

# Leveraging **Quantum Annealer** to **identify event topologies** at High energy colliders

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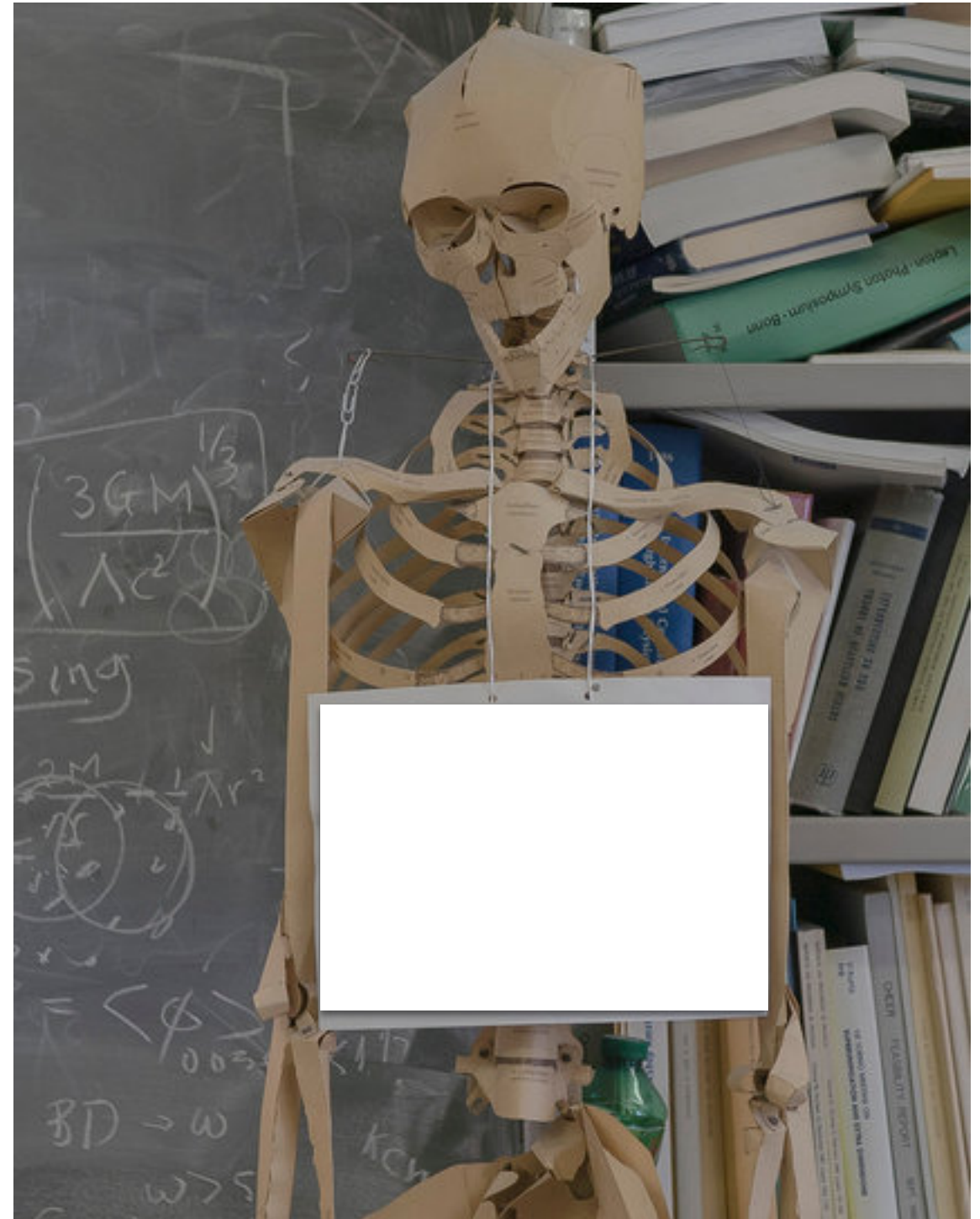
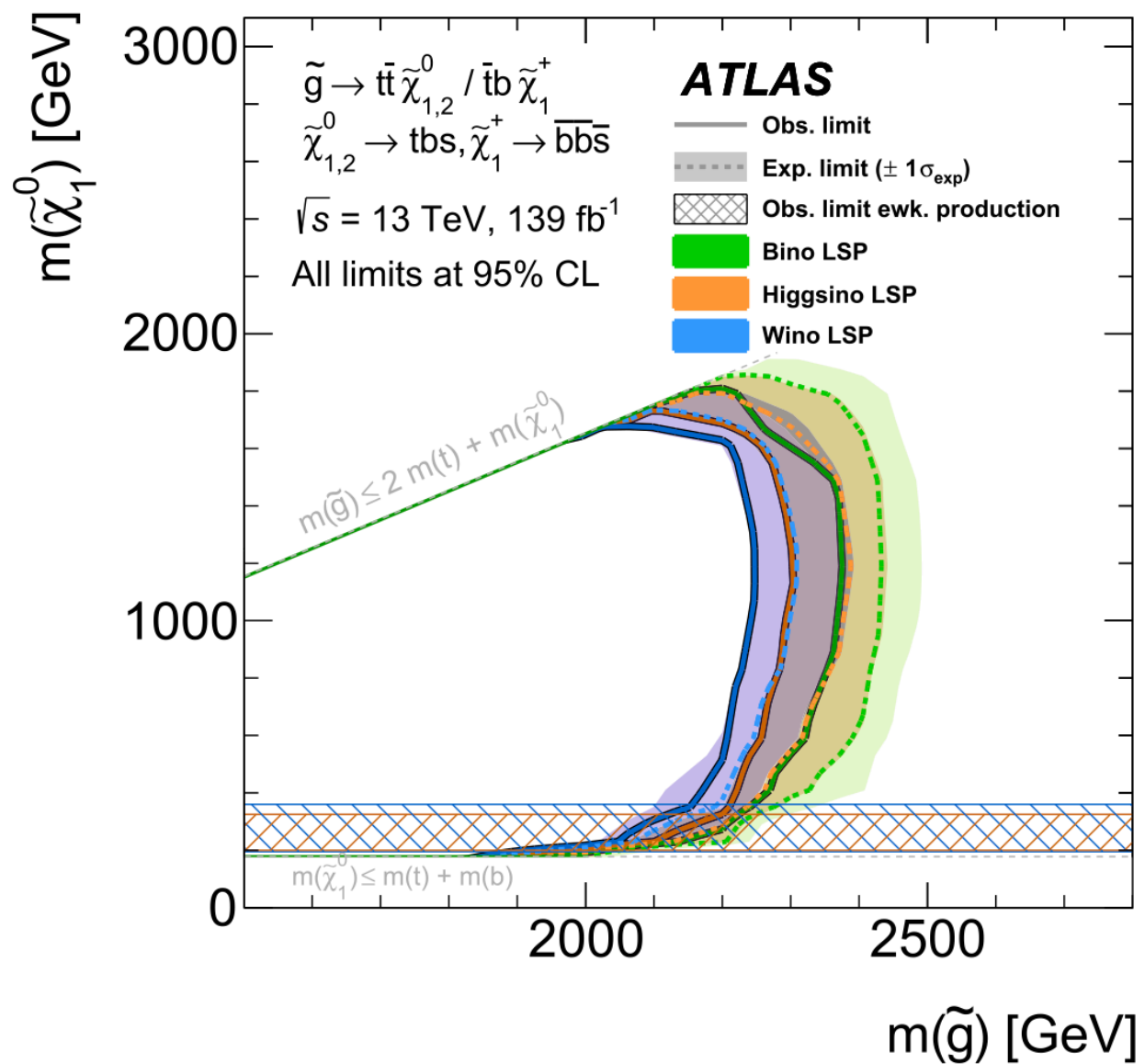
based on arXiv:2111.07806

with Minho Kim, Pyungwon Ko, Jae-hyeon Park


**The 2nd Asian-European-Institutes Workshop for BSM (2022)**

# With the LHC

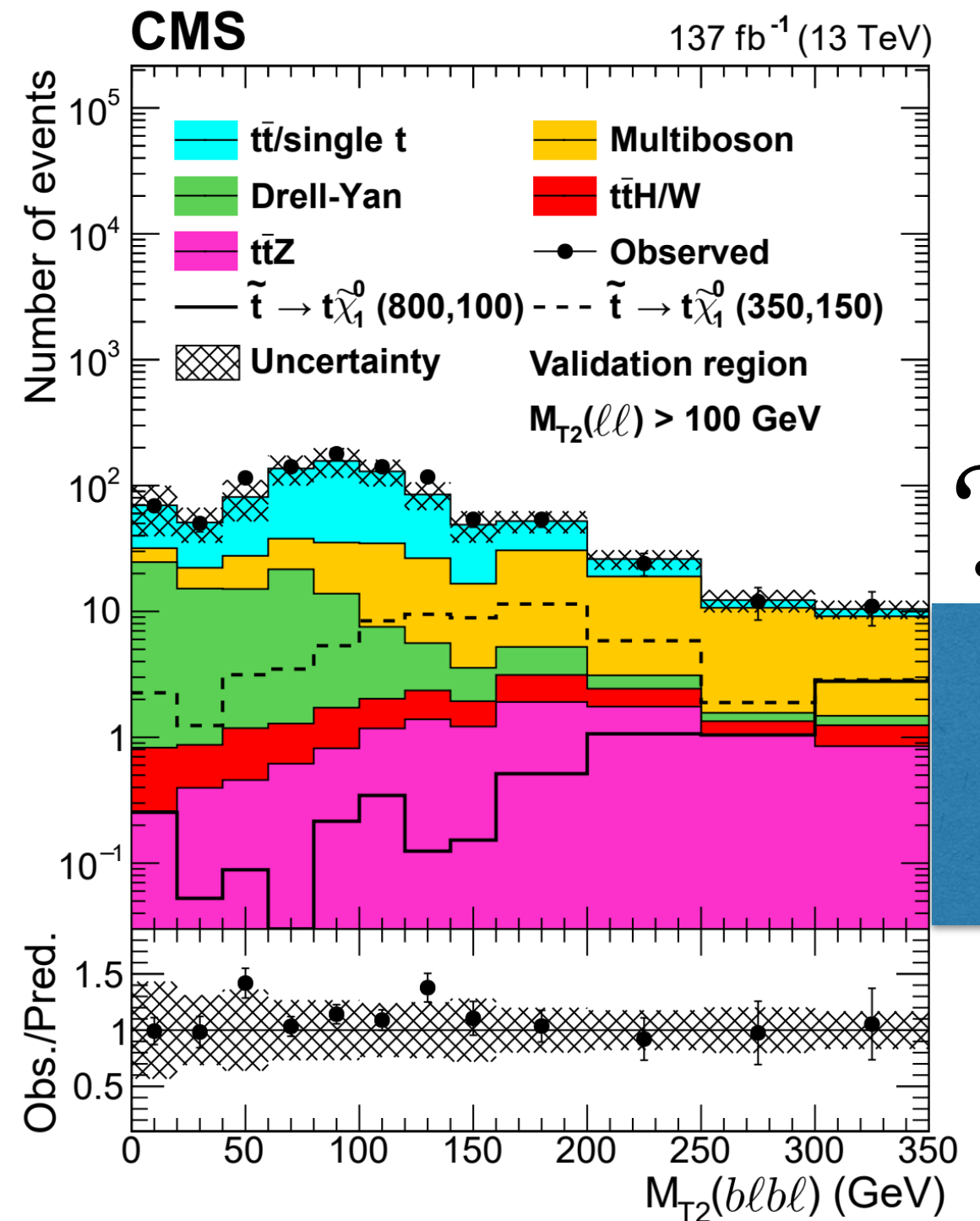
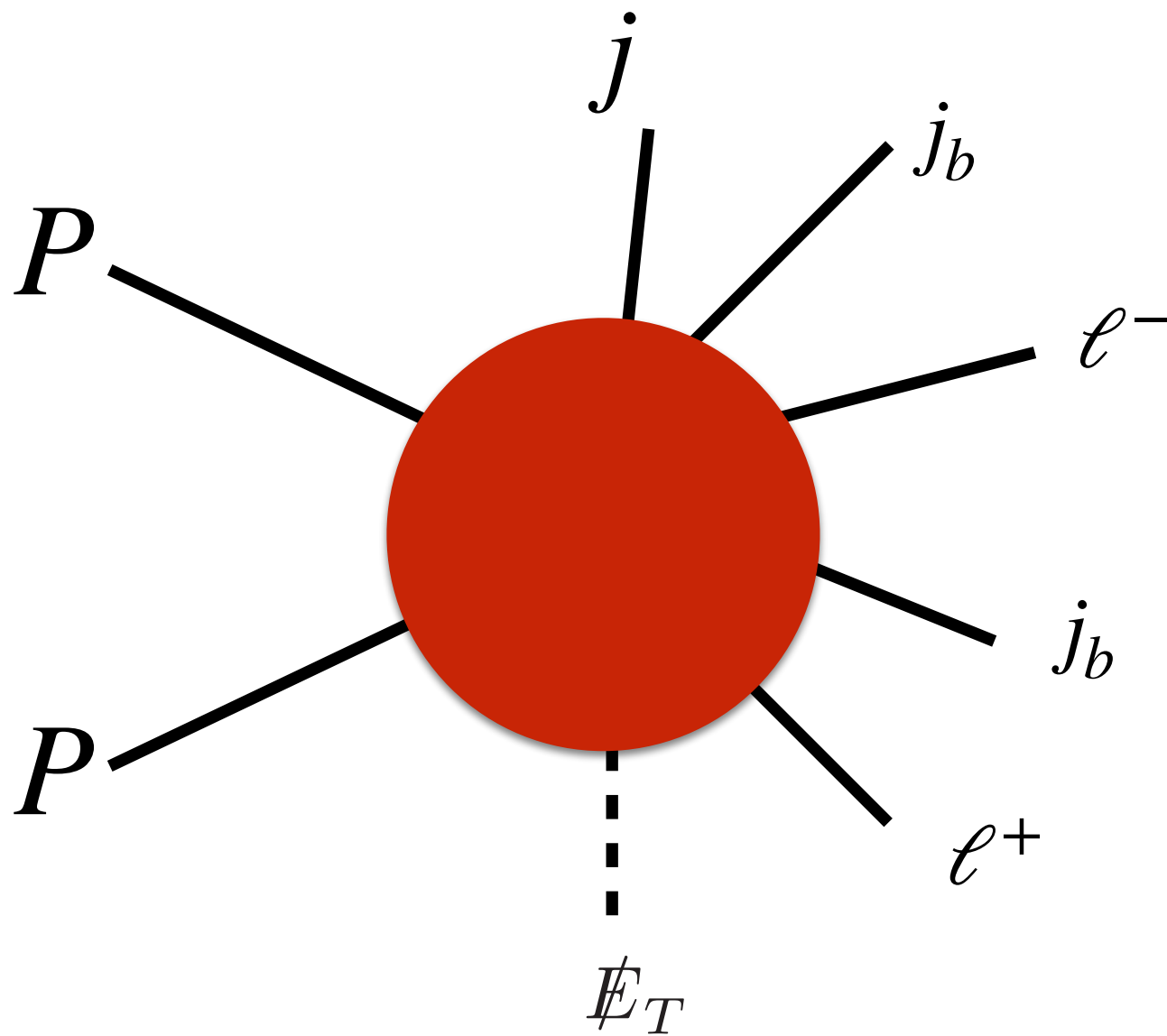
- The long lived king is **dead**!?



# Hunt for new physics afterwards

- 
1. **Anomaly detection** (different from SM expectations)
    - Need to have precise tools (importance of MC)
  2. Try to interpret a new signal with **various** model assumptions or **Model-independent way** so called simplified model
    - For each model, we start with **specific** "feynman-diagram"  
(**event-topology**, without specific spin assignment.)
    - Determine parameters (spin, mass) with various methods

