

國立清華大學

物理系

碩士學位論文

基於 SPA-Net 的雙頂夸克

全強子衰變事件重建

Event reconstruction of
full hadronic Top-quark-pair decays
using SPA-Net

系所組別：物理所物理組

學號姓名：

108022517 何大維 (Ta-Wei Ho)

指導教授：

張敬民 教授 (Prof. Kingman Cheung)

徐士傑 教授 (Prof. Shih-Chieh Hsu)

中華民國一一〇年五月

Event reconstruction of full hadronic Top-quark-pair decays using SPA-Net

A Thesis Presented to
the Department of Physics at
National Tsing Hua University
in Partial Fulfillment for the Requirement of
the Master of Science Degree Program



By
Ta-Wei Ho
Advisor
Dr. Kingman Cheung
Dr. Shih-Chieh Hsu
May 2021

Abstract

The top quarks produced by pp collision in Large Hadron Collider(LHC), have a very complicated process still can't be well-classified today. In this project, we present a novel approach to the 'all hadronic decay' process of Top quarks based on the neural networks with attention mechanism, we call it 'Symmetry Preserving Attention Networks'(SPA-Net). This network identifies the decay products of each quark unambiguously and without combinatorial explosion. This approach performs an outstanding result compared to the existing state-of-the-art method. Our network can correctly assign all hadronic decay in 93.0% of 6 jets, 87.8% of 7 jets, and 82.6% of ≥ 8 jets event respectively.

摘要

在大型強子對撞機 (LHC) 實驗中，經由質子對撞所產生的頂夸克對具有非常複雜的過程以及產物，至今仍無法被非常正確的判別以及重建。在本研究中，我們提出了一個利用新穎的機器學習方法來對雙頂夸克全強子衰變過程進行重建。此方法基於 Attention mechanism，我們稱之為 Symmetry Preserving Attention Networks (SPA-Net)。這個模型架構可以在避免組合性爆炸的前提下對所有的衰變產物進行辨識以及重建。此方法對比於傳統的 χ^2 重建方式，表現出了非常巨大的差異。本方法可以在一、存在 6 jets 條件下正確的重建 93% 的事件；二、存在 7 jets 條件下正確的重建 87% 的事件；三、存在大於 8 jets 條件下正確的重建 82.6% 的事件。



Acknowledgements

I am so grateful for the helps from my colleagues, advisors, collaborators and friends. They gave me a lots of help when I was struggle in the research and technical issue. Their kindless helps me to come over the advantages and receive the degree.

When I doing this project, my advisor and my collaborator, Prof. Kingman Cheung and Prof Shih-Chieh Hsu gave me so much support that makes me can find a path to accomplish this project. Their suggestions and strong knowledge support are my best backup when I struggled in the mist. My colleague, Yi-Lun Chung, who helps me to resolve the problem of simulation package and theoretical problems, is also a person who deserved my most gratitude.

I appreciate all the collaborator who participate in this project, Mike Fenton, Alexander Shmakov, Daniel Whiteson, and Pierre Baldi. I would not finished this project without their help and support. The way that Alexander construct a machine learning model and the way that Mike describing a physics concept really teach me a lot. The guidance of Daniel, Shih-Chieh, and Pierre inspire me to explore more.

我想感謝我的同事、教授、朋友以及合作者。若沒有他們的鼎力相助，我肯定無法順利地完成這個研究。在我遇到各種問題，在困難中掙扎時，他們友善的協助讓我跨越難關，最後終於取得了學位。

張敬民教授以及徐士傑教授作為我的指導教授以及合作者，不論在物理上或是機器學習上，都非常大方地給予我非常多的建議。這些建議讓我能夠以足夠的知識來執行這個研究。他們的建議，每每都在我遇到困難時發揮關鍵的作用，也在我陷入迷霧時給予了我一盞明燈。

鍾沂倫，研究室的博士班學長，也在研究中給予了我相當大的幫助。無論是在各種環境的設定以及工作站的維護，乃至模擬軟體的設定以及物理模型的參數設定，都給了我非常多的幫助。他的協助讓我在維護工作站以及管理上輕鬆了非常多，真的是非常感謝他的幫忙。

Contents

Contents	ii
List of Tables	iii
List of Figures	iv
1 Introduction	1
2 The Top Physics and Machine Learning	3
2.1 The Top Physics	3
2.2 Machine Learning and its application on Particle Physics	5
3 Event Generation	6
3.1 MC samples	6
4 Data analysis and Event reconstruction	8
4.1 Data analysis	8
4.1.1 Event selection	8
4.1.2 Truth matching	10

