

# Applications of Machine Learning to Detecting Fast Neutrino Flavor Instabilities

Focus workshop on collective oscillations and chiral transport of neutrinos

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SFB 1258

Neutrinos  
Dark Matter  
Messengers



# Neutrino Oscillations in Dense Media

- Neutrino evolution in dense neutrino media is **very different** from the one in vacuum and matter

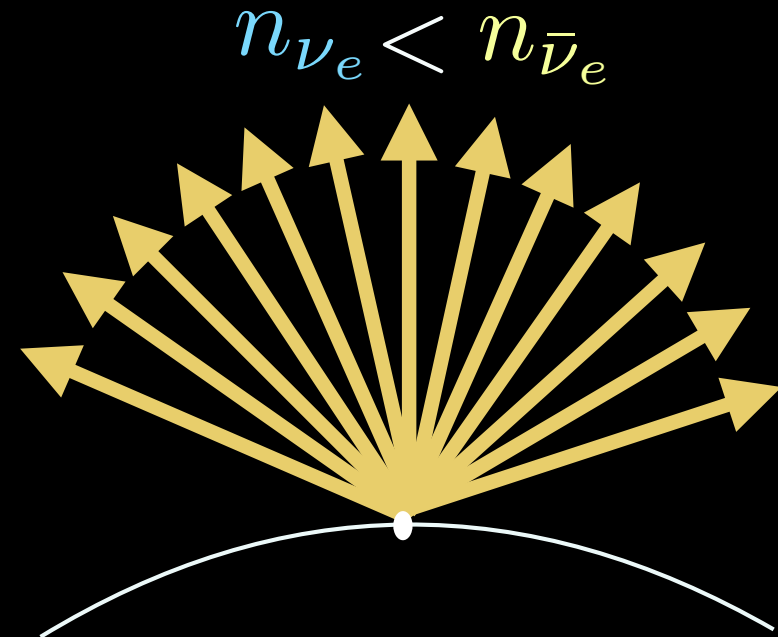
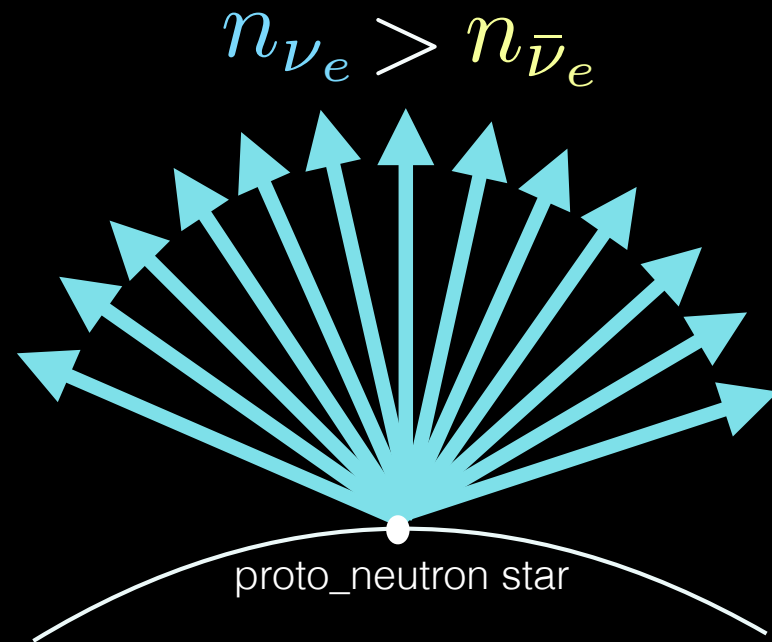
$$i(\partial_t + \mathbf{v} \cdot \nabla)\rho = [H, \rho]$$

$$H = \frac{1}{2} \begin{bmatrix} -\omega \cos 2\theta + \sqrt{2}G_F n_e & \omega \sin 2\theta \\ \omega \sin 2\theta & \omega \cos 2\theta - \sqrt{2}G_F n_e \end{bmatrix} + H_{\nu\nu}$$

$$\sqrt{2}G_F \int \underbrace{d^3q}_{\text{coupling}} (1 - \mathbf{v}_P \cdot \mathbf{v}_q) \underbrace{(\rho_\nu - \rho_{\bar{\nu}})}_{\text{nonlinearity}}$$

# Fast Flavor Conversions

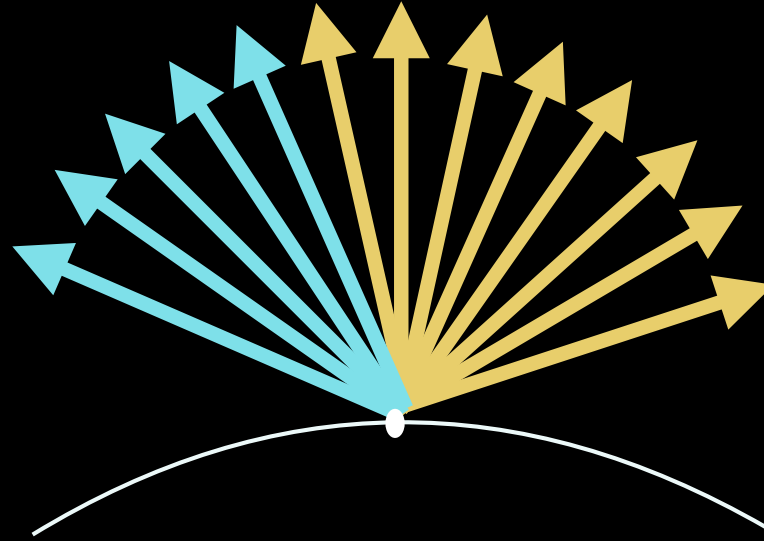
- In our traditional understanding, we assumed that neutrinos are emitted **isotropically** from the surface of the neutrino source
- $f_{\nu_e}(\theta) - f_{\bar{\nu}_e}(\theta)$  is either always **positive or negative**