# Example: anomaly



a simple kinematic variable



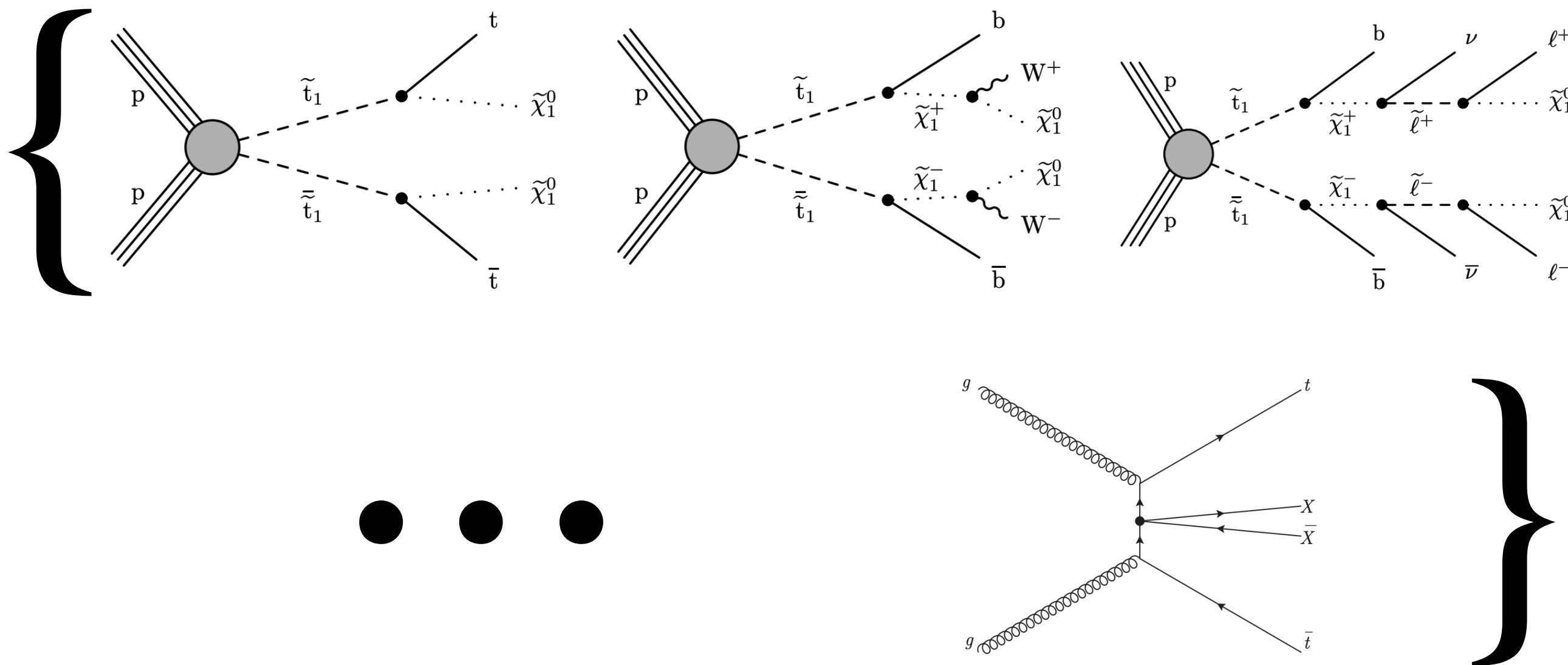
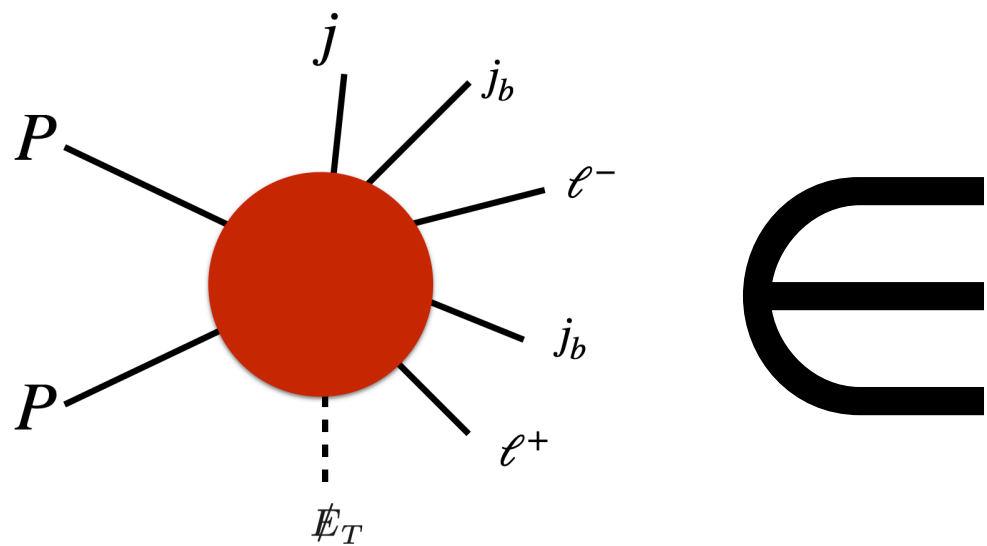
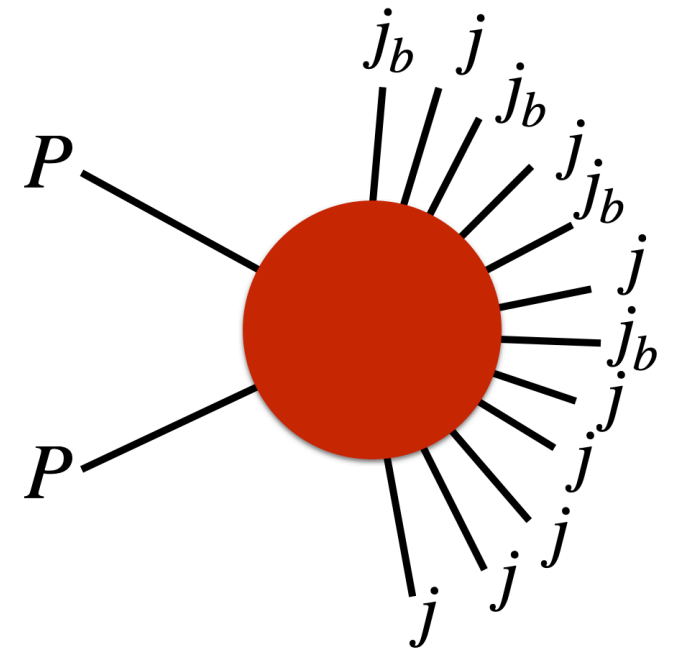
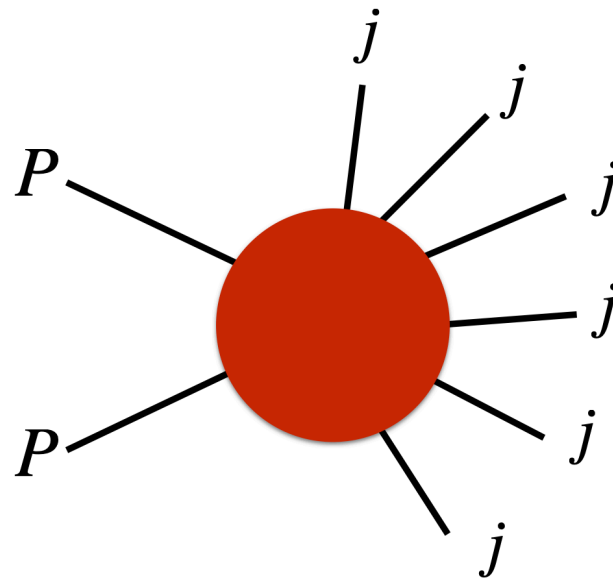
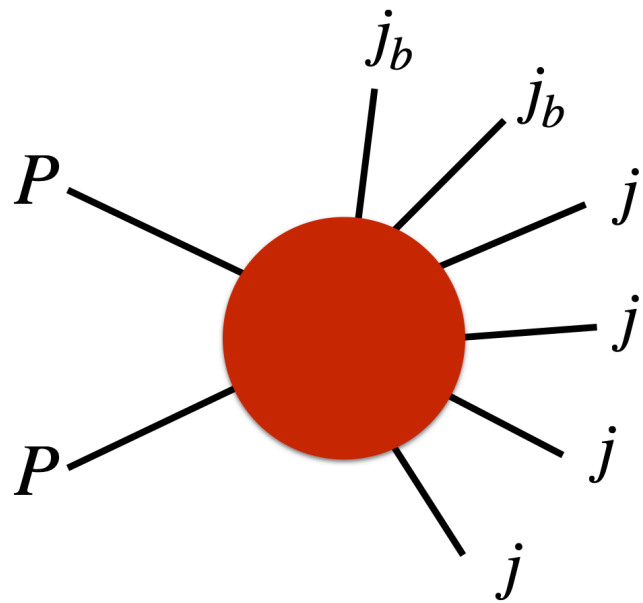


diagram from Lian-Tao Wang et.al. arxiv:1303.6638

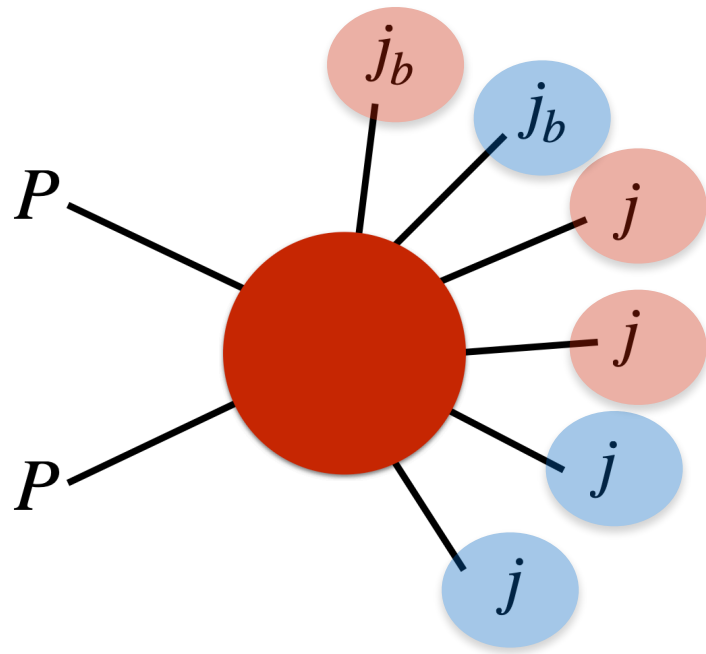
# Purely bottom-up approach

1. Figure out what is the relevant **event-topology** behind anomalous (deviation from SM) events.
  2. Check the **mass** spectrum.
  3. Check **spin** configuration.
- So far, there are very few literatures for **#1**.  
Here I will introduce how one can **identify the event-topology**

# Our examples (multi-jets)



1. Under the a simple assumption:  $pp \rightarrow X, Y \rightarrow \{j_x\} \cup \{j_y\}$   
(No prejudices on  $X$  and  $Y$ )
2. Find a right **combination** to reconstruct  $X$  and  $Y$  particles.  
→ Read off information on **Mass** and **Spin** from event reconstruction.



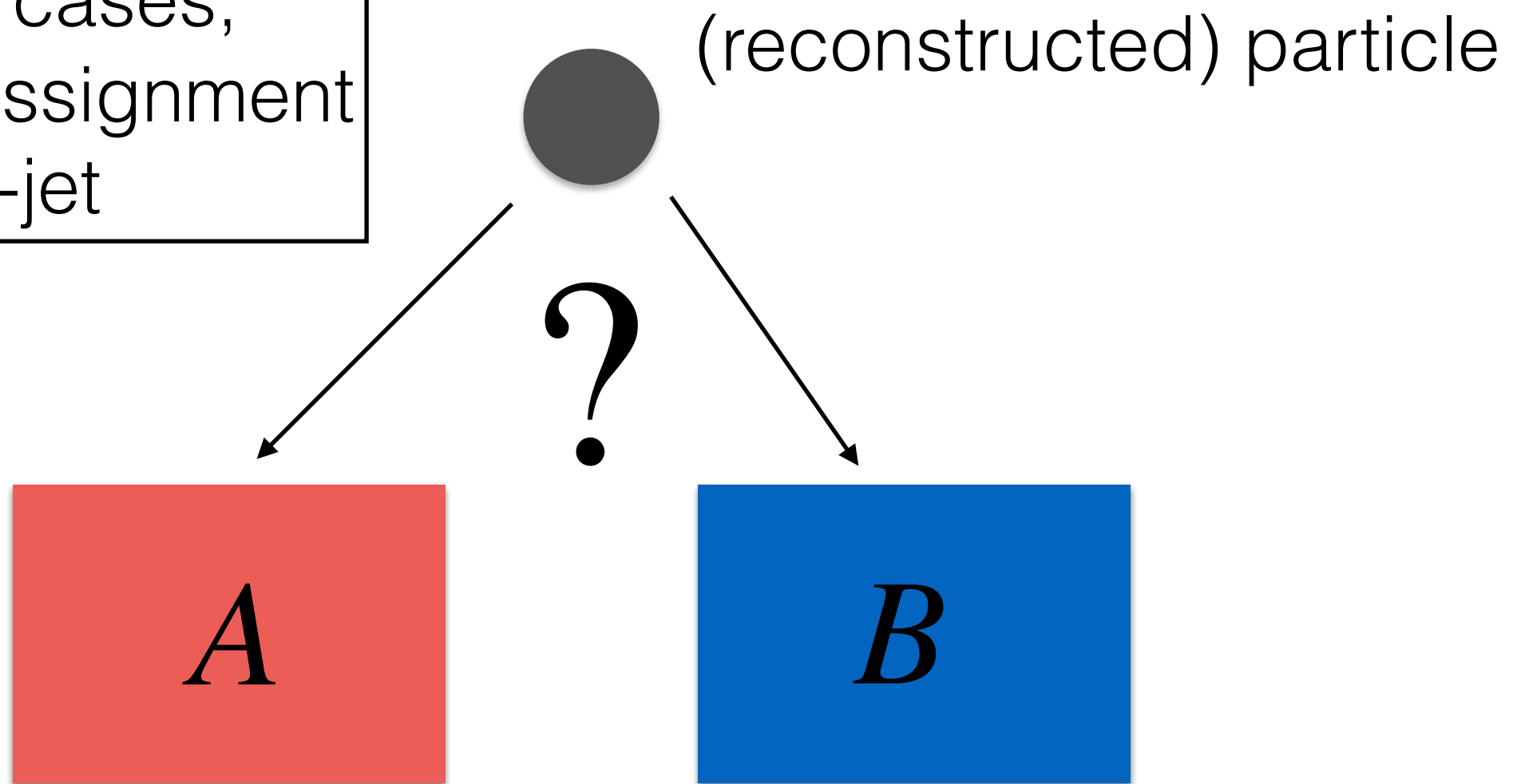
- Standard example of six jets

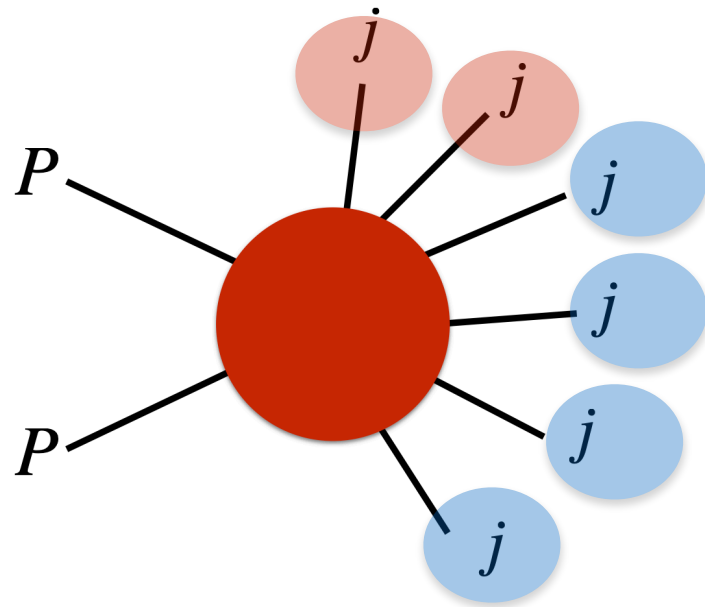
$$pp \rightarrow t\bar{t} \rightarrow \{j_b, (W \rightarrow jj)\} \cup \{j_b, (W \rightarrow jj)\}$$

(when A and B have same mass)

- Right answer is  $(n_A, n_B) = (3, 3)$

$2^6 = 64$  cases,  
no special assignment  
for b-jet



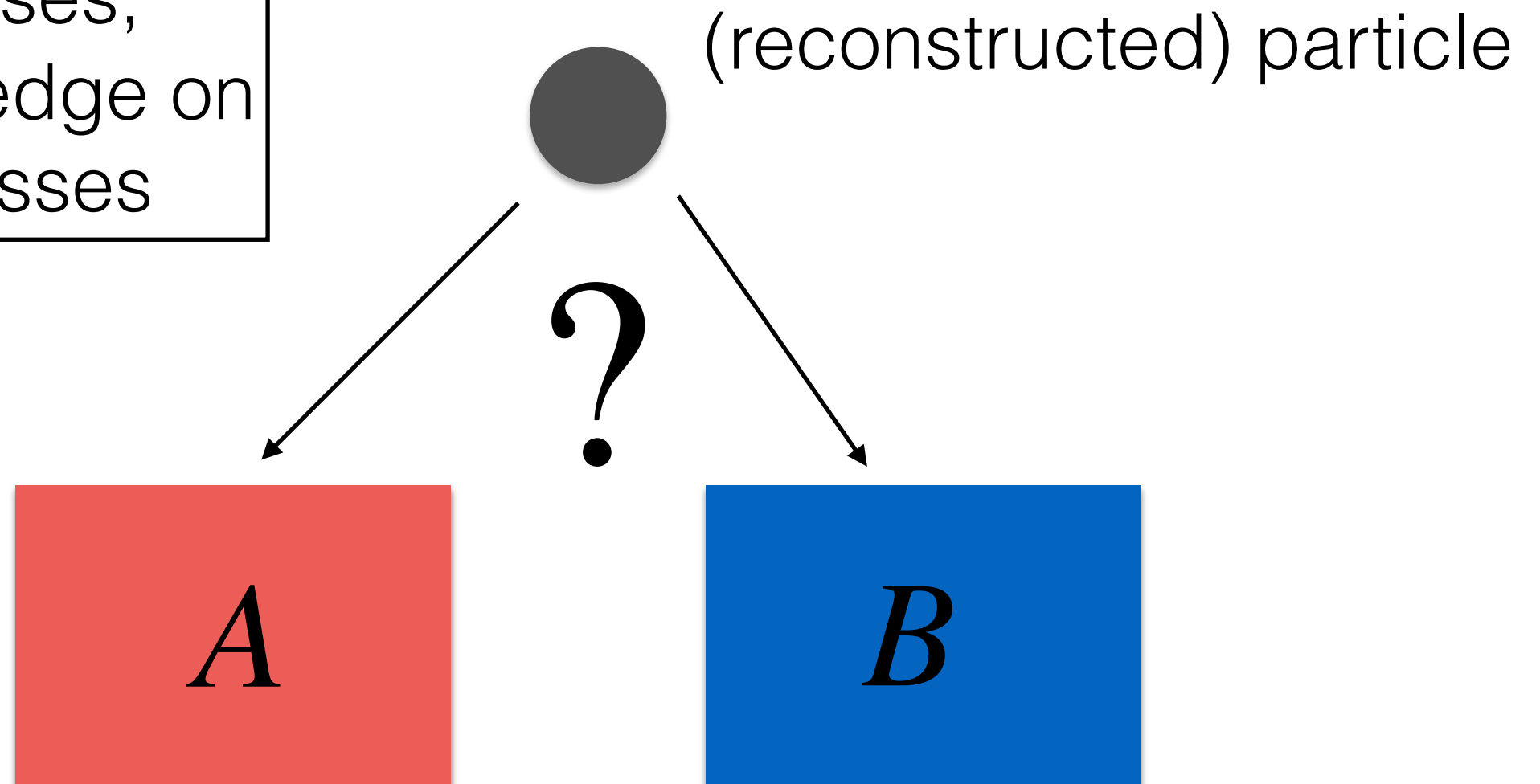


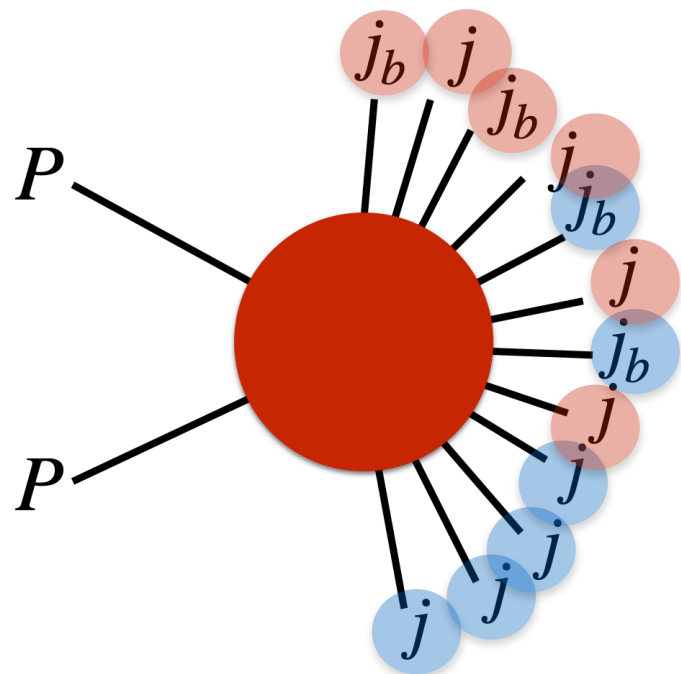
- Different mother particles

$$pp \rightarrow ZH \rightarrow \{j, j\} \cup \{(W \rightarrow jj), (W^* \rightarrow jj)\}$$