4.1.3	Custom barcode system	10
4.2	Event reconstruction	11
4.2.1	χ^2 reconstruction	11
4.2.2	Machine Learning Approach	11
5	Discussion	12
5.1	Invariant mass	12
5.2	Reconstruct efficiency	12
6	Conclusion	13
	Appendix	13
	Reference	14



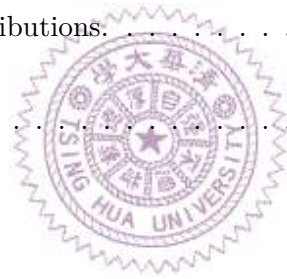
List of Tables

2.1	Top quark pair decay process[3]	4
4.1	Rule of cuts. All the cuts require a kinematic limitation that $p_T > 25$ GeV and $ \eta < 2.5$	9



List of Figures

2.1	The schematic of Top quark decay channels.[7]	4
2.2	How self-attention works.[10]	5
4.1	Cutflow of all hadronic top decay.	9
4.2	Demonstration of distributions.	9
4.3	Design of barcode.	10



Chapter 1

Introduction

At Large Hadron Collider(LHC), two protons collide with very high energy and produce many kinds of products. A process that pp collision produces a pair of Top quark and result in the 6 jets final state is called **Full Hadronic Top-quark-pair decay**. This process has a very complicated signature due to a large number of combinations. These jets produced by the top quark pair is hard to tag as a specific parton correctly. A traditional method is to reconstruct the event using χ^2 reconstruction, but it takes such a long time to compute and cannot provide enough accuracy to reconstruct an event. The importance of studying Top quark and its full hadronic decay channel is 1. Top quark is the most heaviest fundamental particle in standard model and will decay before hadronization, 2. The brach ratio of full hardonic decay is the biggest part in Top quark decay(46%).

For a problem that contains a large amount of data and highly requires computing resources, machine learning can widely provide powerful support on solving the problem and helps to reduce the time-wasting. The machine learning method helps to discover physics phenomena with very outstanding effort. A remarkable discovery that helps by machine learning is the discovery of Higgs Boson. Both CMS and ATLAS groups apply the machine learning method to promote the searching of Higgs Boson. [1][2]

In this thesis, we perform a novel architechture for parton-jet assignment problem. This method is base on the state-of-the-art machine learning technology, Attention mechanism. We call this novel ML model **Symmetry Preserving Attention NETworks (SPA-NET)**. By applying attention networks, the SPA-NET perform a outstanding

performance compare to traditional method while avoiding combinatorial explosion. And thanks to the natural properties of attention network, the network reflect the permutation symmetry naturally and provide a chance to explore in set-based output. We will discuss the Top physics and the concept of machine learning in chapter 2; and explain our event generation and simulation configuration in chapter 3; then introduce how we analyze the dataset and reconstruct the event using traditional method and ML approach in chapter 4. We will discuss our work in chapter 5 and summerize in chapter 6.



Chapter 2

The Top Physics and Machine Learning

2.1 The Top Physics

Top quark, the most massive fundamental particle in Standard Model(SM), is the only one quark that decay semi-weakly. Its large mass leads to short lifetime and decay before hadronization occur. Top quark contains so many propeties that interest us, such like its mass, couplings, and cross section, e.t.c. By measure these properties accurately can bring us a worth understanding of fundamental interactions and the key to Beyond Standard Model.[3]

In recent model, Top quark pair produced by pp collision has three decay modes, **all-hadronic channel**, **semi-leptonic channel**, and **dileptonic channel**. The branch ratio of each channel, has shown in the Table 2.1. The decay width of Top quark predict in SM is[4]:

$$\Gamma_t = \frac{G_F m_t^3}{8\pi\sqrt{2}} \left(1 - \frac{M_W^2}{m_t^2}\right)^2 \left(1 + 2\frac{M_W^2}{m_t^2}\right) \times \left[1 - \frac{2\alpha}{3\pi} s \left(\frac{2\pi^2}{3} - \frac{5}{2}\right)\right] \quad (2.1)$$

Table 2.1: Top quark pair decay process[3]

Decay Channel	Process	Branch Ratio(%)
All-hadronic	$t\bar{t} \rightarrow W^+bW^-\bar{b} \rightarrow q\bar{q}'bq''\bar{q}'''\bar{b}$	45.7
Semi-leptonic	$t\bar{t} \rightarrow W^+bW^-\bar{b} \rightarrow q\bar{q}'b\ell^-\bar{\nu}_\ell\bar{b} + \ell^+\nu_\ell bq''\bar{q}'''\bar{b}$	43.8
Dileptonic	$t\bar{t} \rightarrow W^+bW^-\bar{b} \rightarrow \ell^+\nu_\ell b\ell'\bar{\nu}_{\ell'}\bar{b}$	10.5