- This implies that the **scales** on which flavor conversion could occur are determined by **vacuum frequency**  $\Delta m^2 / 2E \sim 1 \text{ km}^{-1}$

# Fast Flavor Conversions

- **FFC** could occur when there is **crossing** in  $f_{\nu_e}(\theta) - f_{\bar{\nu}_e}(\theta)$



- **Scales** on which flavor conversion can occur is now proportional to  $n_\nu$  and could be  $< 10$  cm

- Neutrino oscillations **can** now occur at densities that had been long thought to be the realm of collisional and scattering processes

# Fast Flavor Conversions

- The angular distributions are **not available**, instead we have only access to their moments

$$I_n = \int d \cos \theta_\nu \cos^n \theta_\nu f_\nu(\cos \theta_\nu)$$

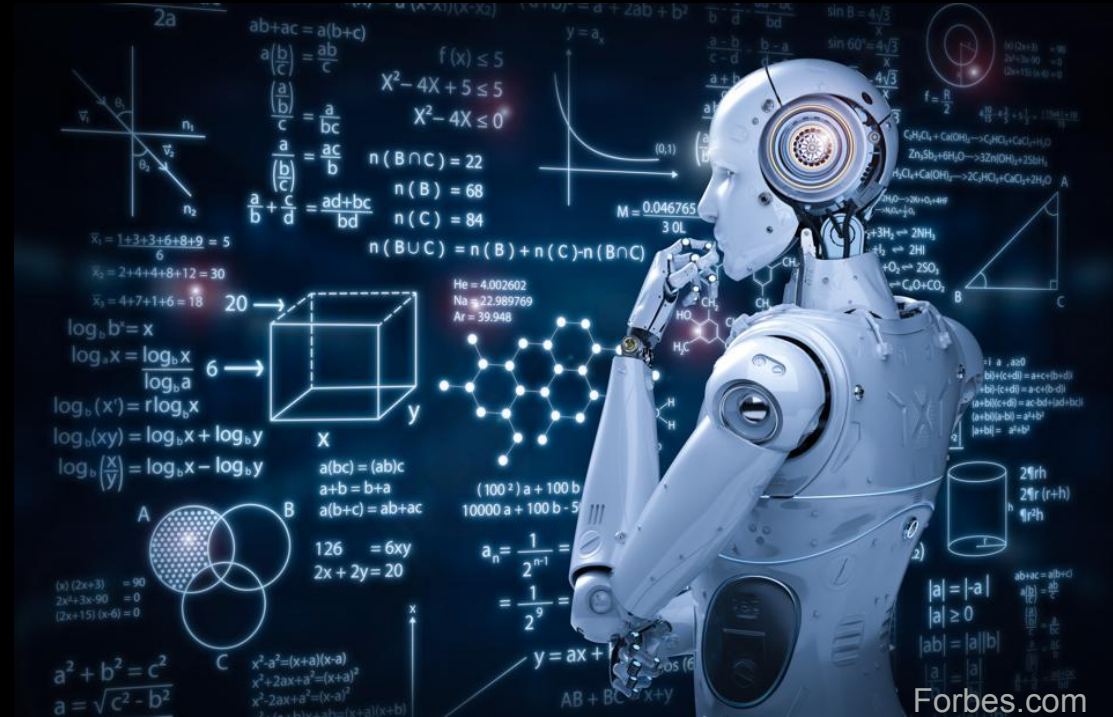
- In M1 closure scheme only the evolution of **zeroth** and **first** moments are followed directly
- We can still make progress! Dasgupta+2018; Abbar2020; Johns+2021; Richers2022;
- But these methods are normally **inefficient** and **very slow**
- FFC can not be detected **on the fly**

# Application of Machine Learning

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- **Machine learning** can help us
- We have **four** feature here:  $I_0$  and  $I_1$  for neutrinos and antineutrinos



# Application of Machine Learning

- For training, we use analytical **maximum-entropy** and **gaussian** distributions