- Right answer is  $(n_A, n_B) = (2, 4)$

$2^6 = 64$  cases,  
no prior knowledge on  
A and B masses





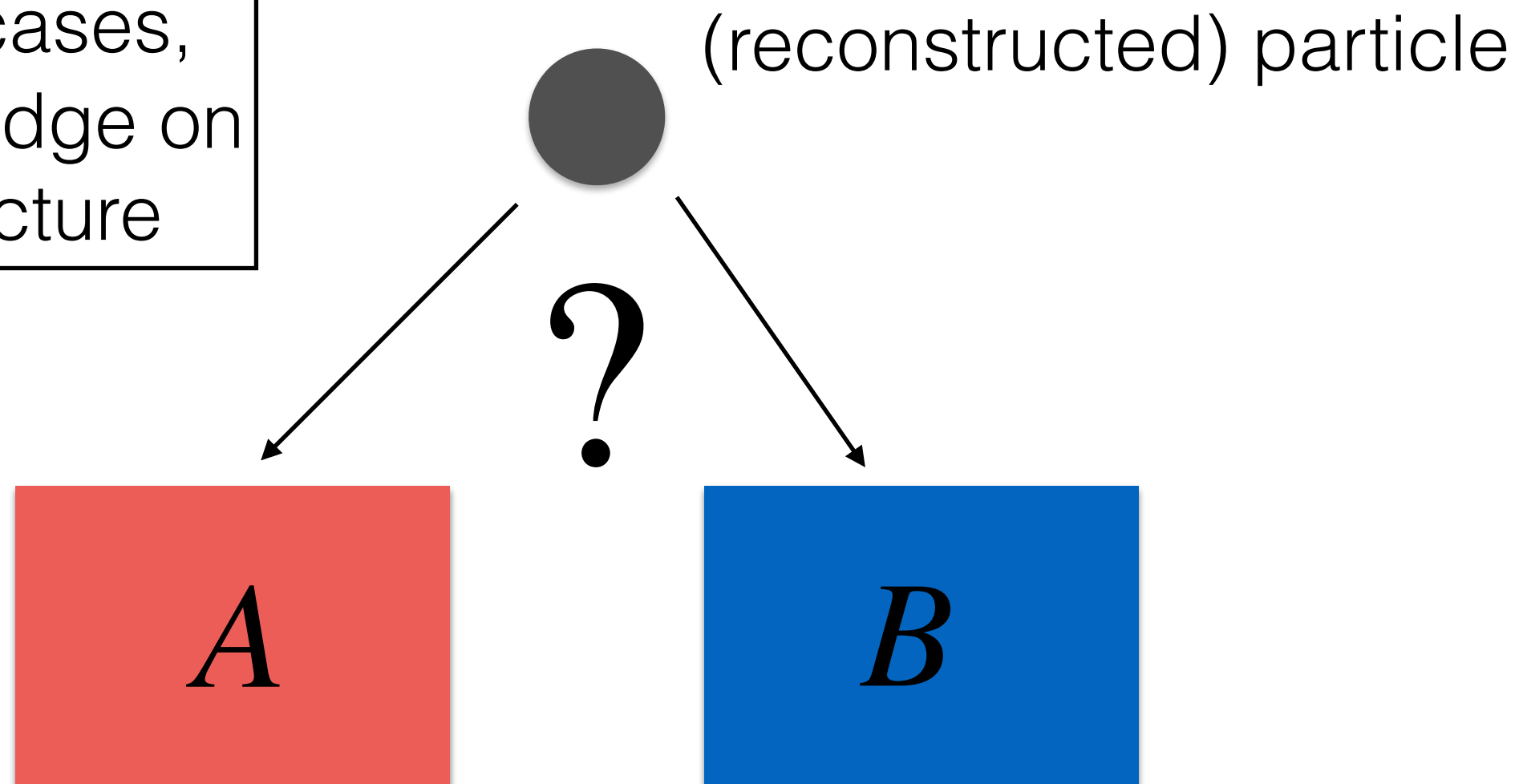
- Complicate situation ( 12 jets)

$$pp \rightarrow o\tilde{o} \rightarrow \{t, \bar{t}\} \cup \{t, \bar{t}\}$$

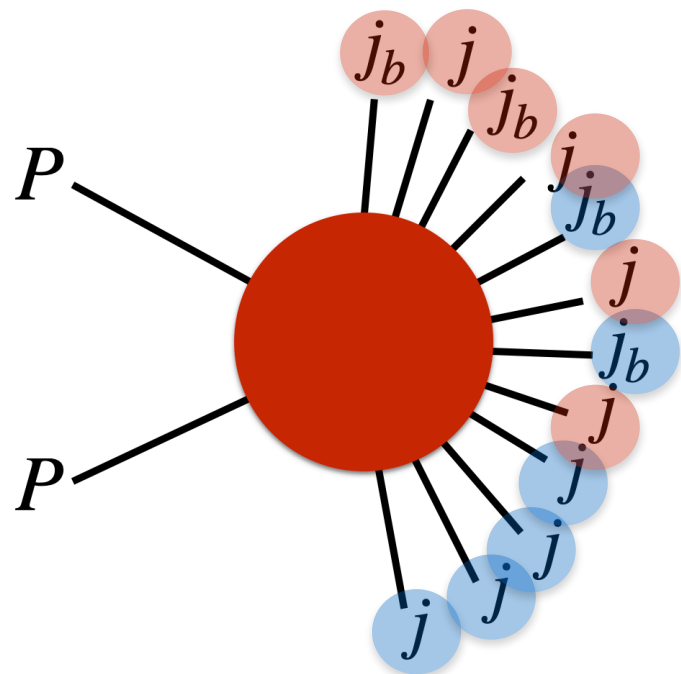
$$o \rightarrow t\bar{t} \rightarrow \{j_b, (W \rightarrow jj)\} \cup \{j_b, (W \rightarrow jj)\}$$

$$\tilde{o} \rightarrow t\bar{t} \rightarrow \{j_b, (W \rightarrow jj)\} \cup \{j_b, (W \rightarrow jj)\}$$

$2^{12} = 4096$  cases,  
no prior knowledge on  
a decay-structure







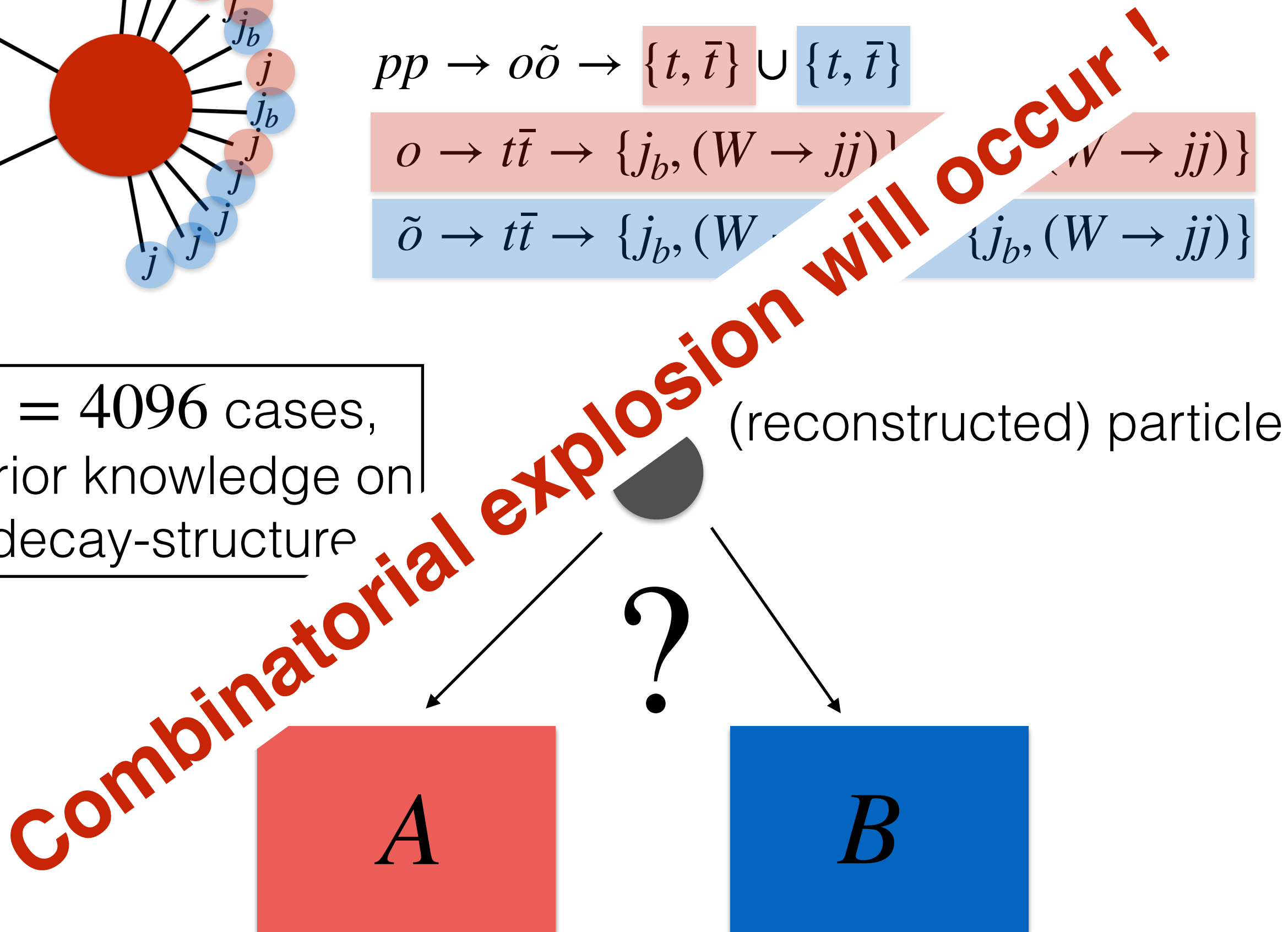
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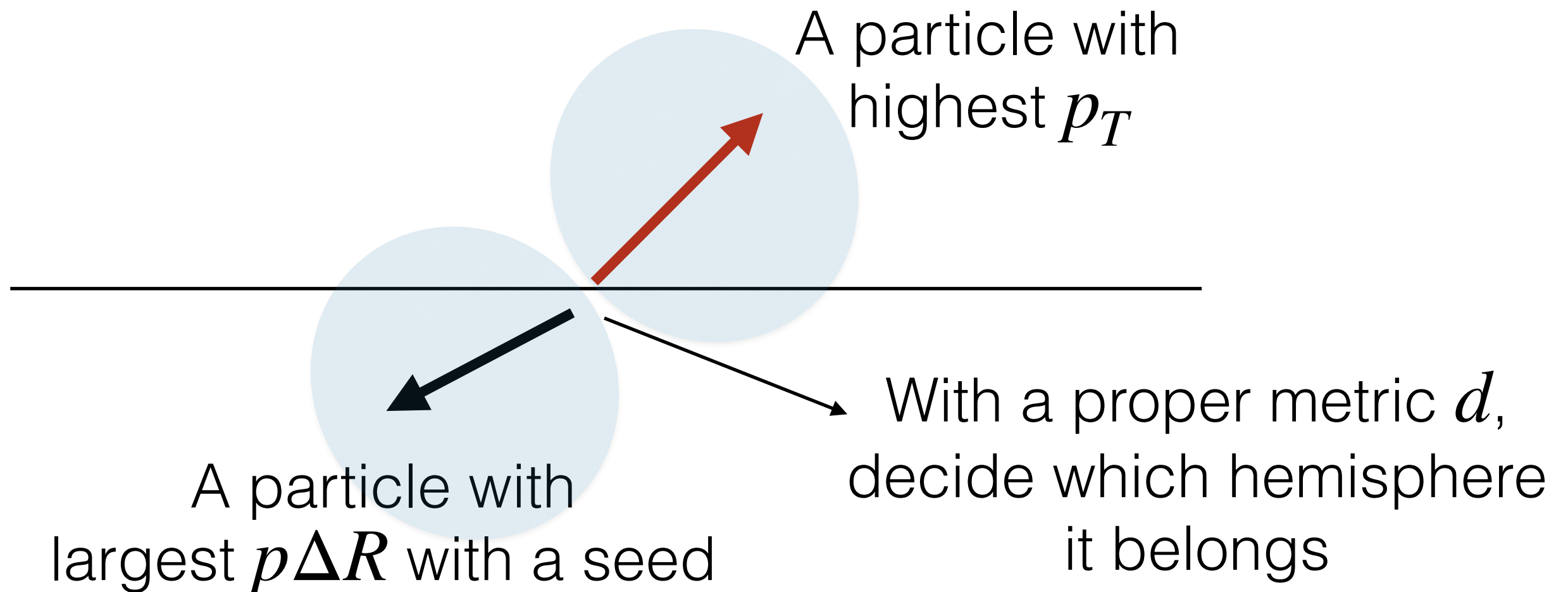


# An algorithm ?

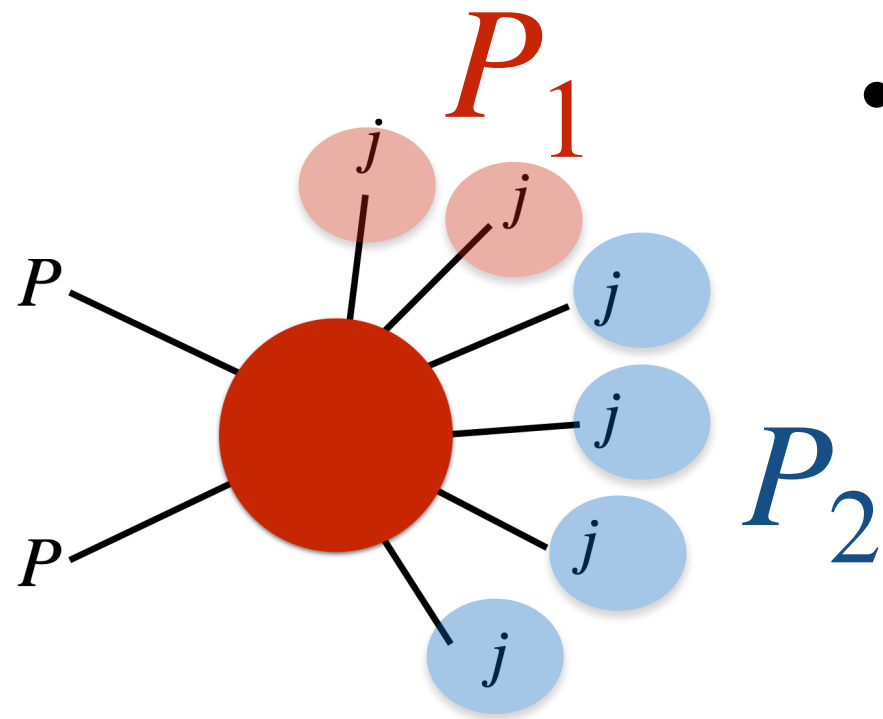
- With the only assumption of  $2 \rightarrow (2 \rightarrow n)$  process
  - No special treatment on any flavor-tagged particle
  - No assumption on  $M_A$  and  $M_B$
  - No assumption on any decaying structure
- What could be a good guide line ?