In recent study, the most precise result of Top quark mass is measured in the lepton+jets channel due to its good signal-to-background ratio and the presence of one neutrino final state. Although all-hadronic channel has the most probability to appears in the Top quark pair decay process, but it couldn't provide a precise mass measurement due to its poor signal-to-background ratio. The poor signal-to-background ratio of all-hadronic channel is due to the difficult QCD background. The CMS and ATLAS group approach a precision of Top mass measurement using all-hadronic channel with 0.65% and 1.1%. [5][6]

The channel we are interested in this project is the **jet-parton assignment problem in all hadronic decay channel**. The reason that we are interested in this channel is the resolved 6 jets signature and the potential of machine learning method apply on the ambiguous event reconstruction problem. There exist 6 jets in the final state, 2 b-jets and 4 quark jets, they can be separated into two groups (b, q, q') and (\bar{b}, q'', q''') . A schematic of the decay products is shown in 2.1.

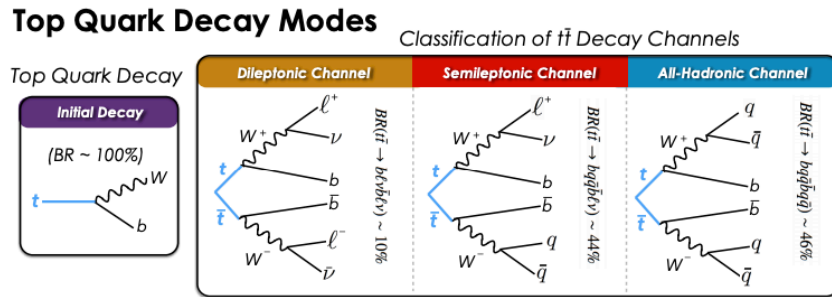


Figure 2.1: The schematic of Top quark decay channels.[7]

2.2 Machine Learning and its application on Particle Physics

Machine Learning has been applied to most of the region in recent age, so dose particle physics. From the search of higgs boson(neural network and BDT) to the b-tagging technology(BDT[8]), physicist already applied several kinds of machine learning method to recent researchs.

In a nut shell, machine learning can break into several cases, it can help to do classification, regression, and the clustering problem. It can not only helps to accelerate the computation of well-defined problem, and also find a new path to unsolved area. We wil use a state-of-the-art machine learning technology, attention mechanism. Attention mechanism is a technology base on the evolution of RNN.[9] The attention mechanism will not only consider local relationship and the sequence neighbor, but also calculate the global relation base on the self-attention calculation shown in Figure 2.2. Using this novel architecture, we will train on the relationship between each jet and try to figure out the correct pair information.

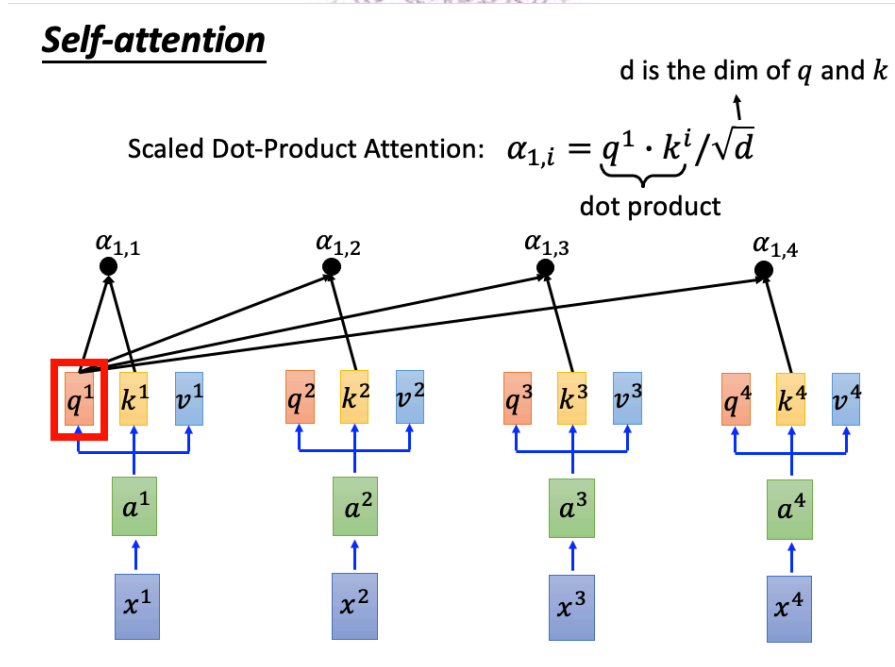


Figure 2.2: How self-attention works.[10]