$$f_{\nu}(\cos \theta_{\nu}) = \exp(-\eta + a \cos \theta_{\nu})$$

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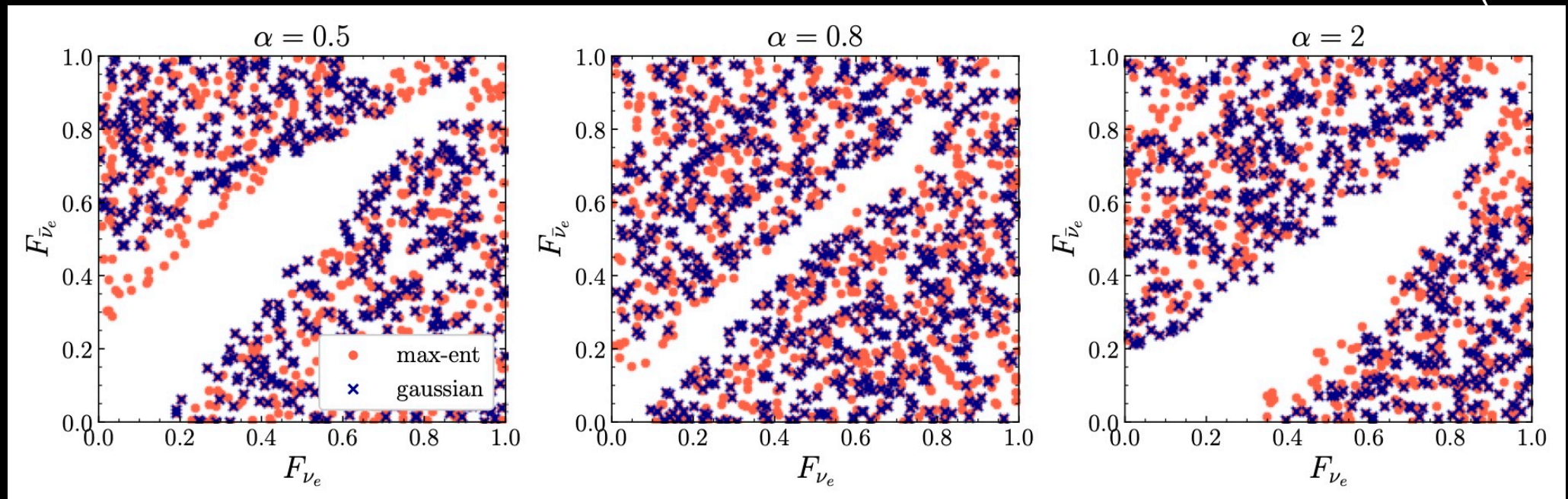
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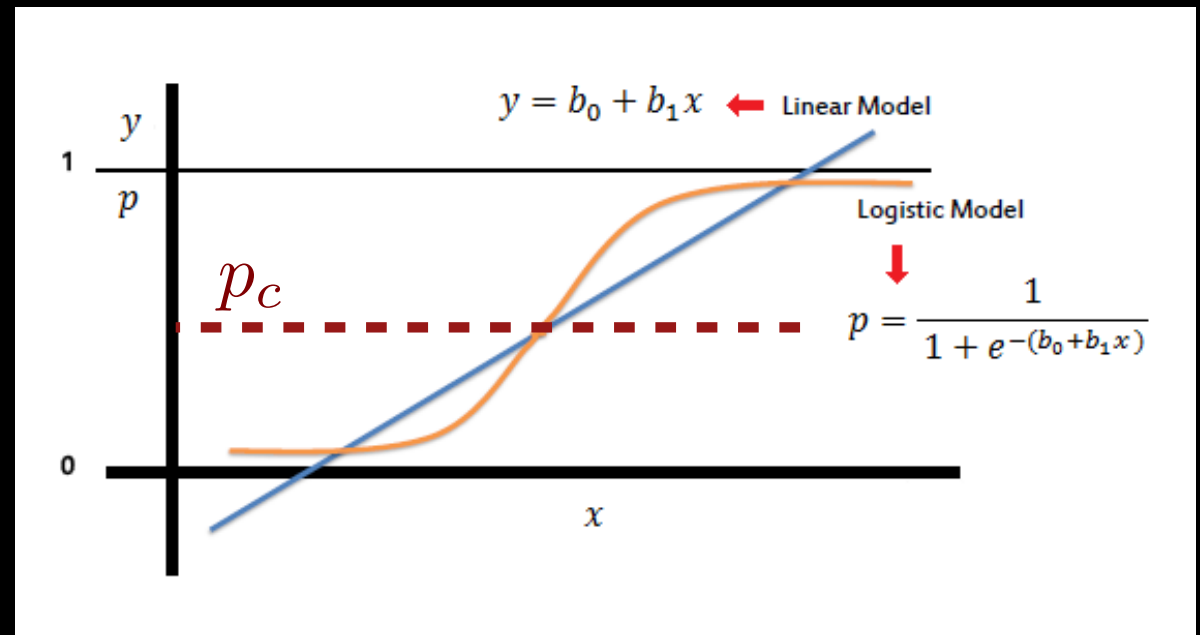
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Abbar (2023)



# Logistic Regression

- Based on finding a line that separates the data points, in which a **logistic** function is applied on the top of the linear one so that one can decide on the basis of some final values which are in (0,1)