# A Classic algorithm

- Hemisphere method: a **seed**-based method (iterative and converge)



# Non-geometric algorithm

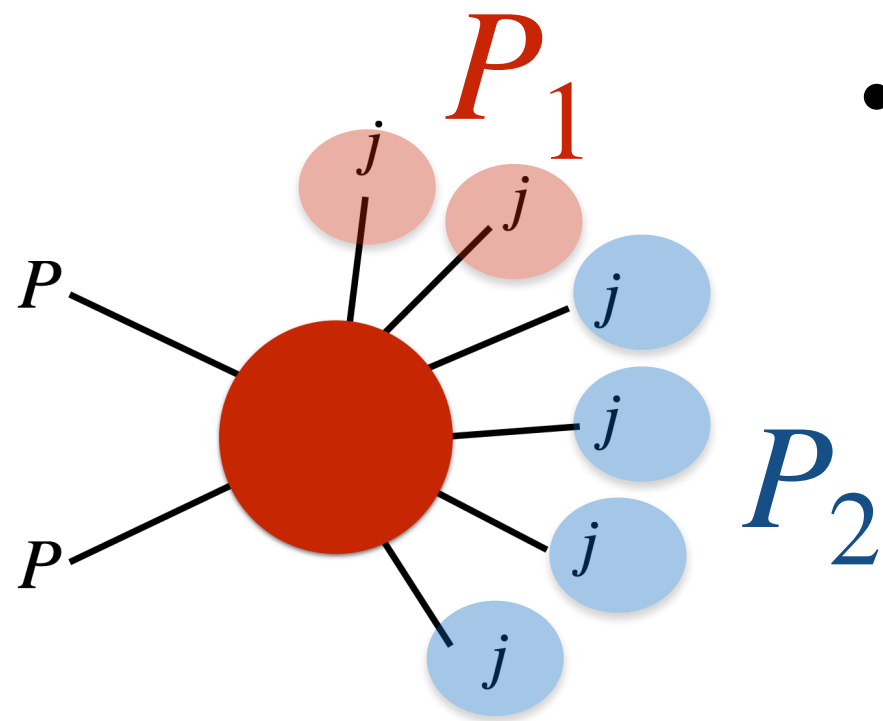


- For each assignment, calculate invariant mass

$$(M_A^2, M_B^2) = (P_1^2, P_2^2)$$

- Try to **minimize** the mass difference  $H = (M_A^2 - M_B^2)^2$

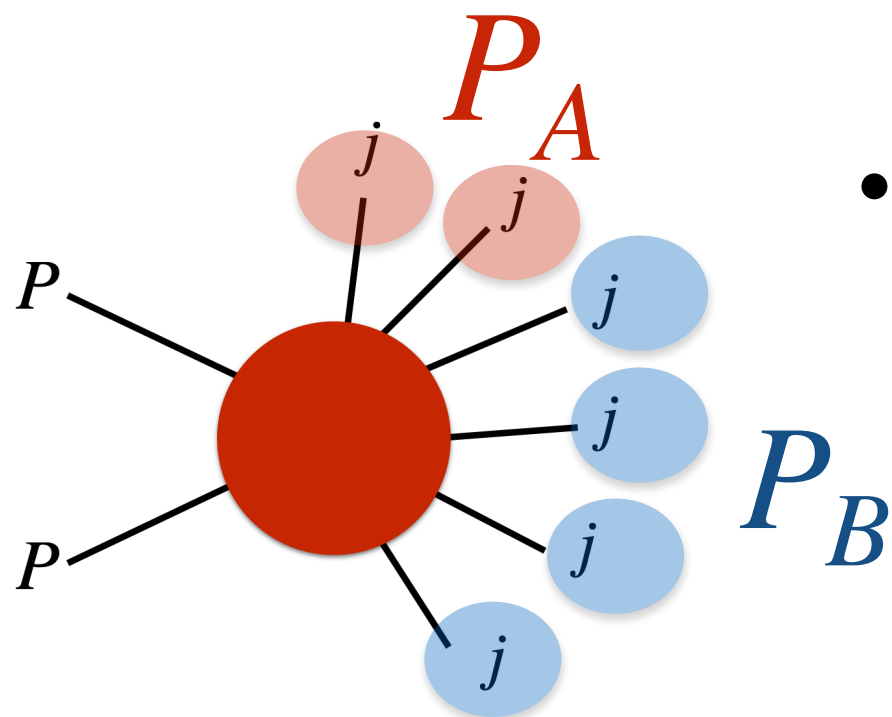
# Non-geometric algorithm



- For each assignment, calculate invariant mass

$$(M_A^2, M_B^2) = (P_1^2, P_2^2)$$

- Try to **minimize** the mass difference  $H = (M_A^2 - M_B^2)^2$
- **How can we deal with the case of  $M_A \neq M_B$  ?**



- Try to **minimize** the mass difference

$$H = (M_A^2 - M_B^2)^2$$

- **How can we deal with the case of  $M_A \neq M_B$  ?**

+ (even with  $M_A = M_B$ ) we need to handle

- 1) off-shell mass due to the width of A and B
- 2) from smearing effects due to imperfect detectors

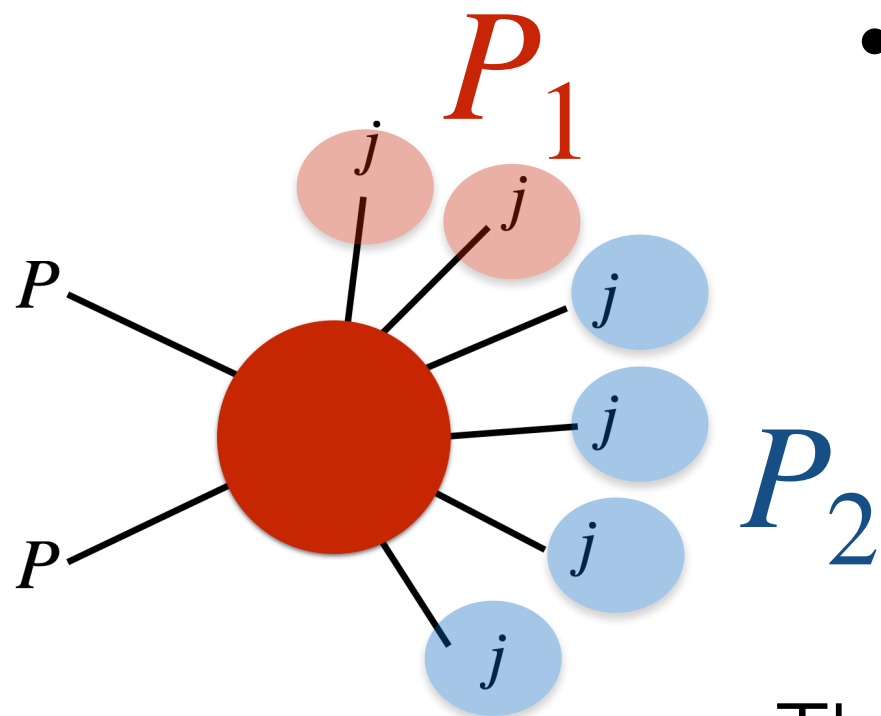
- One suggestion: Add a regularization term of  $\lambda(P_1^2 + P_2^2)$   
( $\lambda$  is a dimension full "**hyper-parameter**")



- 2 → 2 process:  $\{p_i\} \rightarrow P_1 \cup P_2$   
Using a binary operation  $x_i \in \{0,1\}$

For  $p_i$  to be either in  $P_1$  ( $x_i = 1$ ) or in  $P_2$  ( $x_i = 0$ )

$$P_1 = \sum_i p_i x_i, \quad P_2 = \sum_i p_i (1 - x_i)$$



- Try to **minimize**

$$H = (P_1^2 - P_2^2)^2 + \lambda(P_1^2 + P_2^2)$$

**for each "assignment" ?!**

- This problem now becomes well-known...

# Minimization using Ising model

- If we replace  $x_i \rightarrow \frac{1 + s_i}{2}$  with  $s_i \in \{+1, -1\}$

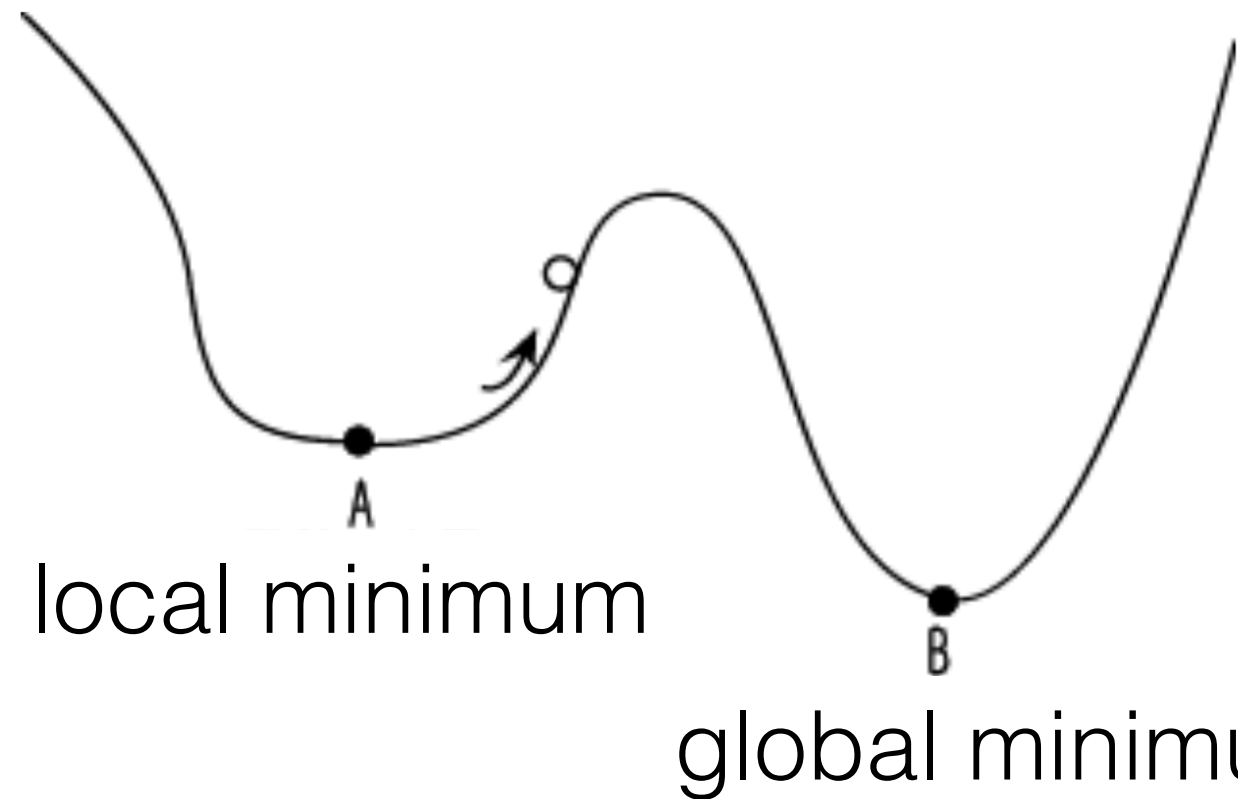
$$H = (P_1^2 - P_2^2)^2 \rightarrow H + \lambda (P_1^2 + P_2^2)$$

$$= \sum_{i,j} \left( C_{ij} + 2\lambda S_{ij} \right) s_i s_j + \sum_i \left( J_i - 2\lambda \sum_j S_{ij} \right) s_i$$

- To maintain the importance of original  $H$ ,

we take  $\lambda = \frac{\min(C_{ij})}{\max(S_{ij})}$

# "Classic" minimization method (for ising hamiltonian)



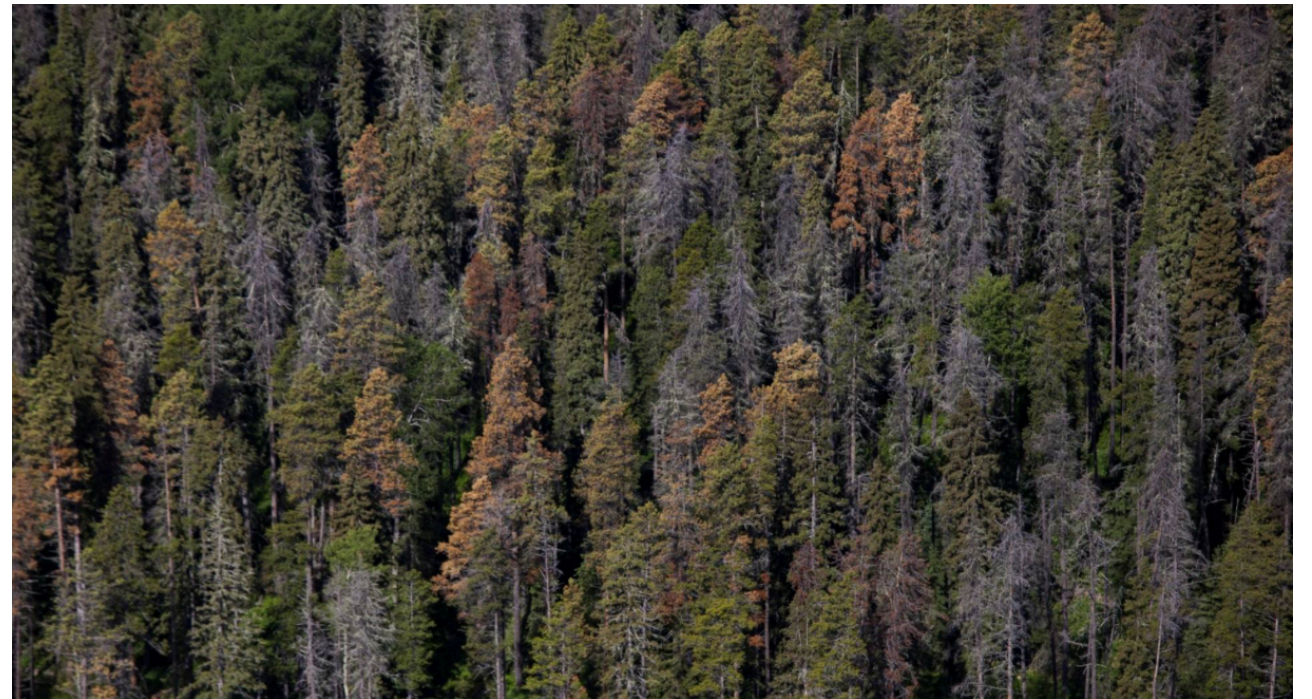
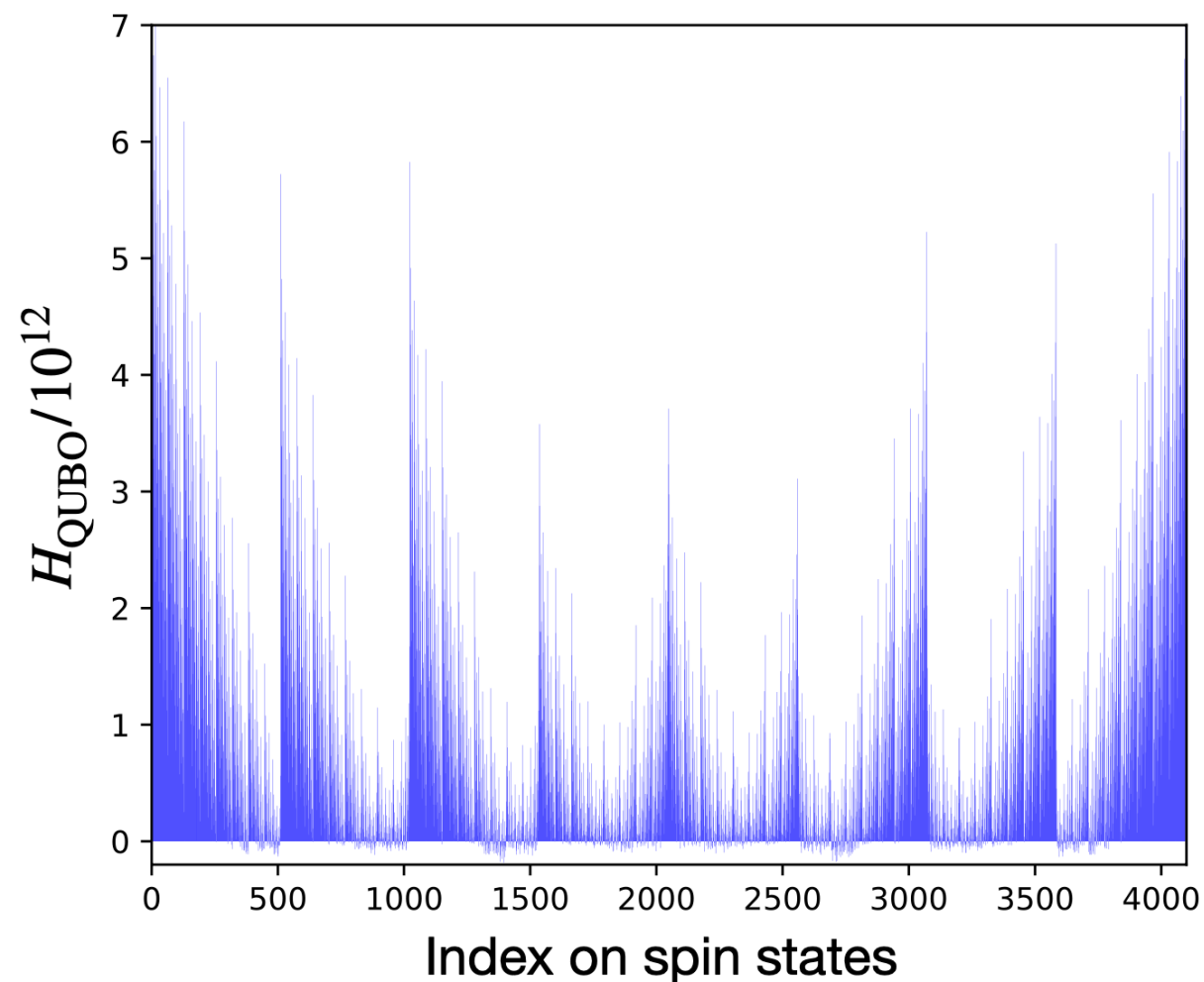
## Simulated annealing

- Go to the next spin state  $s_n \rightarrow s_{n+1}$ 
    - 1) If  $E_n > E_{n+1}$  : go to the lower energy
    - 2) If  $E_n < E_{n+1}$  , go with a probability of  $e^{-\frac{E_{n+1} - E_n}{k_B T}}$  to **jump out**
- (A "temperate  $T \rightarrow 0$ . With large T, SA can jump out local minimum)

But our "mindless"  
=**minimally assumed** Collider example  
is **not so easy**  
**for a classical SM**

# Combinatorial complexity arises (for a random Ising model)

Landscape of energy distribution



$\uparrow \uparrow \uparrow \uparrow \uparrow \uparrow \rightarrow \uparrow \uparrow \uparrow \uparrow \uparrow \downarrow \rightarrow \dots \rightarrow \downarrow \downarrow \downarrow \downarrow \downarrow \downarrow$  ( $n_{\text{spin}} = 2^{12} = 4096$ )

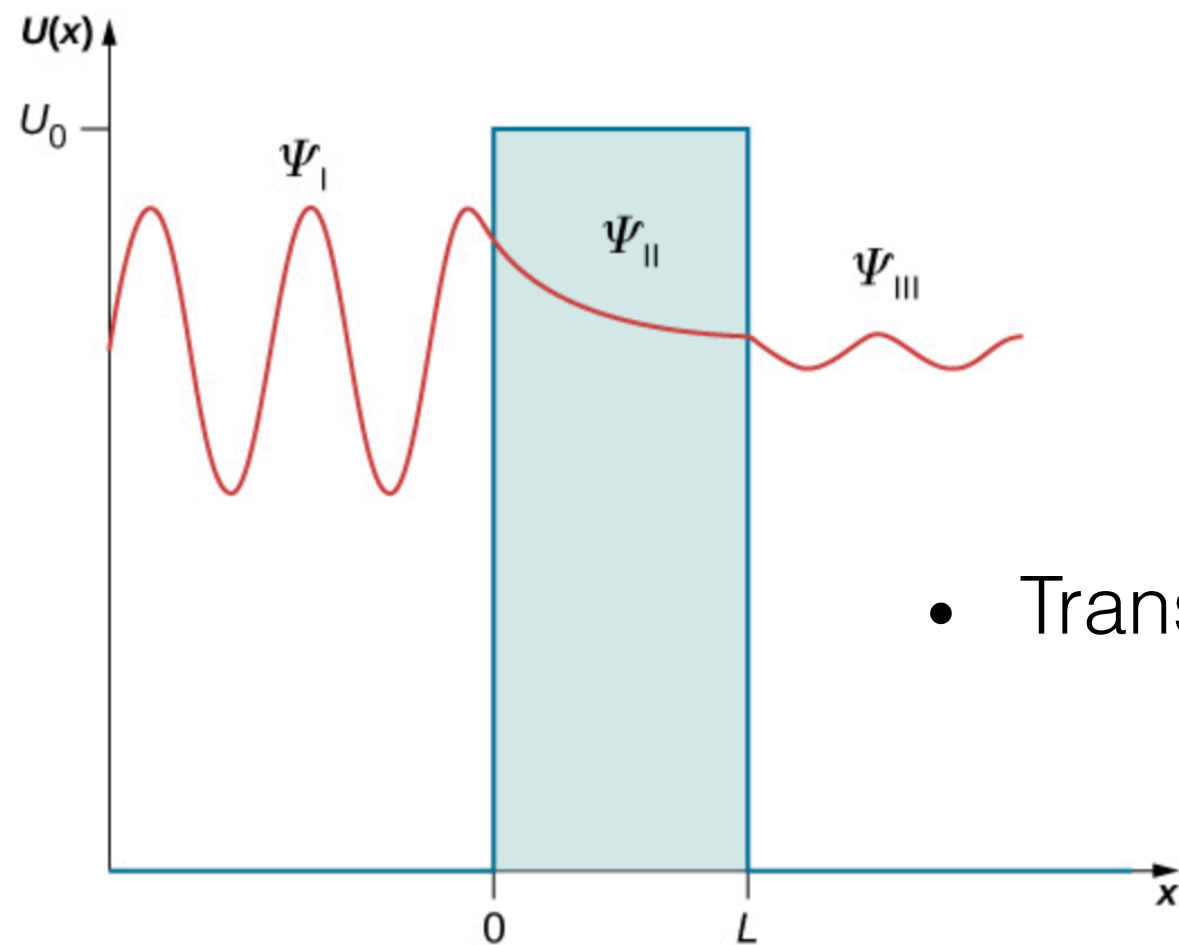
**SA cannot jump this random potential!**

Any solution  
we can have?



