedure for the selection of muon track candidates in the CBM experiment

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Outline

- Dilepton physics
- Muon setup of CBM
- Muon identification
- Cut optimization with machine learning procedure
- Reconstruction results
- Conclusions
- Next steps

Dimuon physics: highlights

Thermal radiation

Thermal Dileptons as Fireball Thermometer and Chronometer Ralf Rapp, Hendrik van Hees arXiv:1411.4612 Fireball model T. Galatyuk et al. Eur. Phys. J. A 52 (2016) 131





Charm production at threshold energies in cold and dense matter:

sub-threshold charm production via $N^* \rightarrow \Lambda_c + D$ $N^* \rightarrow N + J/\psi$

Sub-threshold charm production in nuclear collisions J. Steinheimer, A. Botvina, M. Bleicher, arXiv:1605.03439





Muon Chamber Systemglobal tracking, muon identification via hadron(MUCH)absorption in massive material

Transition Radiation global tracking **Detector (TRD)**

Time of Flight Detectorparticle identification via time measurement(TOF)

Configurations

low beam energies (up to 4 AGeV for Au beam)





μ reconstruction chain

STS track reconstruction ↓ extrapolation of the tracks through downstream detectors up to TOF



selection of the muon track candidates



calculation of the invariant mass spectra

Parameters for muon track candidate selection

- Vertex
 χ²/NDF
- STS
 - number of hits
 - χ²/NDF
- MUCH
 - number of hits
 - χ^2/NDF

- TRD
 - number of hits
 - χ²/NDF
- TOF*
 - mass cut



* rejection of punch through hadrons

Cut optimization

• <u>Manual:</u>

• χ^2 of the reconstructed tracks



mass cut in TOF



Artificial Neural Network

χ^2_{MUCH} vs. χ^2_{STS}





Test of ANN for $\omega \rightarrow \mu \mu$ reconstruction

Background: central UrQMD events @ 8 AGeV/c ω→μμ from PLUTO generator TMultiLayerPerceptron https://root.cern.ch/doc/master/classTMultiLayerPerceptron.html

ANN input: preselection

- χ^2/ndf
 - in primary vertex < 10</p>
 - in STS < 10
 - in MUCH < 10
 - in TRD < 10
- Number of hits
 - in STS > 5
 - in MUCH > 8
 - in TRD > 0
 - in TOF > 0



 ω + central Au+Au @ 8 A GeV/c

input parameters

 ω + central Au+Au @ 8 A GeV/c

purity

0.2

0.4

Ω

efficiency

0.6

0.8

ANN output

Positive tracks

Negative tracks



efficiency = signal passing ANN cut / total signal purity = signal passing ANN cut / (signal+background passing ANN cut)

Analysis results

ω + central Au+Au @ 8 A GeV/c

significance

ω efficiency, %

ω/background ratio



ω + central Au+Au @ 8 A GeV/c

Comparison with manual cuts

cuts	ε _ω %	ω background ratio	norm. significance
manual	1.21	0.016	1
ANN ±	1.21	0.040	1.58

manual cuts

 χ^{2}_{Vertex} <3, χ^{2}_{STS} <3, χ^{2}_{MUCH} <3 STS>6, MUCH>9, TRD σ_{TOF} =3



$\frac{\omega + central Au + Au @ 8 A GeV/c}{(for same efficiency)}$



14

ω + Au+Au @ 12 A GeV/c

Robustness of ANN

- ANN was trained using for UrQMD+PLUTO+GEANT3
- The output weights were used for DCMQGSM-SMM+PLUTO+GEANT4

ANN cuts	ε _ω %	ω background ratio	norm. significance
from training of DCMQGSM-SMM with G4	1.0	0.018	1
from training of UrQMD with G3	1.0	0.018	0.98

reconstructed background using the same set of manual cuts







ANN for physics performance



Simulation input for Au+Au @ 8 AGeV/c

Dimuon signals – PLUTO

- empirical angular-distribution parametrizations for selected processes
- resonance excitation in hadronic interactions
- nucleon-nucleon elastic scattering
- Multiplicities from Thermal-FIST and UrQMD



- Background UrQMD
 - low and intermediate energies (Vs<5 GeV): the interactions between known hadrons and their resonances
 - high energies (Vs>5 GeV): excitation of colour strings and their subsequent fragmentation into hadrons

Reconstruction steps

- CBM setup with thick absorber for measurements of intermediate and high dimuon mass regions
- Track, momentum and vertex reconstruction in STS
- Track extrapolation through MUCH and TRD up to TOF
- $\hfill Selection of the <math display="inline">\mu\hfill -track$ candidates using ANN training for MUCH with thick (1m) absorber



Low and intermediate mass region

central Au+Au @ 8 AGeV/c with dimuon signals



invariant mass

signal-to-background ratio



J/ψ results

central Au+Au @ 8 AGeV/c

invariant mass



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efficiency





PLUTO input



reconstructed J/ ψ



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Conclusions

- ANN was successful used for optimization of the selection cuts for muon analysis.
- ANN selection gives 3 times better ω-to-background ratio and 60% better significance for the same ω reconstruction efficiency as with manual set of cuts.
- The set of weights produced using ANN for simulation setup UrQMD+PLUTO+GEANT3 was tested for DCMQGSM-SMM+PLUTO+GEANT4. The results are comparable.

Next steps Input update PLUTO DCMQGSM-SMM GEANT4



Signal reconstruction via background subtraction

Multi-differential analysis





without selection





manual cuts

78

23456

0

1





10²

9 10

 χ^2_{STS}



1 2 3 4 5 6

ANN PID

background

Selection of μ^+_{ω}



manual cut from fit



cut from ANN



Track reconstruction

- STS track reconstruction (general for all physics cases). The π -mass hypothesis (m_{π} = 139.57018(35) MeV/c²) will be used for:
 - track finder procedure;
 - momentum reconstruction;
 - primary vertex reconstruction.
- Extrapolation of the STS reconstructed tracks through downstream detectors (MUCH, TRD, TOF). The μ -mass hypothesis (m_{μ} = 105.6583755(23) MeV/c²) will be used for calculation of energy loss of particles in thick materials (absorbers in MUCH), see Bethe-Bloch equation.
- The primary tracks passing through all detector subsystems will be selected as muon track candidates.