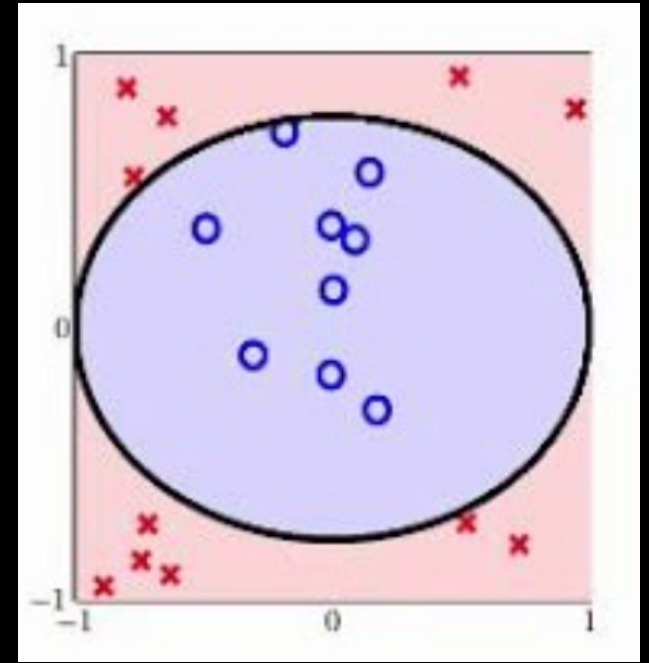


<http://www.elusives.eu>

# Logistic Regression

- one should first make **non-linear transformations**

$$x, y \rightarrow x, y, x^2, y^2$$



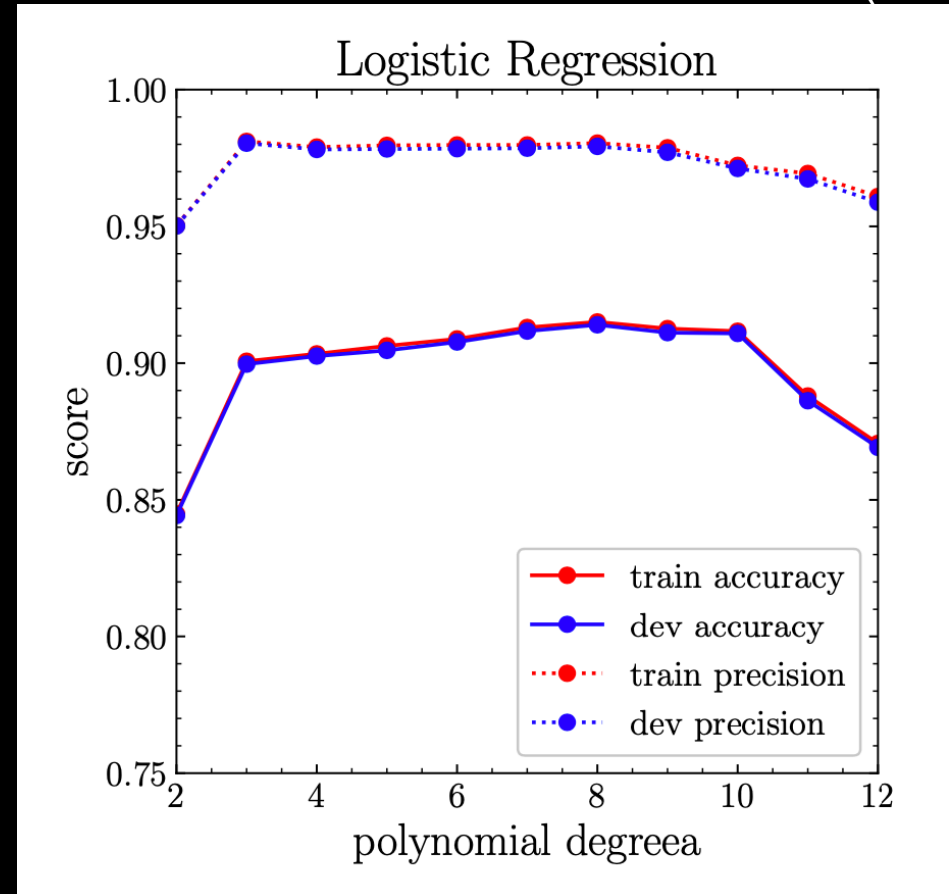
Abu Mostafa

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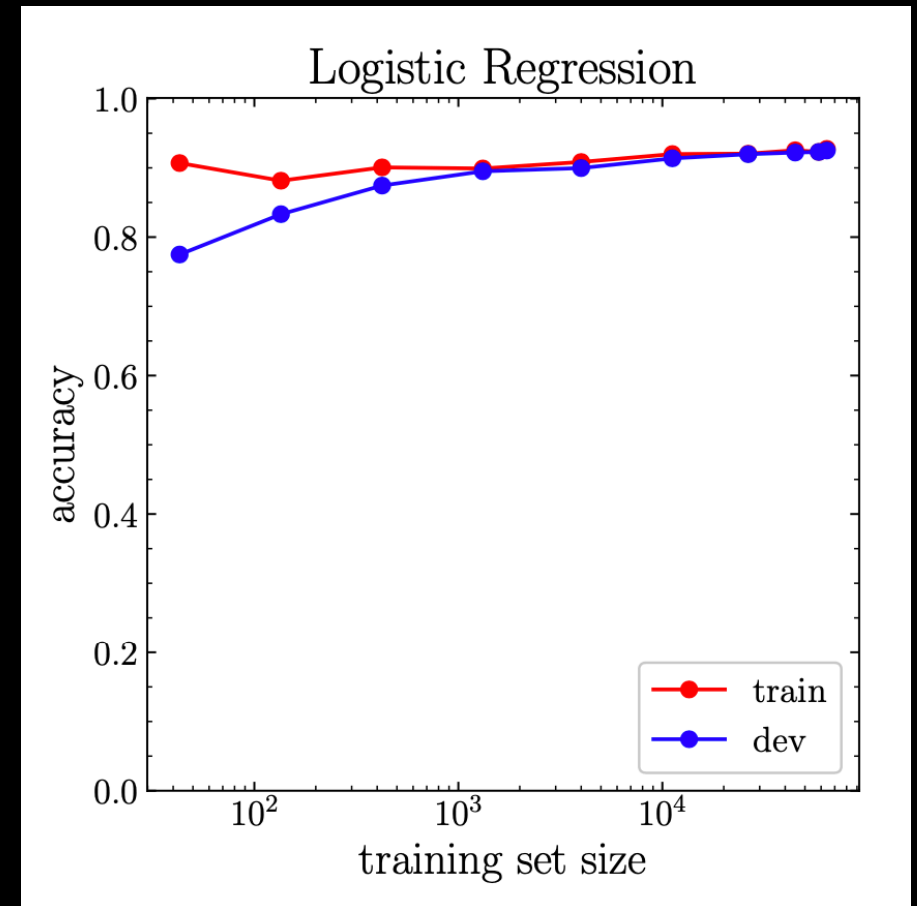
$$x, y \rightarrow x, y, x^2, y^2$$