# The Quantum thing...

- In the class of **undergraduate QM**, we learned



- Transmission coefficient  $T \propto e^{-2L\sqrt{U_0 - E}}$

1) The effect of energy difference becomes mild

2) Effective for **shallow barrier** !

- If there is a **machine which can realize Quantum tunneling**,  
**our problem is a simple and good example** to demonstrate  
an **advantage from Quantum tunneling**

# Quantum Computer

- Gate type : IBM just announced 433 qubit QPU.  
(An application of this type: Yamazaki's talk)
- **Quantum annealer**: over 5000 qubits (Here)



currently 433 Qubits (IBM Eagle)

## Scaling IBM Quantum technology

IBM

IBM Q System One (Released)

(In development)

Next family of IBM Quantum systems

2019

2020

2021

2022

2023

and beyond

27 qubits  
*Falcon*

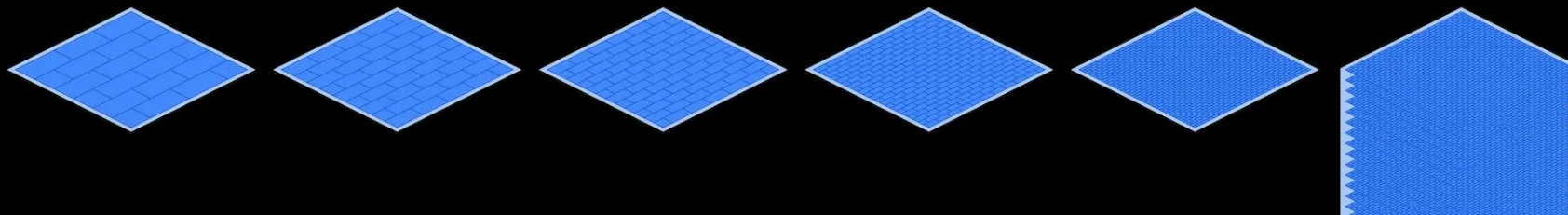
65 qubits  
*Hummingbird*

127 qubits  
*Eagle*

433 qubits  
*Osprey*

1,121 qubits  
*Condor*

Path to 1 million qubits  
and beyond  
*Large scale systems*



Key advancement

Key advancement

Key advancement

Key advancement

Key advancement

Key advancement

Optimized lattice

Scalable readout

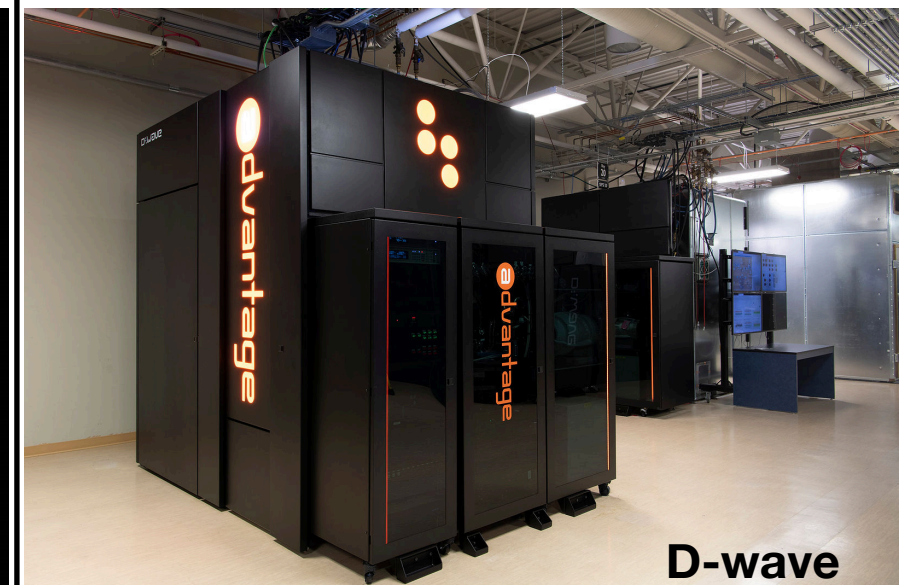
Novel packaging and controls

Miniaturization of components

Integration

Build new infrastructure,  
quantum error correction

## Quantum Annealer



D-wave

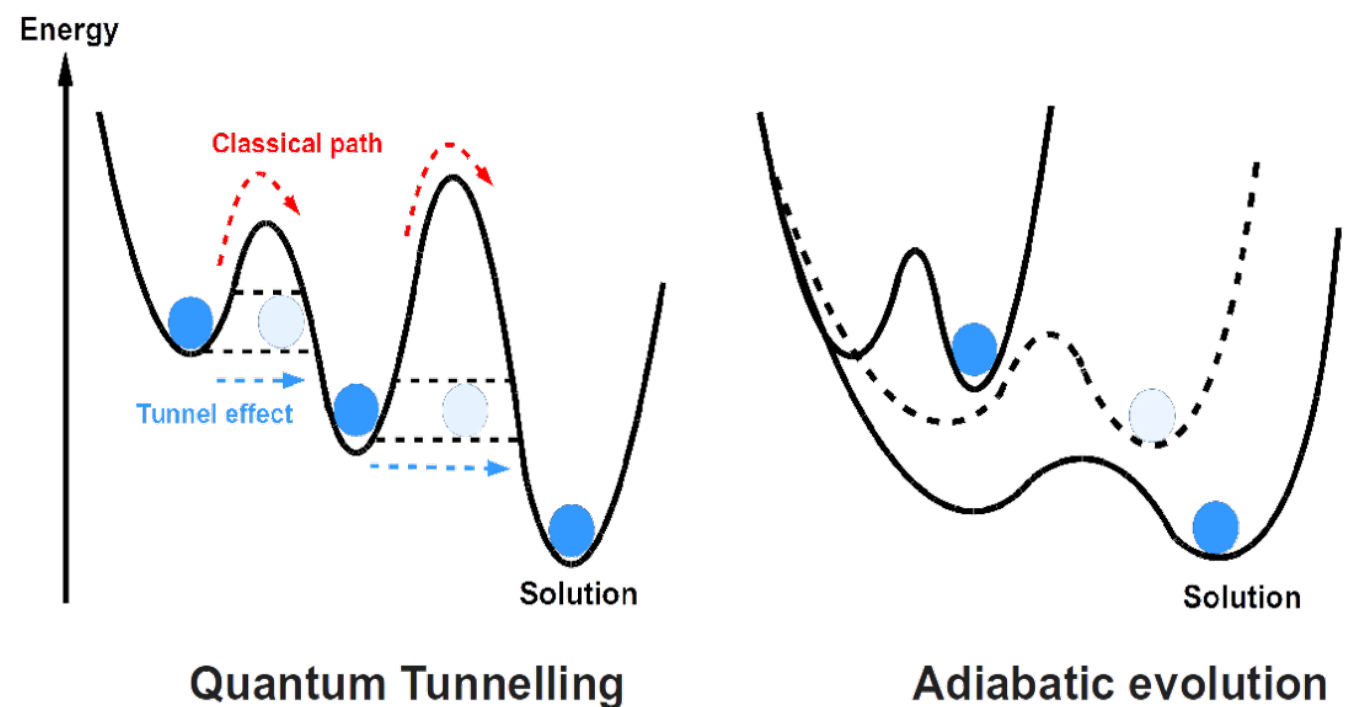
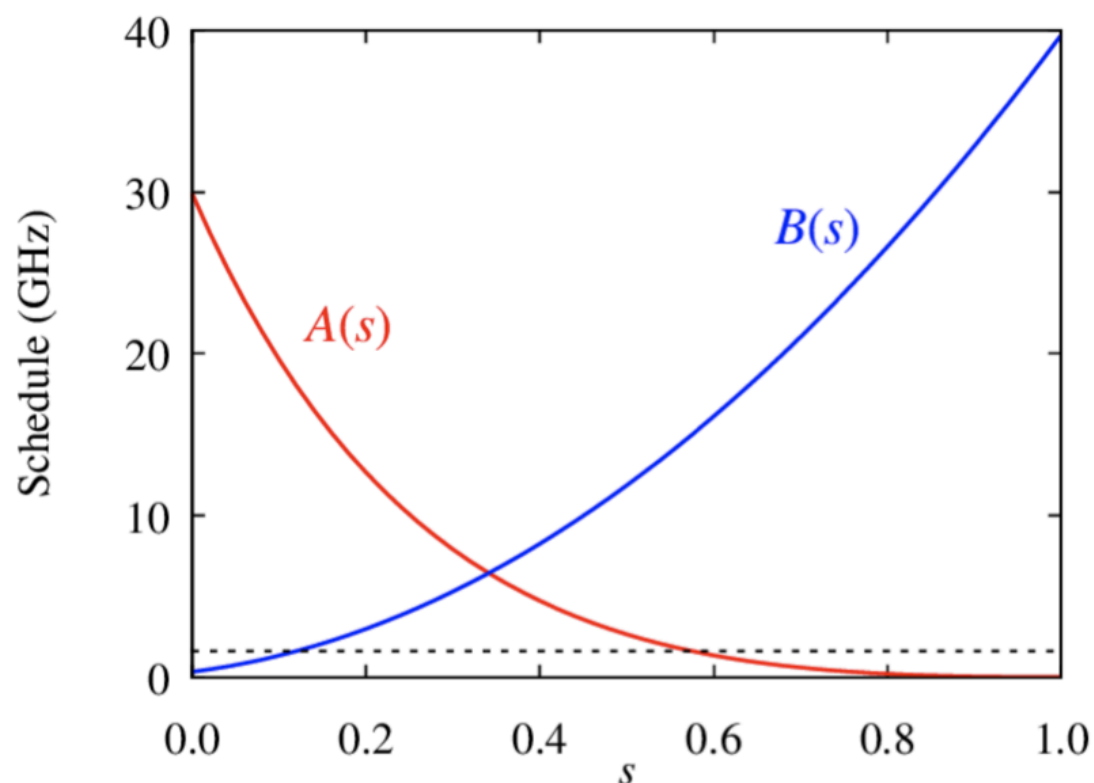
Temperature: below  $1.5 \times 10^{-2} \text{K}$   
dimension:  $3\text{m} \times 2.1\text{m} \times 3\text{m}$   
Weight: 3800kg  
Power: (max) 25kW

# Quantum Annealing method

- With adiabatic theorem, we can find the ground state of a complicate hamiltonian  $H_{\text{QUBO}}$  starting from simple  $H_0$ .

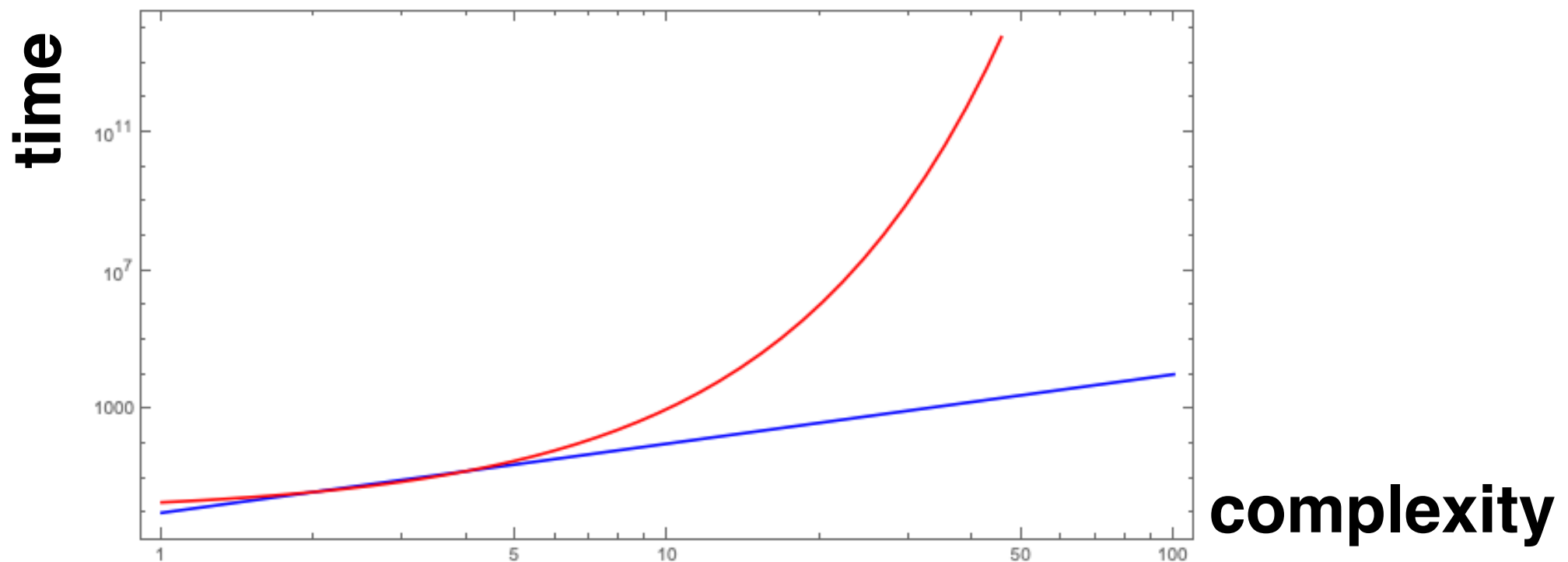
$$H_{\text{QA}} = A(s) H_0 + B(s) H_{\text{QUBO}} \quad \text{with } H_0 = \sum \sigma_i^x \text{ and } H_{\text{QUBO}} = \sum J_{ij} \sigma_i^z \sigma_j^z + \sum h_i \sigma_i^z$$

(T. Kadowaki and H. Nishimori, Quantum annealing in the transverse ising model, 1998)



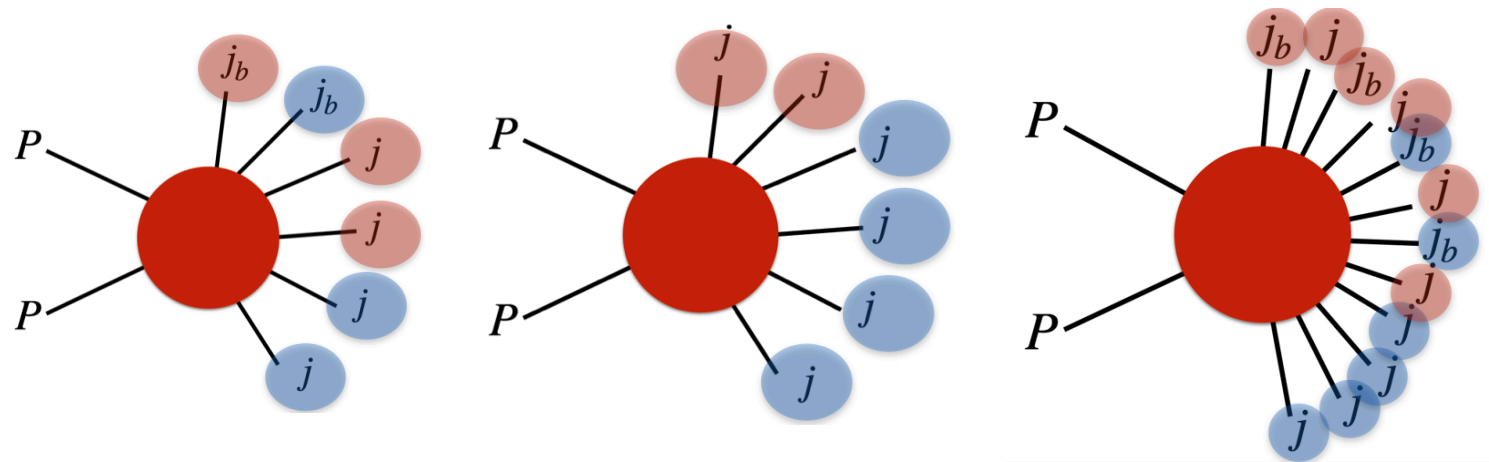
# (small) Quantum advantage

- QA v.s. Brute-force scanning:  
The required time (mostly preparation time  $T_{\text{QUBO}}$ )  
of QA machine:  $T_{\text{QUBO}} = \mathcal{O}(n^2)$   
The complete scanning with  $n$  input takes  $\mathcal{O}(2^n)$