Chapter 3

Event Generation

3.1 MC samples

For data preparation, we generate our dataset using a custom docker image with MadGraph_aMC@NLO(v2.7.2), Pythia8(v.8.2), and Delphes(v3.4.2) for showering, hadronization, and detector simulation. We apply the ATLAS parametrization during detector simulation. The data are generated at Leading order, including quantum chromodynamics(QCD). The top mass is configured as $m_{top} = 173 \text{ GeV}$. We force the W quark decay hadronically into a (q, q') pair. Following is our configuration:

```
generate p p > t t~ QED=0, (t > W+ b, W+ > j j), (t~ > w- b~, w- > j j)
output <file_path>
launch <file_path>
shower=Pythia8
detector=Delphes
analysis=OFF
done
set nevents = 10000
set iseed = 1
Delphes/cards/delphes_card_ATLAS.tcl
done
exit
```

Listing 3.1: Configuration for generating samples

To get a more general performance, we scan the iseed value from 1 to 30000, each value has around 100 file with 10 thousand events before event selection. The reason for scanning iseed value is that the iseed value is the key to the random generation. Originally, the program will choose the iseed value randomly and generate different samples. By scanning the issued value, we can check whether the network can works well on differnet iseed number or not.



Chapter 4

Data analysis and Event reconstruction

4.1 Data analysis



4.1.1 Event selection

The top all hadronic decay channel has 2 b-jets and 4 quark jets, all of them in our configuration is not in boosted region. Follow the event selection used in the reference[7], we apply a event selection that an event should at least exists **2 b-jets** and **4 quark jets** satisfied p_T larger than **25 GeV** and $|\eta|$ less than **2.5**. A cutflow table and figure can help us to understand how much events are killed by the selections. We may apply 5 cuts and see the evolution of survived event numbers. The rule of cuts is shown in Table 4.1, and the cutglow is shown in Figure 4.1. As the result, we found around 1820% events will survived after the event selection.

Table 4.1: Rule of cuts. All the cuts require a kinematic limitation that $p_T > 25$ GeV and $|\eta| < 2.5$.

#Cut	Number of b-jets	Number of quark jets
C1	0	4
C2	0	5
C3	0	6
C4	1	6
C5	2	6

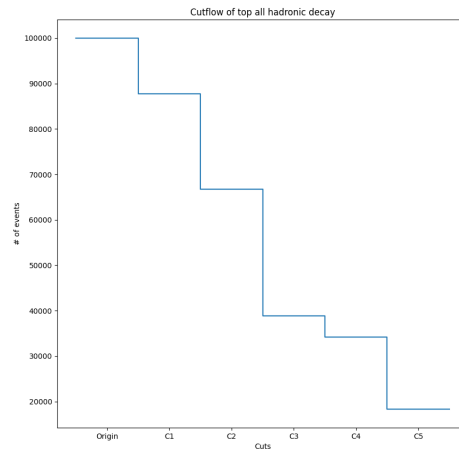


Figure 4.1: Cutflow of all hadronic top decay.

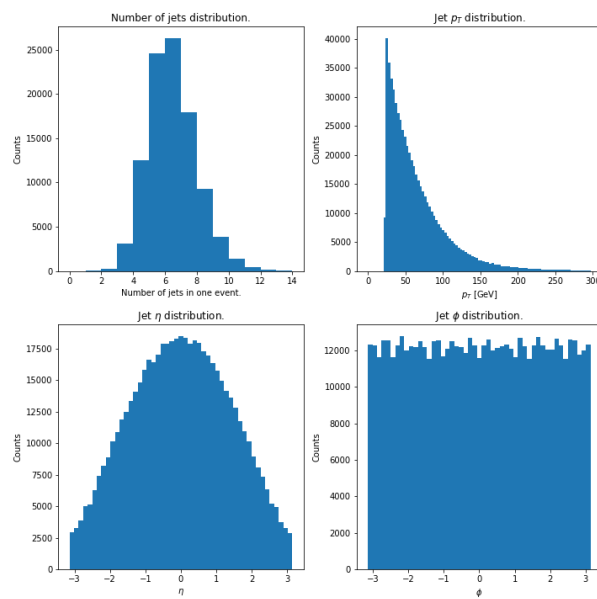


Figure 4.2: Demonstration of distributions.