- In our problem, maximum accuracy is reached for  $n = 9$  Abbar (2023)



# Logistic Regression

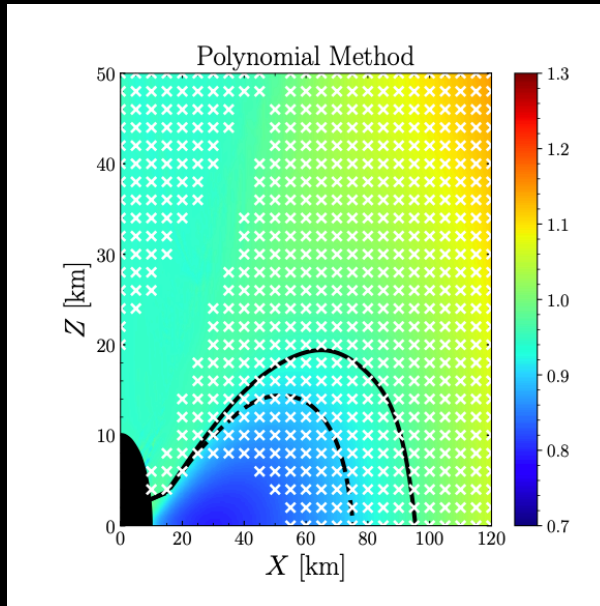
- In order to overcome the **overfitting**, one needs at least a few thousand points



Abbar (2023)

# Logistic Regression

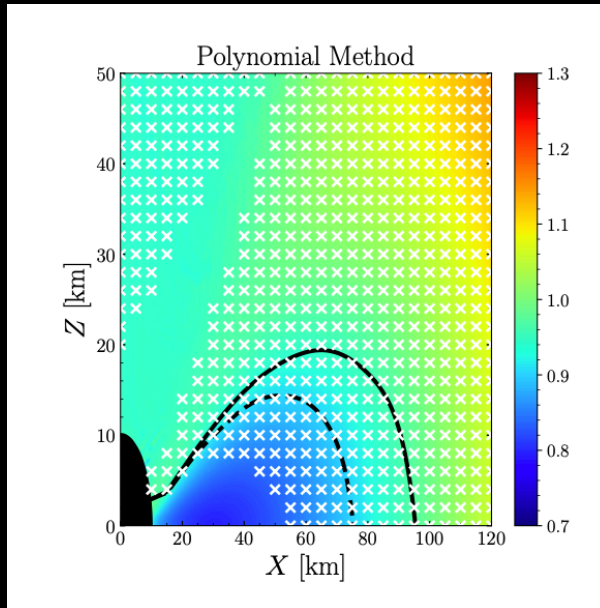
- LR performs well on NSM remnant simulation data



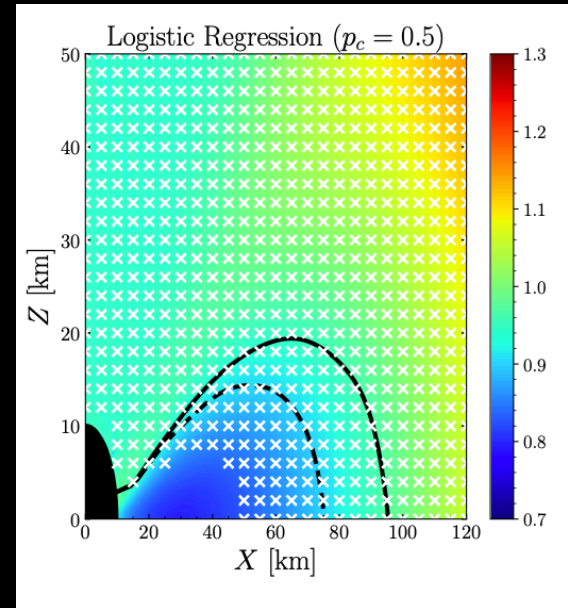
Just+2022

# Logistic Regression

- LR performs **well** on NSM remnant simulation data



Just+ (2022)