# (big) Quantum advantage

- QA v.s. SA

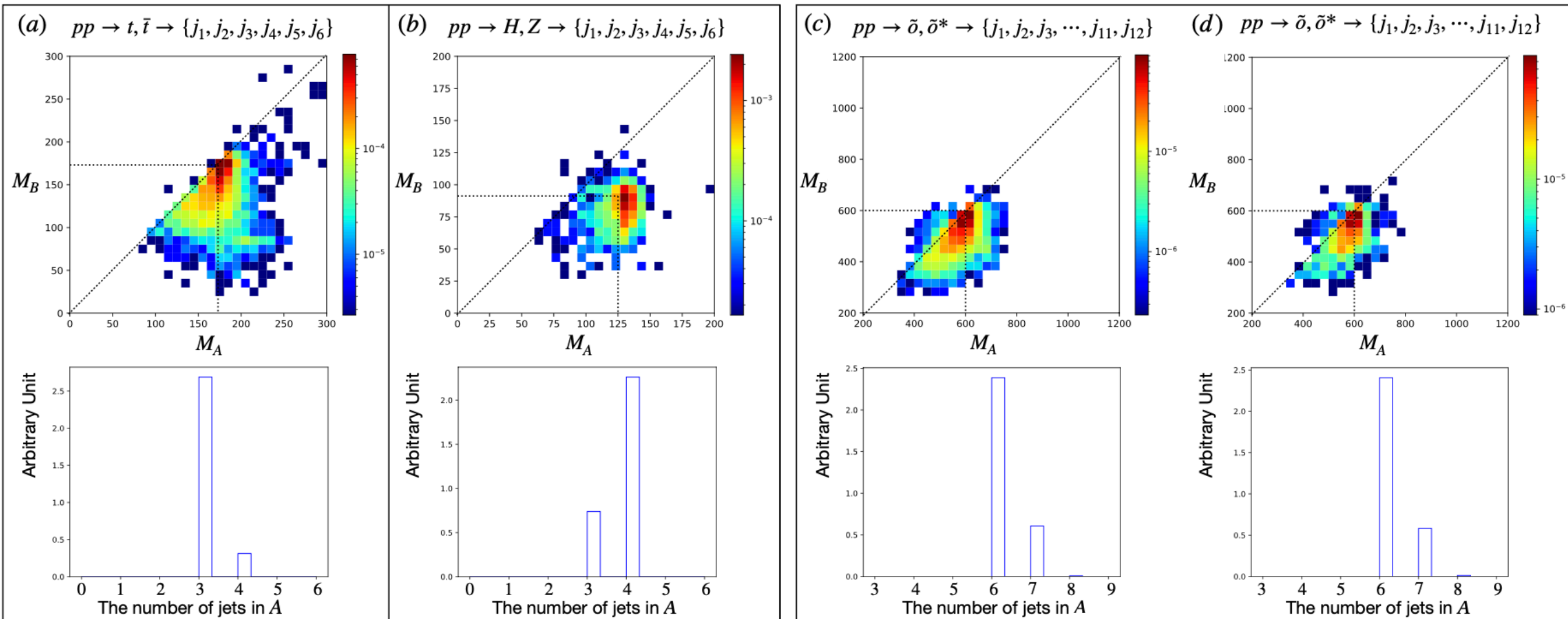


Process	$pp \rightarrow t\bar{t}$ (2 $\rightarrow$ 6 )	$pp \rightarrow HZ$ (2 $\rightarrow$ 6 )	$pp \rightarrow \tilde{o}\tilde{o}^*$ (2 $\rightarrow$ 12 )
Quantum annealing	100%	100%	74.3%
Simulated annealing	36.7%	45.7%	1%

Percentage to get a **global minimum energy state**  
(**does not guarantee** a true combinatorial assignment)



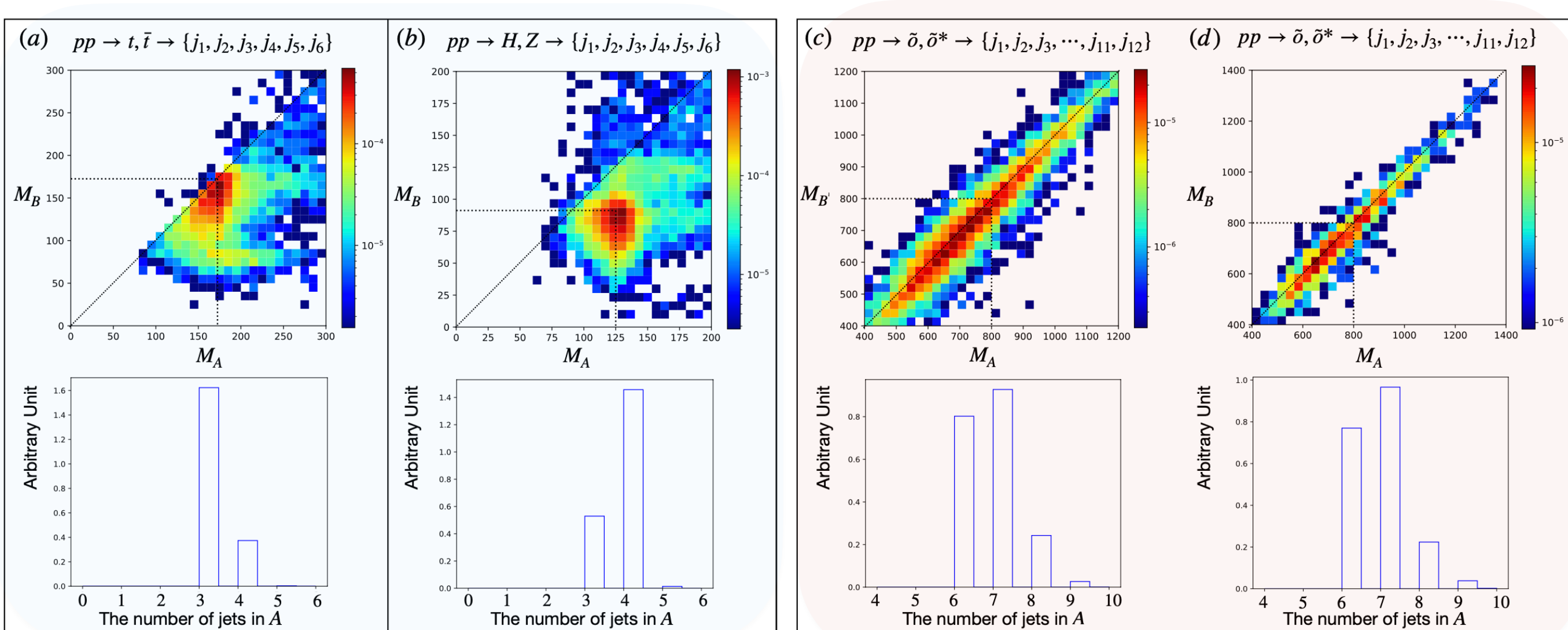
# results



- Madgraph  $\rightarrow$  Pythia (ISR/FSR/MPI turned off)  $\rightarrow$  Delphes

\* a to c: brute force scanning for  $H_{\text{QUBO}}$  to check the fidelity of our algorithm  
d is from D-Wave computer (expensive...)

# results



- Madgraph  $\rightarrow$  Pythia (**ISR/FSR/MPI turned ON**)  $\rightarrow$  Delphes

(As we give a priority to hardest jets,  
**effect of hard ISR is emerging** for hard scale,  
 here  $2m_{\tilde{o}} = 1.2\text{TeV}$ )

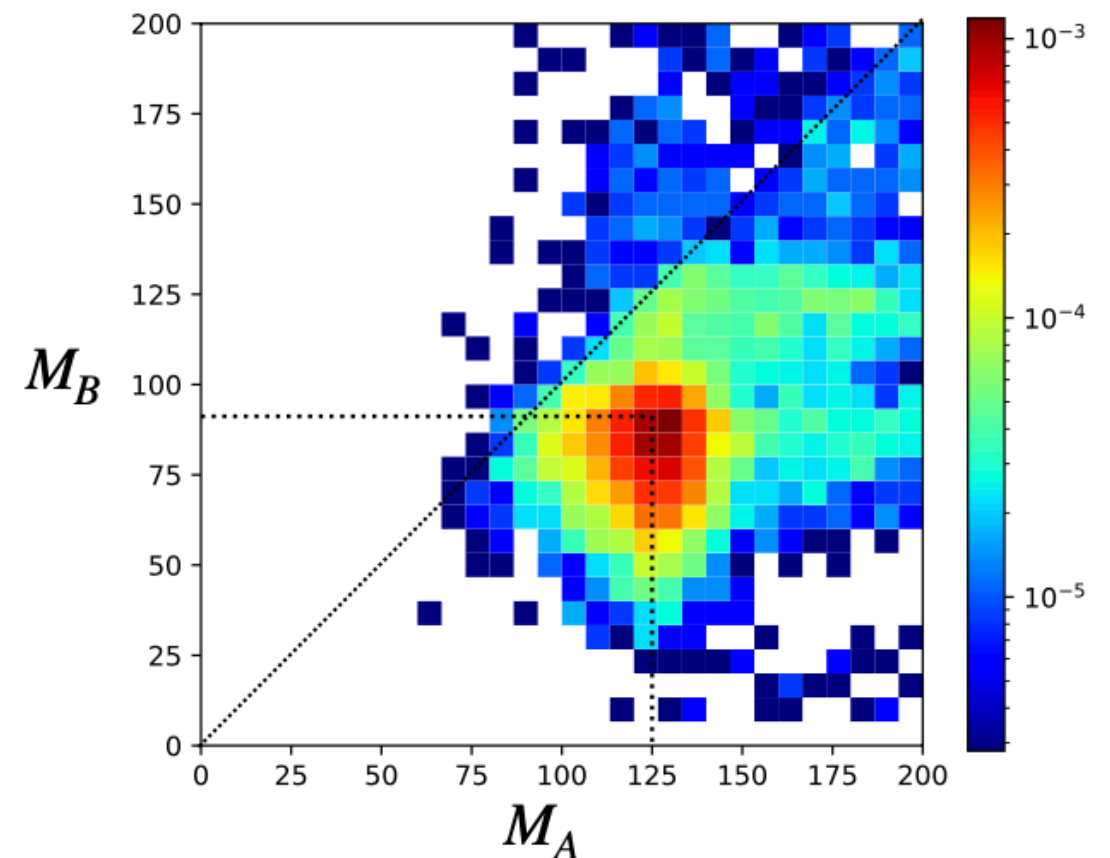
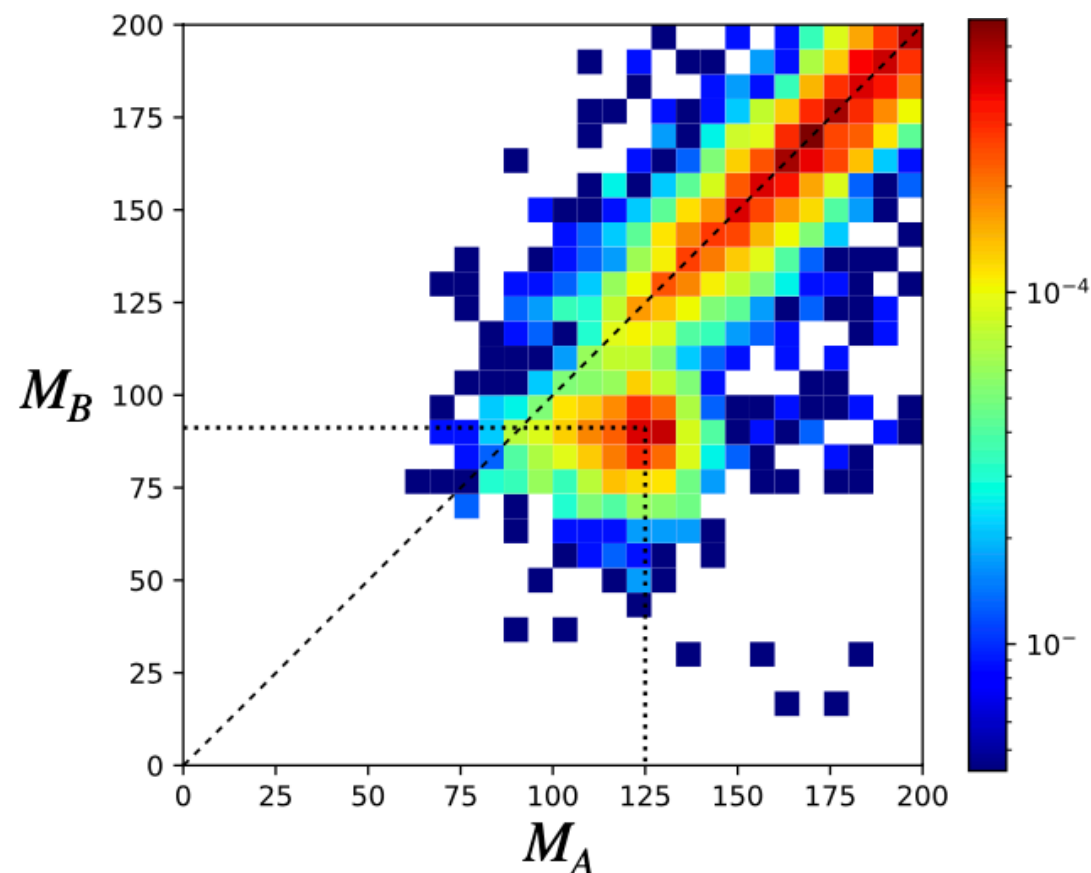
# Effect of additional constraints

$$H = (P_1^2 - P_2^2)^2 \rightarrow H + \lambda (P_1^2 + P_2^2)$$

- For different mother particle cases:  $pp \rightarrow HZ$

$$H = (P_1^2 - P_2^2)^2$$

$$H \rightarrow H + \lambda (P_1^2 + P_2^2)$$



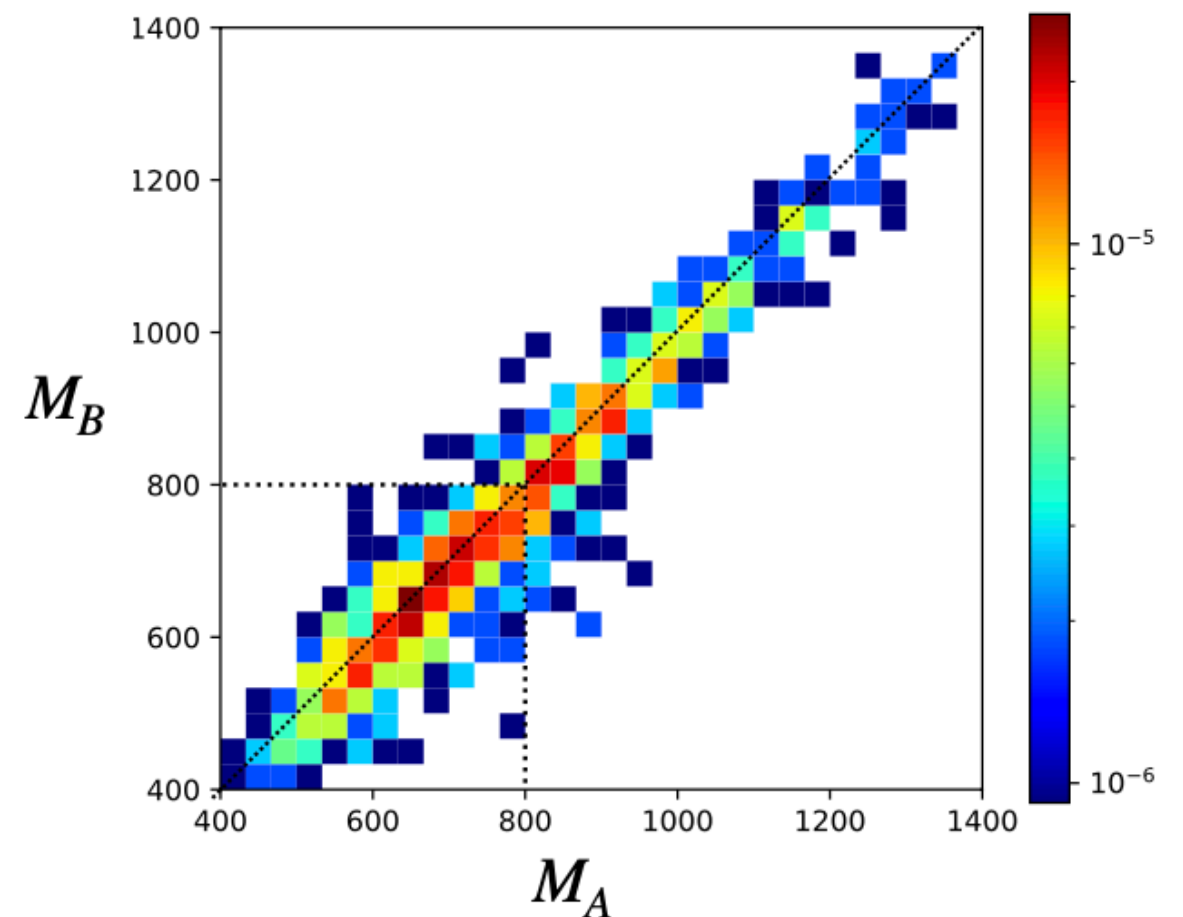
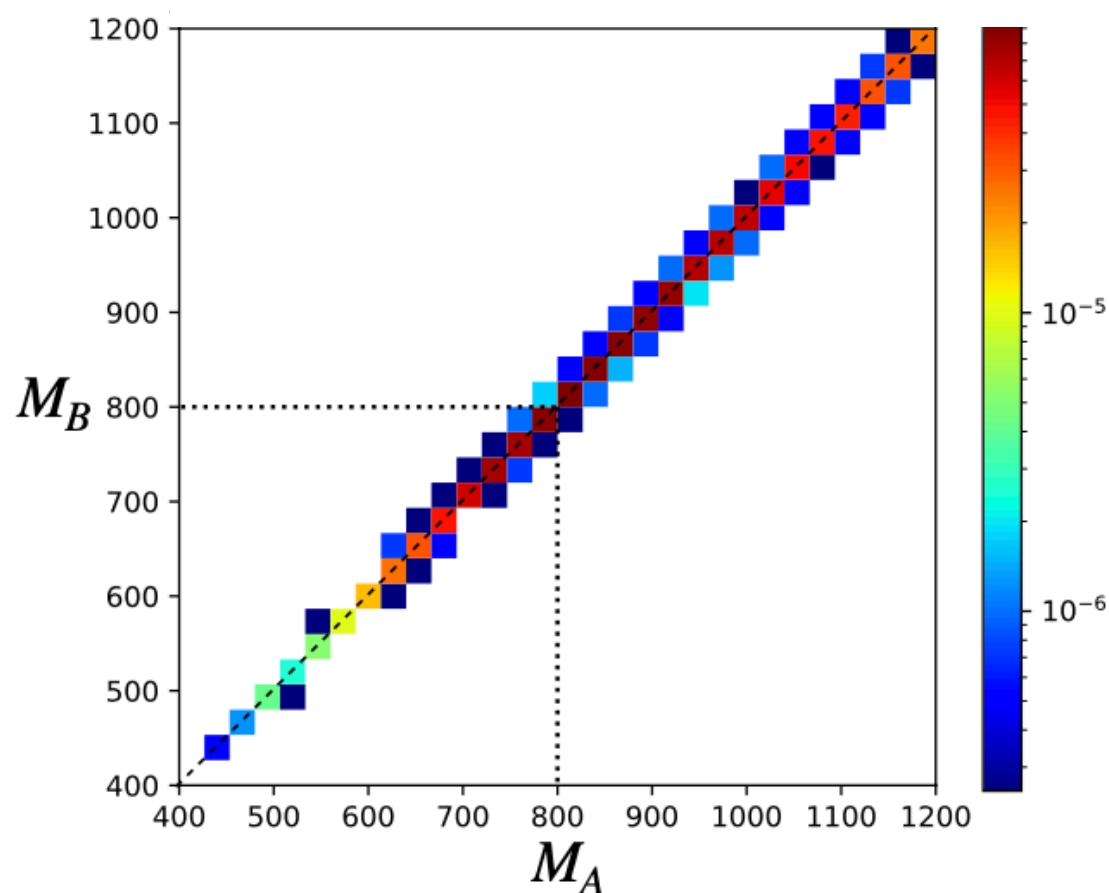
# Effect of additional constraints