4.1.2 Truth matching

The **truth matching**, which is also called **ΔR matching**, is to match the detector simulation(i.e. jet information generate by Delphes) data to truth record(i.e. parton level information). To calculate the ΔR value, we will find the daughters of top quarks, W boson and b quark. After the daughters of top quarks is found, we will find the daughters of W bosons. Finally, we will get six partons which are comes from the decay of top quark pairs. This six partons can match to the jets identically by considering their distances. The formula of calculating ΔR is:

$$\Delta R = \sqrt{\Delta\eta^2 + \Delta\phi^2} \quad (4.1)$$

By using the kinematic properties provide in parton level and detector simulation information, we can calculate the the ΔR value between each parton and jets. Using the result of calculation, we may assign each parton to a specific jet.

4.1.3 Custom barcode system

To specify the relation between each parton, and the relation between mothers and daughters, we design a barcode system which helps us to declare the relationship.

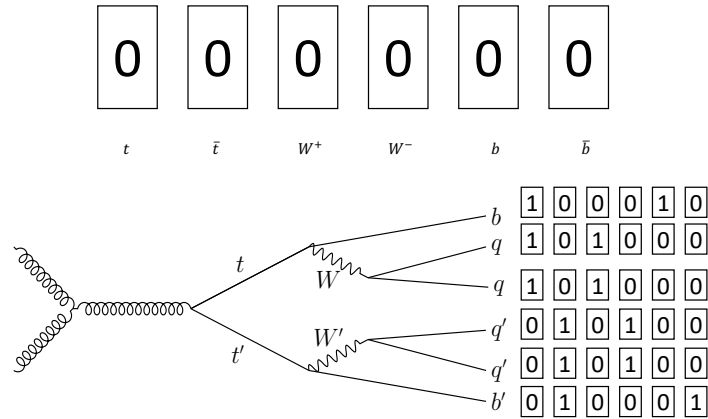


Figure 4.3: Design of barcode.

In Figure 4.3, we define a six-digits barcode, the first two digits is to show which top quark is the mother of this parton, the last four digits of sequece is to declare which

daughter of top quark is the mother of parton. In case, we can use this barcode system to break six parton(jet) candidates into two subset which contains 3 elements.

4.2 Event reconstruction

4.2.1 χ^2 reconstruction

4.2.2 Machine Learning Approach

...



Chapter 5

Discussion

In this chapter, ...

5.1 Invariant mass

5.2 Reconstruct efficiency



Chapter 6

Conclusion



Bibliography

- [1] G. Aad *et al.* [ATLAS], “Observation of a new particle in the search for the Standard Model Higgs boson with the ATLAS detector at the LHC,” Phys. Lett. B **716**, 1-29 (2012) doi:10.1016/j.physletb.2012.08.020 [arXiv:1207.7214 [hep-ex]].
- [2] S. Chatrchyan *et al.* [CMS], “Observation of a New Boson at a Mass of 125 GeV with the CMS Experiment at the LHC,” Phys. Lett. B **716**, 30-61 (2012) doi:10.1016/j.physletb.2012.08.021 [arXiv:1207.7235 [hep-ex]].
- [3] P. A. Zyla *et al.* [Particle Data Group], PTEP **2020**, no.8, 083C01 (2020) doi:10.1093/ptep/ptaa104
- [4] A. Quade, “Top quark physics at hadron colliders,” Eur. Phys. J., 48, (2008)
- [5] A. M. Sirunyan *et al.* [CMS], “Measurement of the top quark mass in the all-jets final state at $\sqrt{s} = 13$ TeV and combination with the lepton+jets channel,” Eur. Phys. J. C **79**, no.4, 313 (2019) doi:10.1140/epjc/s10052-019-6788-2 [arXiv:1812.10534 [hep-ex]].
- [6] M. Aaboud *et al.* [ATLAS], “Top-quark mass measurement in the all-hadronic $t\bar{t}$ decay channel at $\sqrt{s} = 8$ TeV with the ATLAS detector,” JHEP **09**, 118 (2017) doi:10.1007/JHEP09(2017)118 [arXiv:1702.07546 [hep-ex]].
- [7] T. McCarthy, “Measurement of the Top Quark Mass in the All-Hadronic Top-Antitop Decay Channel Using Proton-Proton Collision Data from the ATLAS Experiment at a Centre-of-Mass Energy of 8 TeV,” CERN-THESIS-2015-275.
- [8] M. Paganini [ATLAS], J. Phys. Conf. Ser. **1085**, no.4, 042031 (2018) doi:10.1088/1742-6596/1085/4/042031 [arXiv:1711.08811 [hep-ex]].

- [9] A. Vaswani *et al.*, “Attention is all you need,” Advances in Neural Information Processing Systems, NIPS (2017)
- [10] Hung-Yi, Lee, “NTU course lecture note-Transformer”