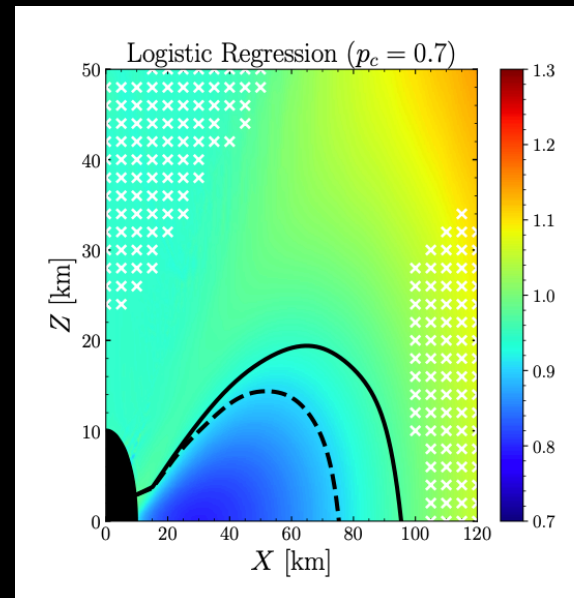
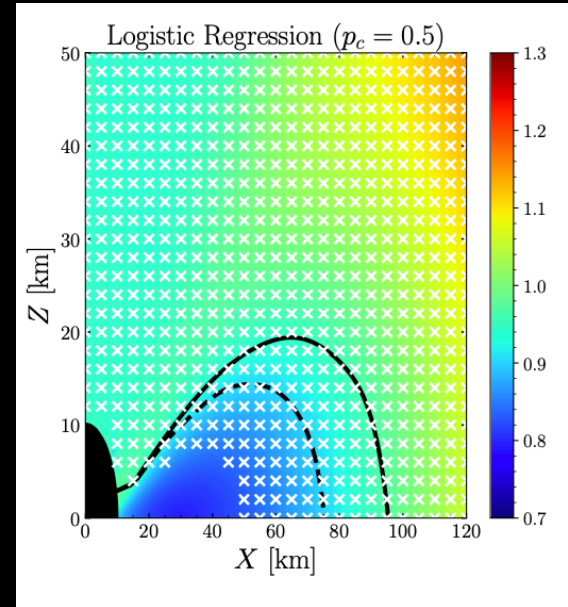
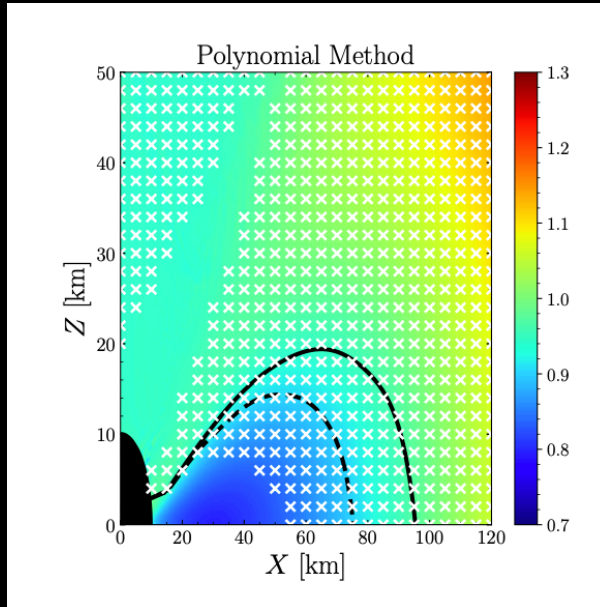
Abbar (2023)

- Another **justification** for using parametric angular distributions



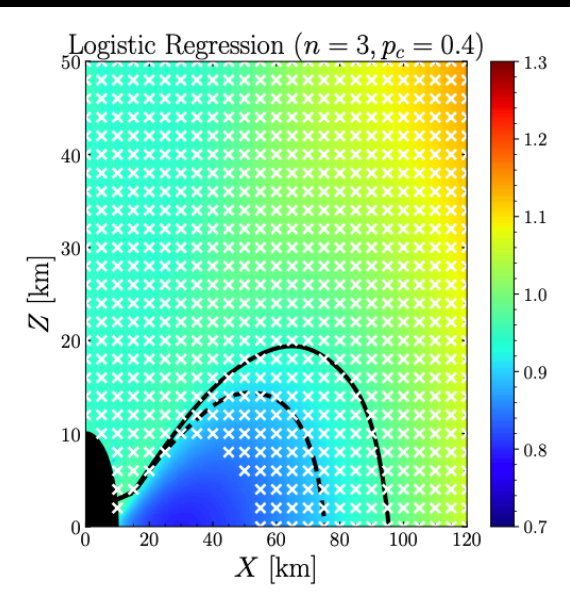
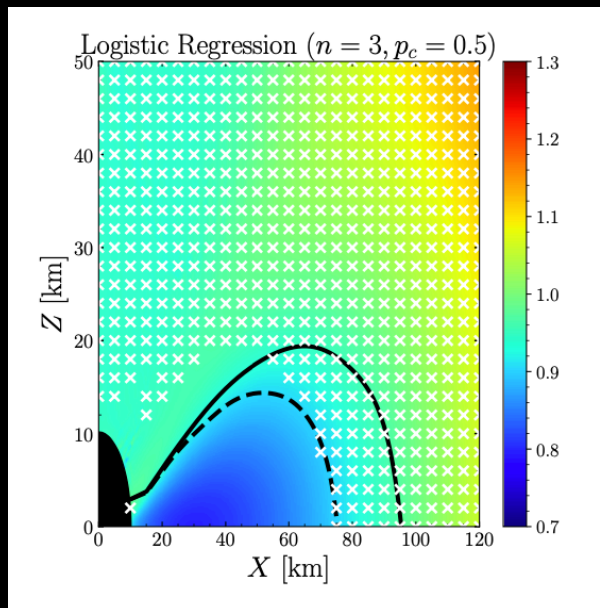
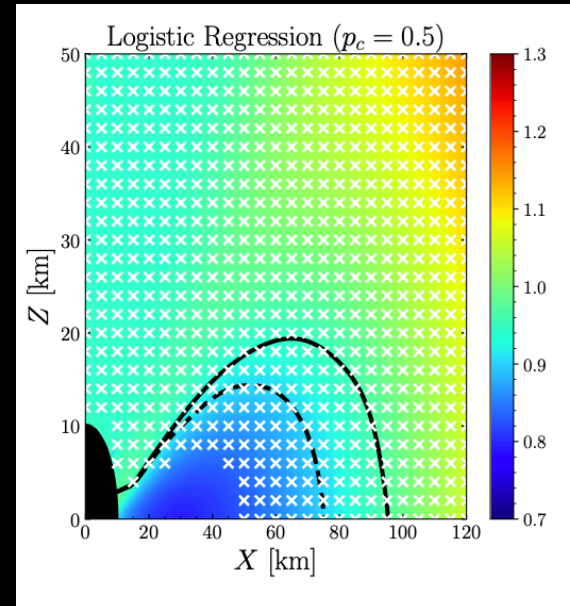
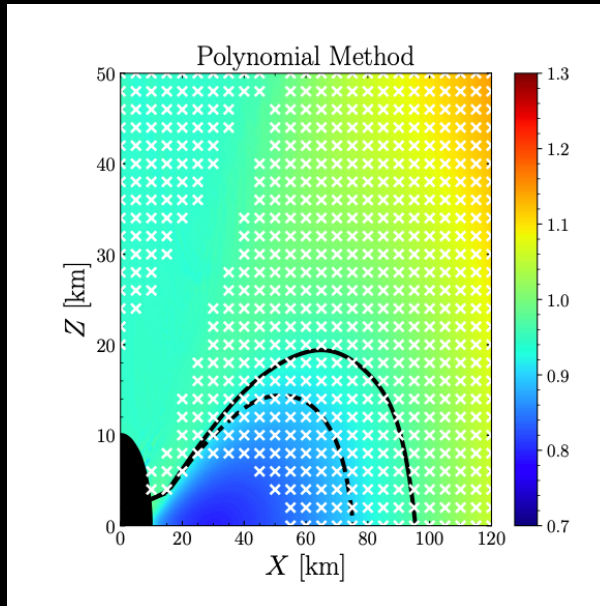
# Logistic Regression