$$H = (P_1^2 - P_2^2)^2 \rightarrow H + \lambda (P_1^2 + P_2^2)$$

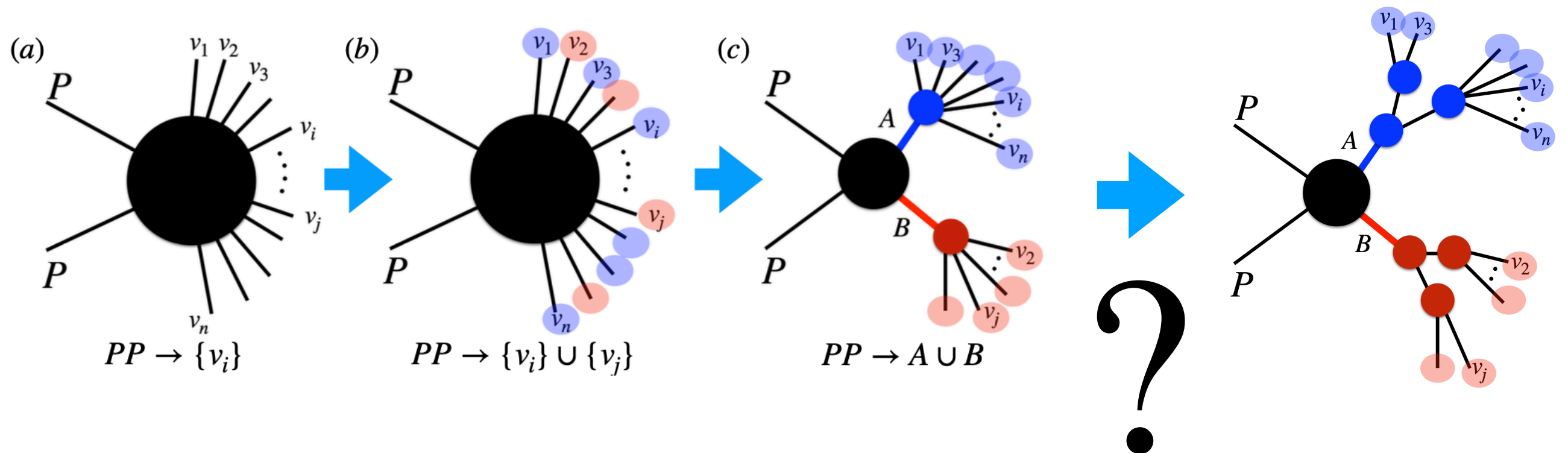
- For smearing effects :  $pp \rightarrow \tilde{o}\tilde{o} \rightarrow t\bar{t}t\bar{t}$

$$H = (P_1^2 - P_2^2)^2$$

$$H \rightarrow H + \lambda (P_1^2 + P_2^2)$$



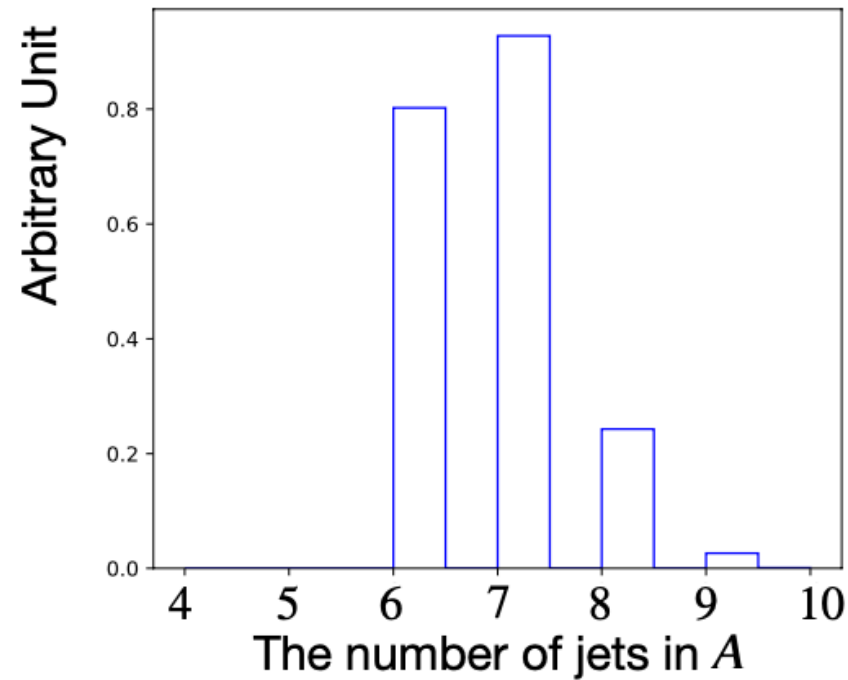
# Sequential algorithm



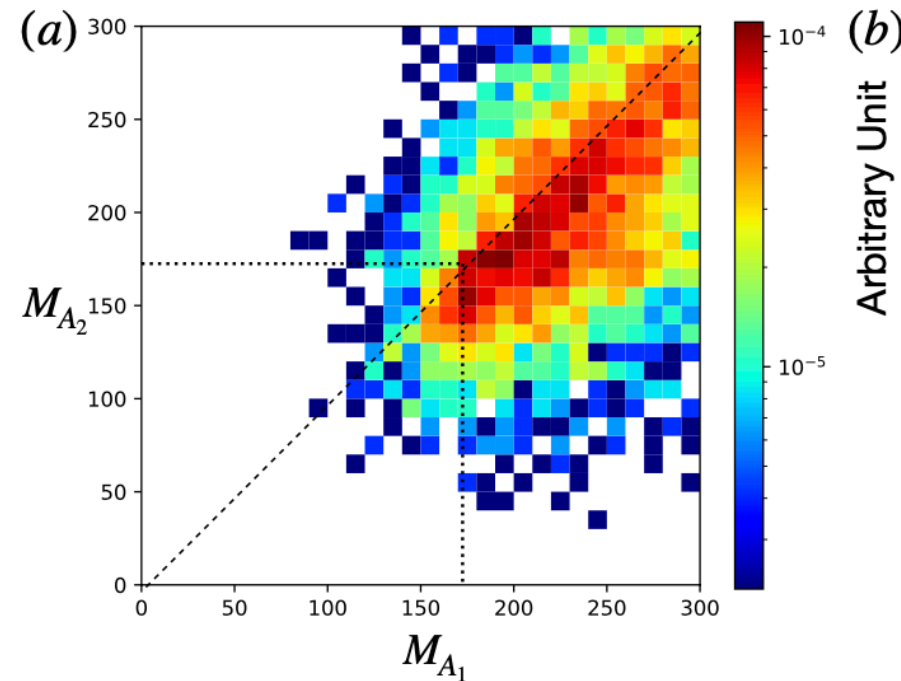
$$H_{\text{QUBO}}^{(A)} = \sum_{ij=1}^{\ell} J'_{ij}{}^{\alpha} s_i^{\alpha} s_j^{\alpha} + \sum_{i=1}^{\ell} h'_i{}^{\alpha} s_i^{\alpha},$$

$$H_{\text{QUBO}}^{(B)} = \sum_{ij=1}^m J'_{ij}{}^{\beta} s_i^{\beta} s_j^{\beta} + \sum_{i=1}^m h'_i{}^{\beta} s_i^{\beta},$$

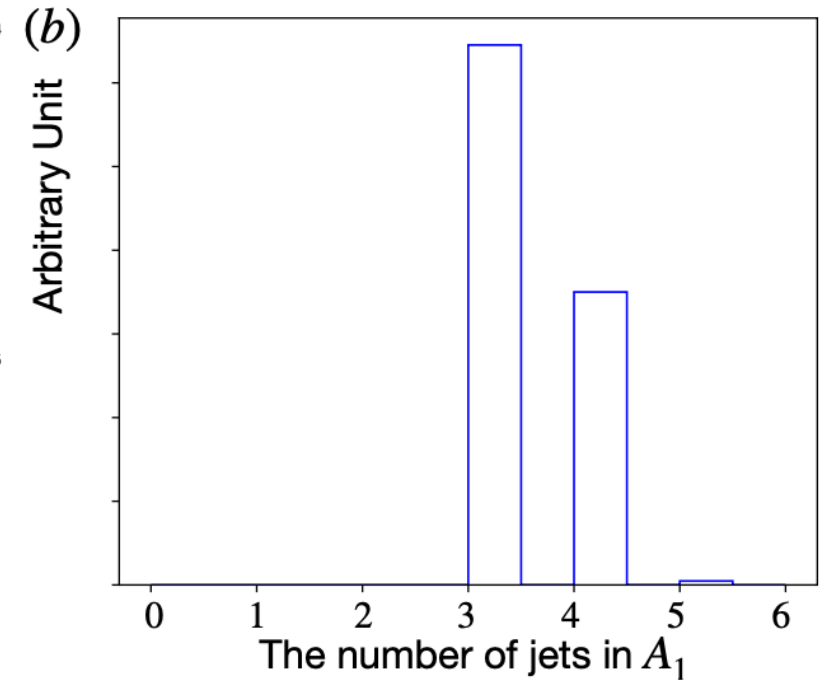
- For 12 hard-jets production, it would be worthy if we can check whether this is four-tops events or not !



$$pp \rightarrow A, B$$



$$A \rightarrow A_1, A_2$$



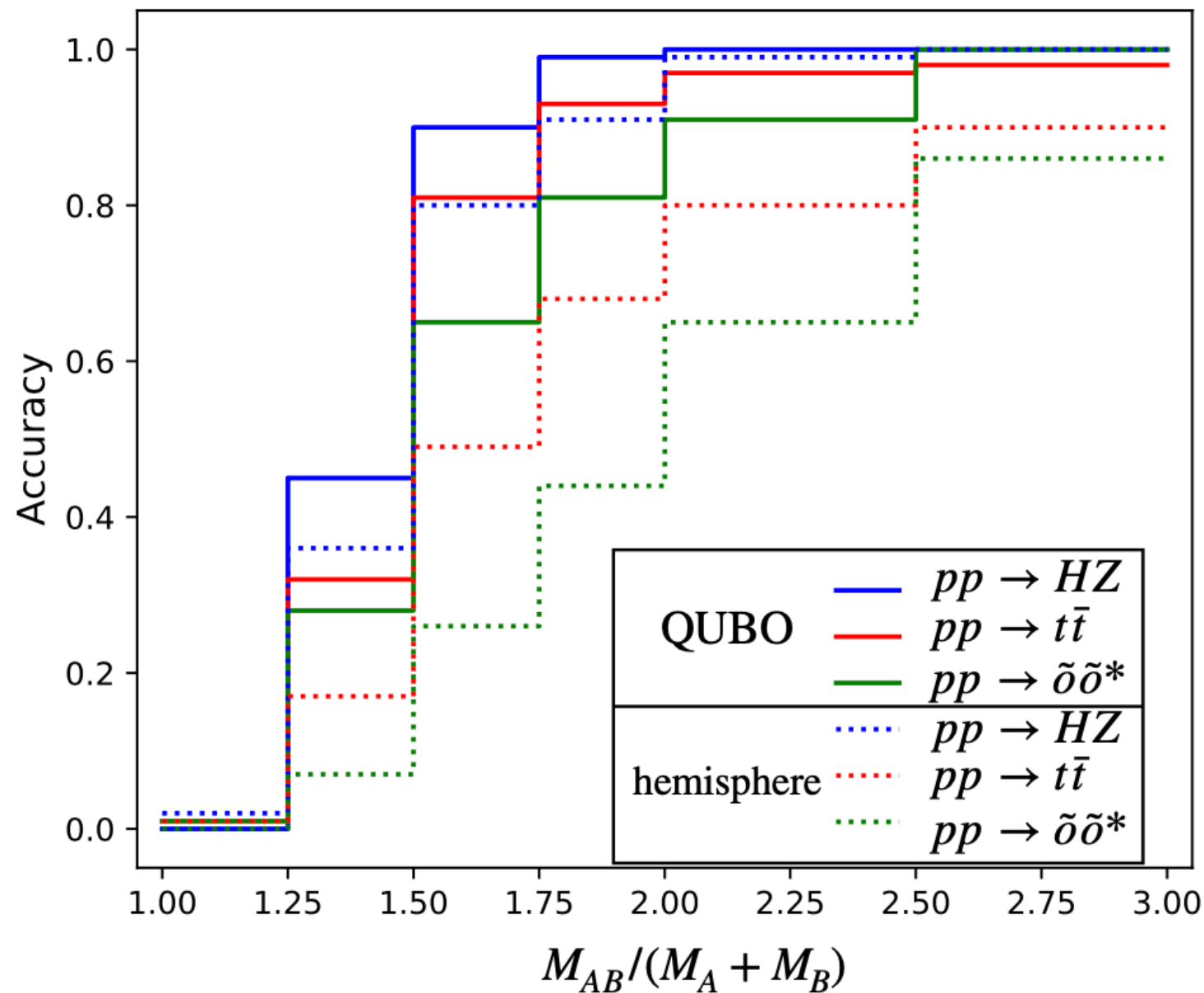
- We can "guess" that  $A_i = t(\bar{t})$  as their **mass** and **number of children** are identical to the case of a top-quark.

# Bench mark?

- There are not many studies on identifying event-topology. (as far as I have searched... if I missed, plz let me know)
- Hemisphere method: **seed-based** algorithm  
(**our algorithm is seedless** one)

Process		$pp \rightarrow t\bar{t}$ Eq. (7a)	$pp \rightarrow HZ$ Eq. (7b)	$pp \rightarrow \tilde{o}\tilde{o}^*$ Eq. (7c)
Algorithm	QUBO	47.3%	89.5%	15.1%
	Hemisphere	33.6%	86.2%	5.84%

(Parton-level analysis with detector cuts)

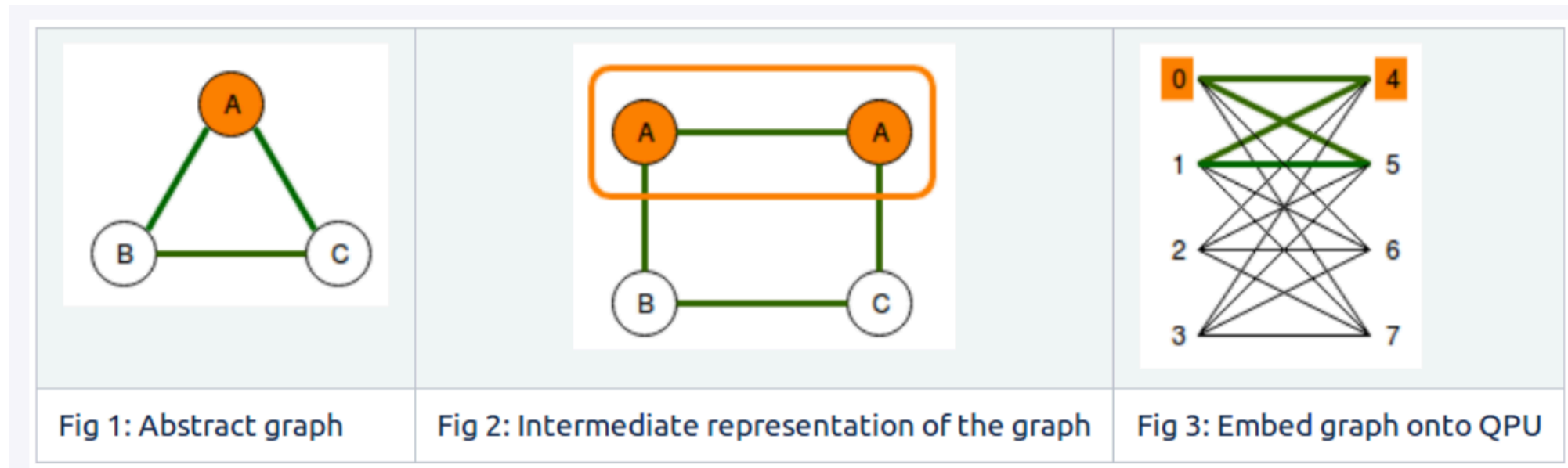


- Performance of an algorithm based on "**seed**" becomes weak when particles are **not boosted enough to develop structures**.
- Lorentz boost factor  $\gamma_A = \frac{E_A}{M_A} = \frac{M_{AB}}{2M_A}$  (for A=B case)