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# Logistic Regression

- Calculations with  $n = 3$  also perform **relatively** well



# Application of Machine Learning

- All our ML algorithms can reach good accuracies

<b>Logistic Regression (93%)</b>			
	precision	recall	$F_1$ -score
no crossing	83%	93%	88%
crossing	97%	93%	95%

<b>KNN (n=3) (95%)</b>			
	precision	recall	$F_1$ -score
no crossing	90%	90%	90%
crossing	96%	96%	96%

<b>SVM (95%)</b>			
	precision	recall	$F_1$ -score
no crossing	92%	90%	91%
crossing	96%	97%	97%

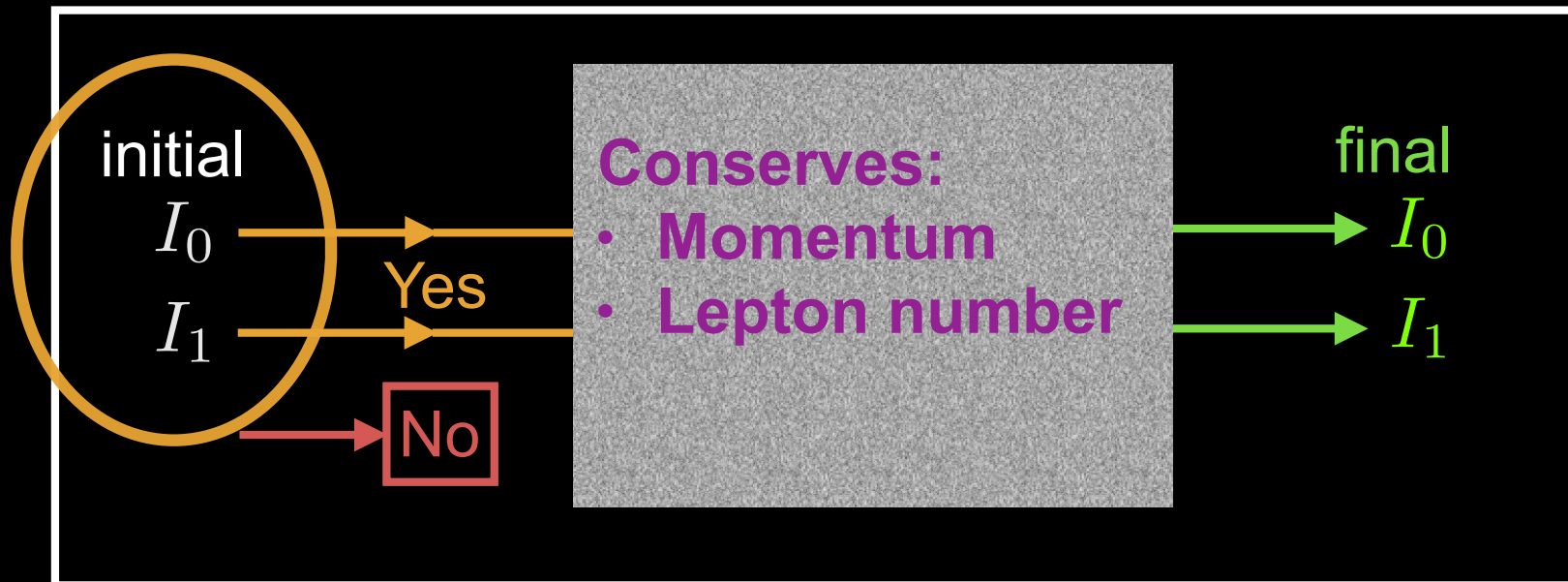
  

<b>Decision tree (94%)</b>			
	precision	recall	$F_1$ -score
no crossing	89%	88%	89%
crossing	96%	96%	96%

Abbar (2023)

# Future Directions

- Improve the ML algorithms with more realistic data
- Three-flavor effects: **muon** creation
- Machine learning methods prove to be very promising regarding the detection of FFI

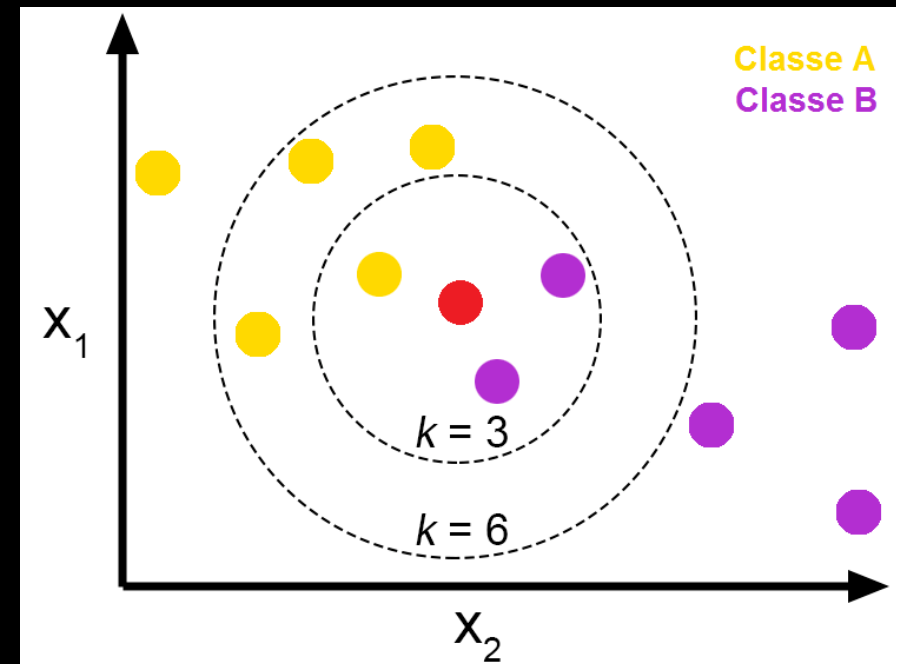


# Summary

- Machine learning is fascinating!

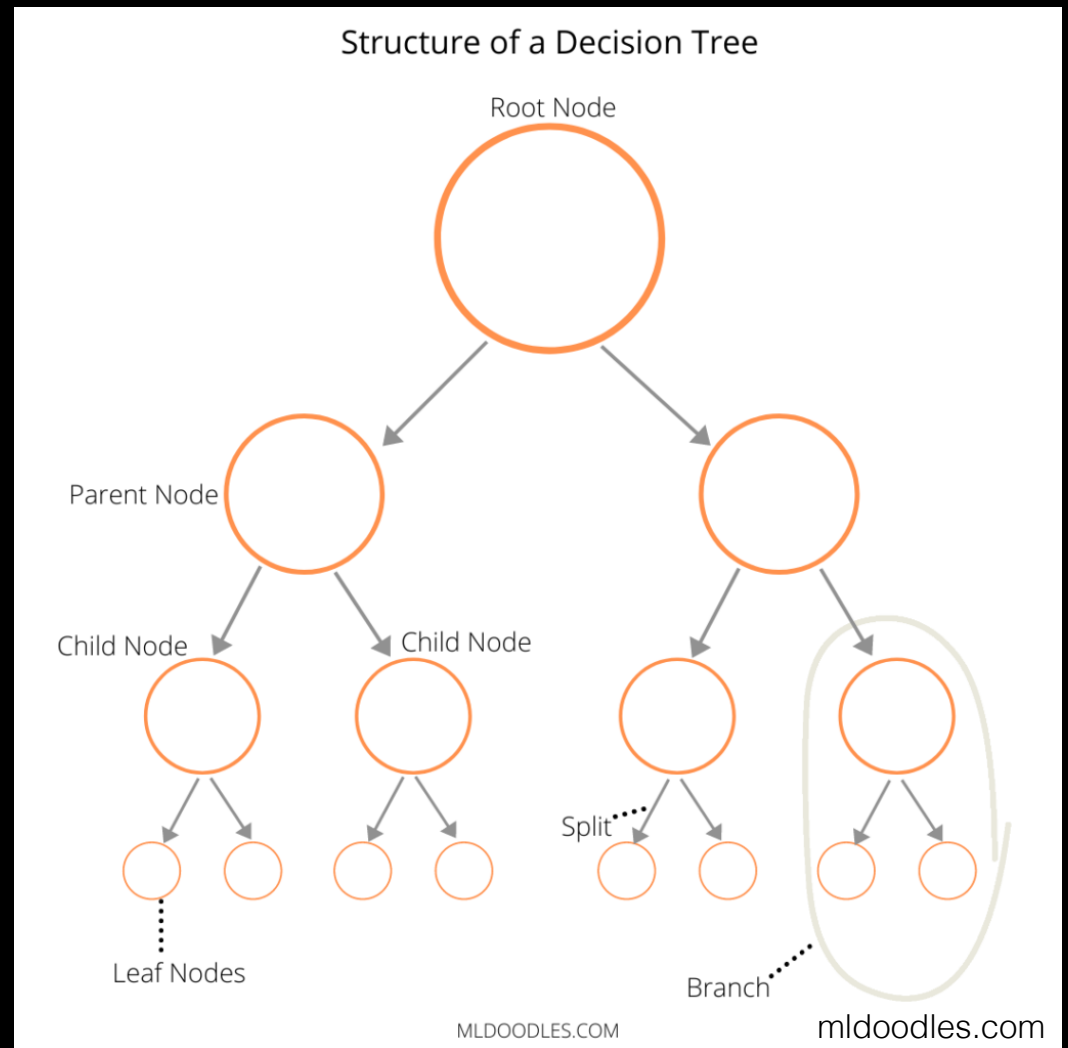
# KNN

- KNN is one of the simplest forms of machine learning algorithms mostly used for classification. It **classifies the data point on how its neighbor is classified.**



# Decision Tree

- In decision tree, one makes decision using a tree-like structure. At each node, one of the features is selected and the branching occurs.



# SVM

- **Support Vector Machine** is a classification based on finding a line that classifies the data points, maximises the margins