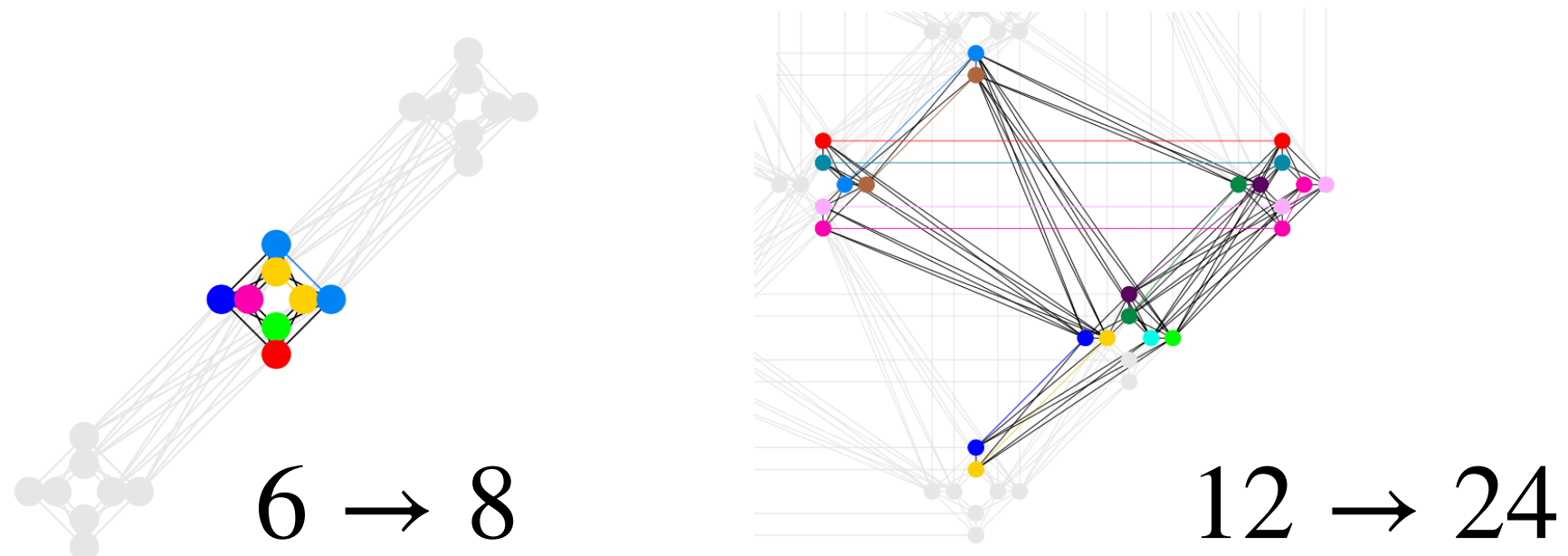


# Current limits for QA

- Number of couplers is limited
  - **spin-chain** method to encode a hamiltonian (connections)

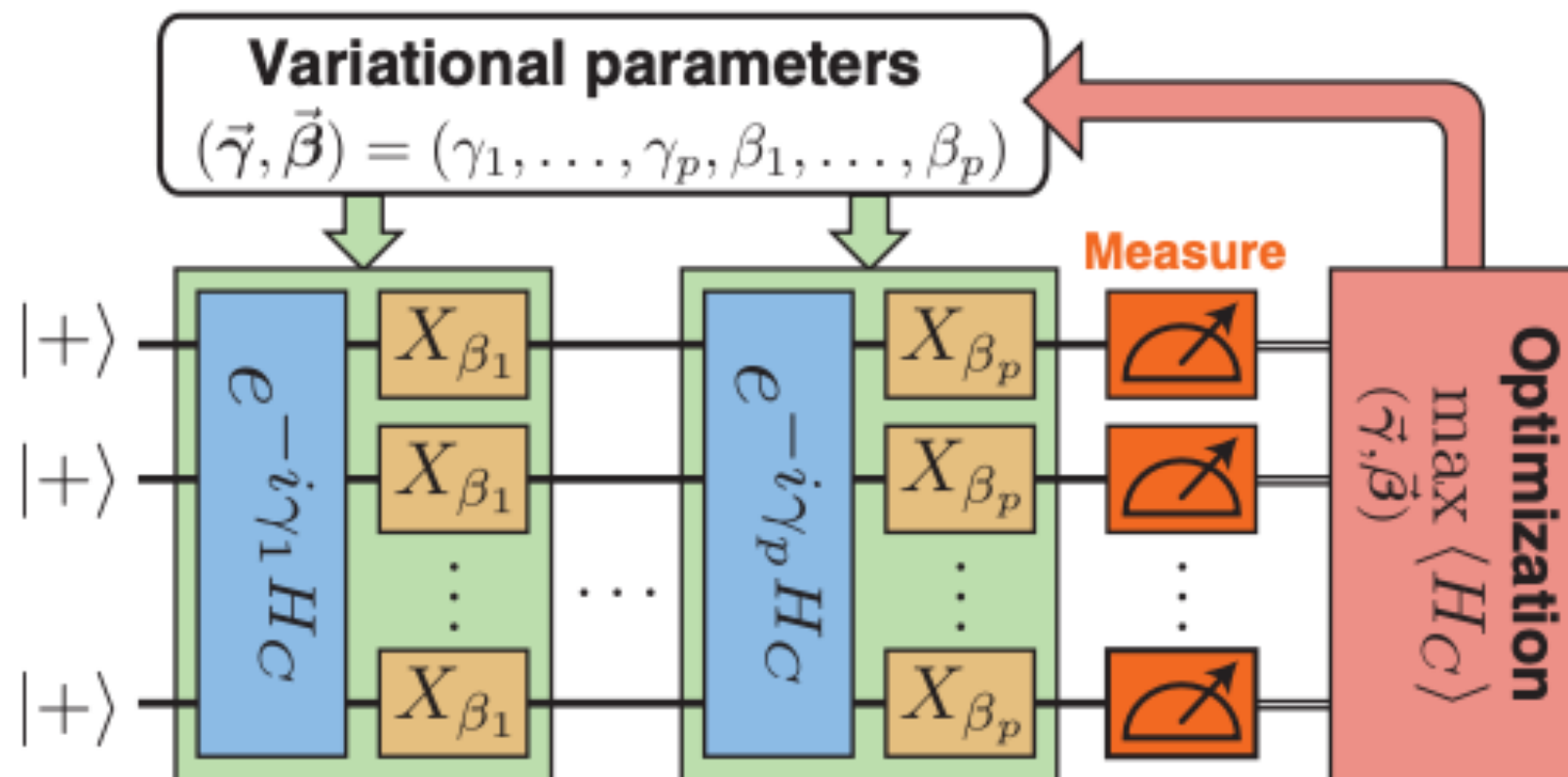


- Number of required qubits for our problem

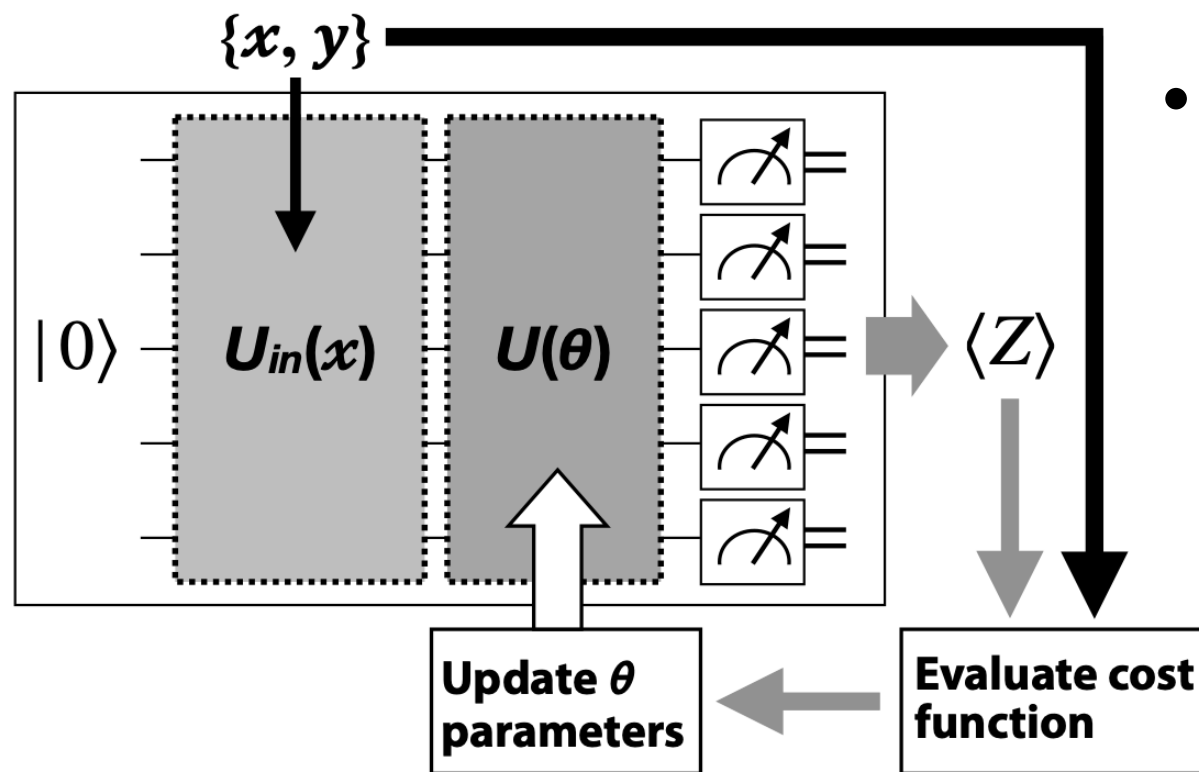


# Conclusion@QC

- I presented a **simple quantum annealing method for clustering reconstructed particles**.
- Gate-based QC can be used via a variational algorithm.
- I am very interested in this new possibility (Now ongoing)



- Hybrid Quantum Classifier: Variational Quantum Approaches

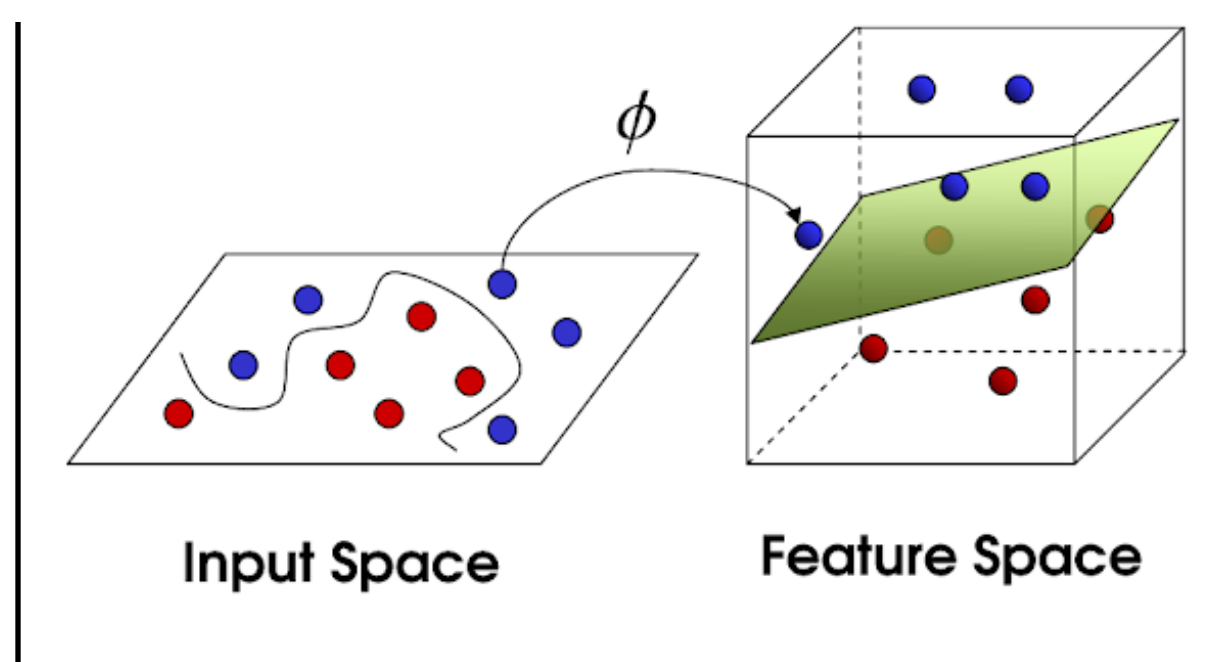


- Supervised Machine Learning with a label  $y$  and expectation  $\langle Z \rangle$  from QC.

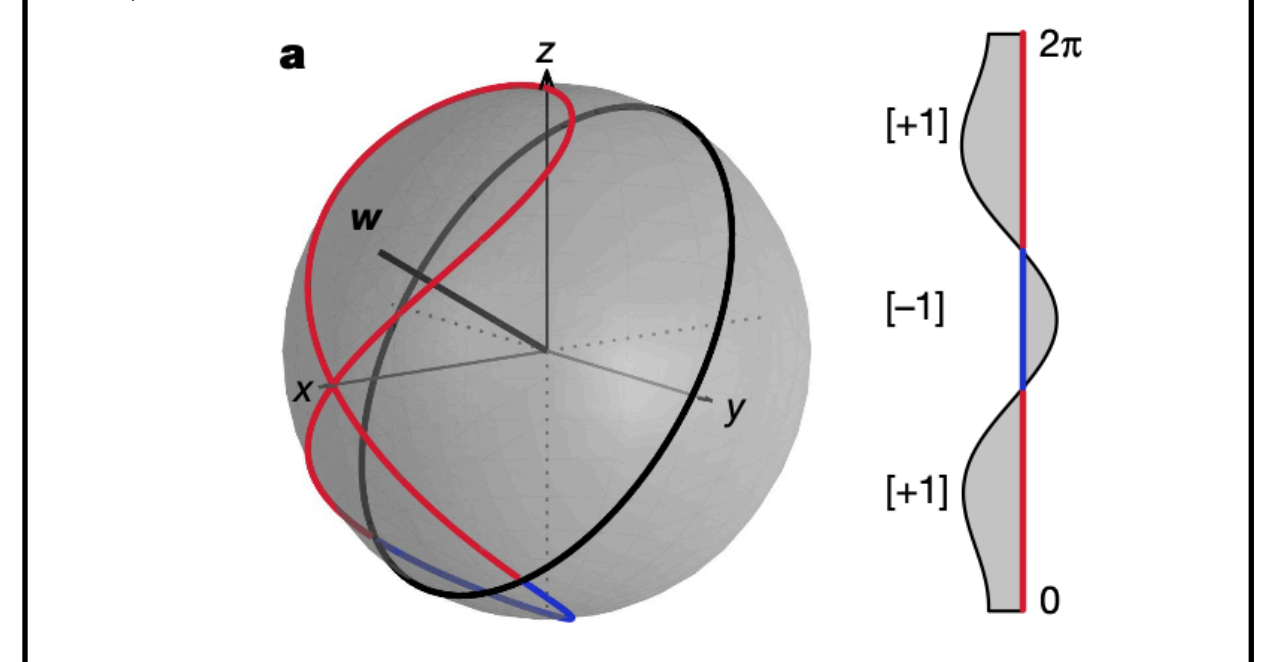
Update learnable parameters  $\theta$  by a classical computer

- Mapping input data **to an exponentially large Quantum Hilbert Space**.

### classical ML Kernel Method



### Quantum kernel function



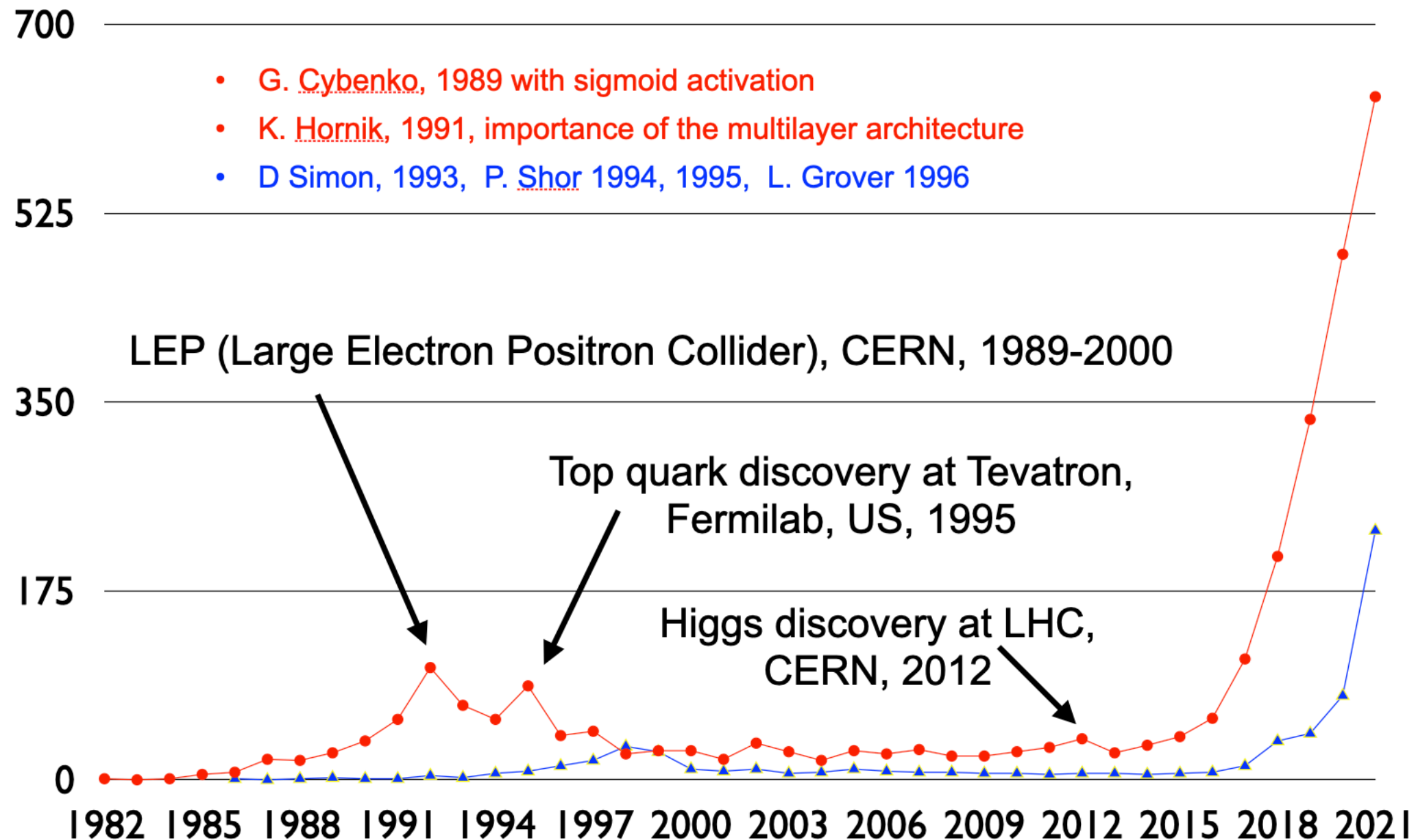
# Conclusion

Credit to KC. Kong

Data is obtained via **InspireHEP**

● The number of papers (in high energy physics) that has a keyword “Machine Learning”, “Deep Learning”, “Artificial Intelligence” or “Neural Networks” in their title.

▲ The number of papers that has a keyword “Quantum Computer”, “Quantum Computing”, “Quantum Annealing” or “Quantum Machine Learning” in their title.



- As a **desperate** seeker, we have tried to take advantages of new computing methods, ML, QC, QML.
- In this talk, I presented a **bottom-up** collider algorithm to identify a new physics from a signal (if we can have)
- There could be many examples to demonstrate **Quantum Advantage** in the field of HEP.
- **Stay tuned...**