

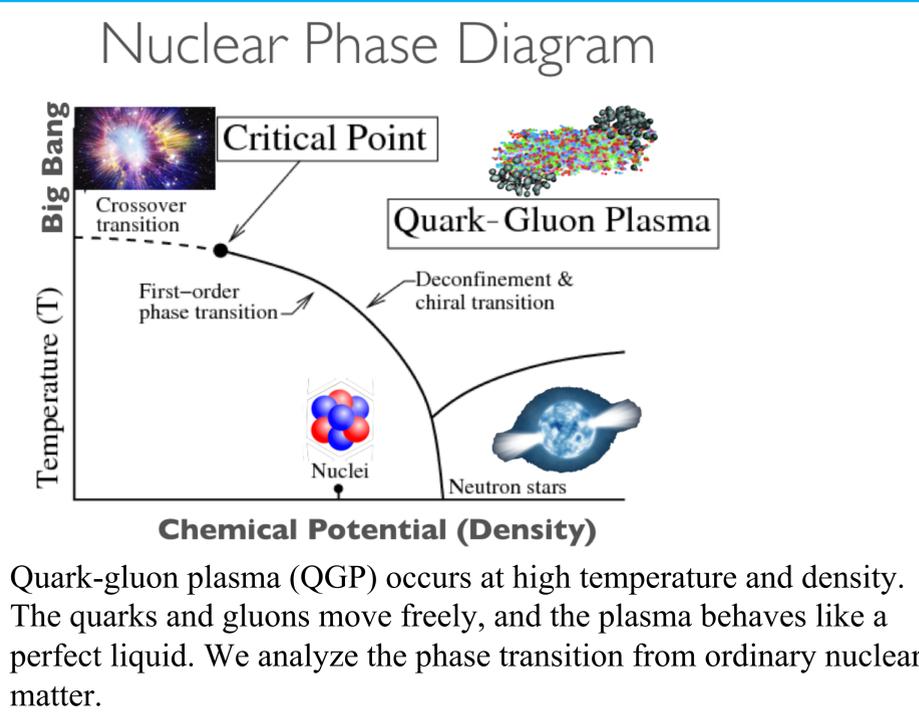
# Machine learning for classifying the chiral phase transition in AdS/QCD

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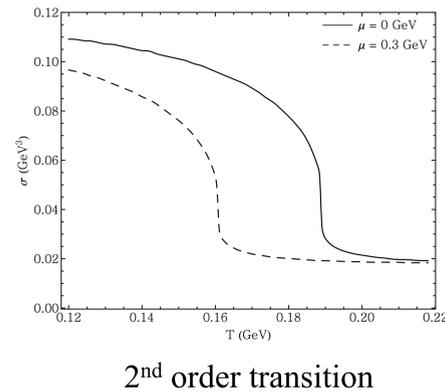
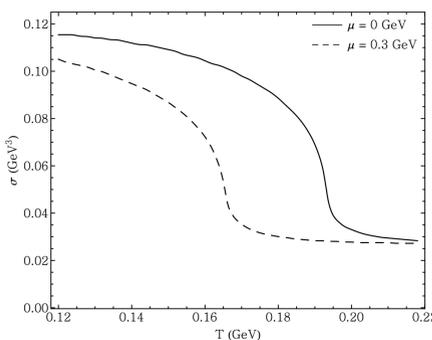
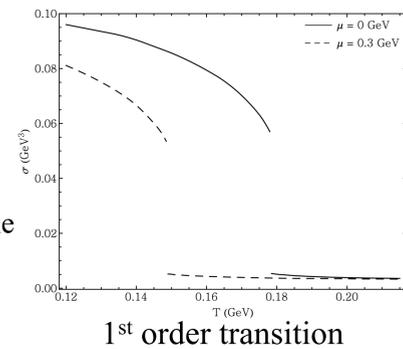
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## Introduction – Physics Background



## Transition Types

- Chiral parameter  $\sigma$  measures difference in behavior between particles with left-handed spin and particles with right-handed spin
- $\sigma = 0$  indicates plasma formation
- We are interested in the *order* of the phase transition

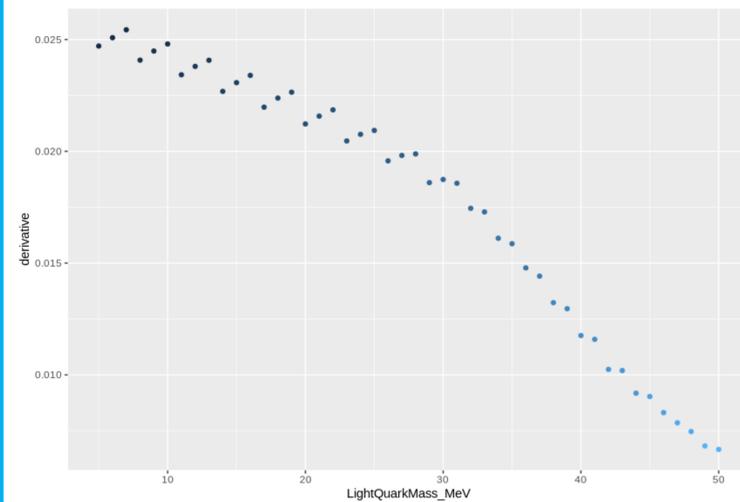
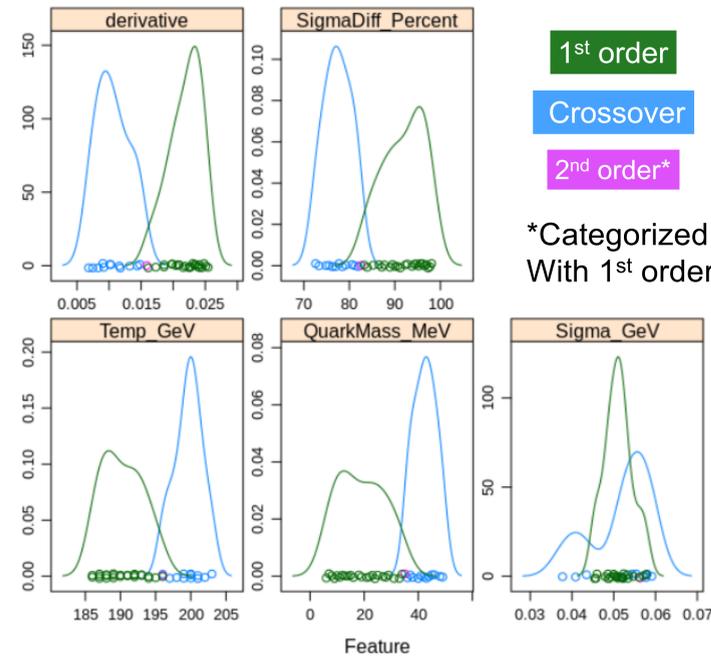


## Machine Learning Model

Types of input data for machine learning

- Maximum derivative of chiral parameter  $\sigma$**
- Percent change of chiral parameter**
- Temperature at maximum derivative
- $\sigma$  value at maximum derivative**
- Quark mass

Bolded variables are used in the machine learning method



- Maximum derivative is an indicator of the phase transition type
- 1<sup>st</sup> order at small quark mass
- Crossover above 35 MeV

The machine learning model combines four classification algorithms:

- classification and regression trees (CART)
- k-nearest neighbors (kNN)
- support vector machines (SVM) with a linear kernel
- random forest (RF).

## Training vs. Testing

3 flavor model:  
Light quark mass equals strange quark mass

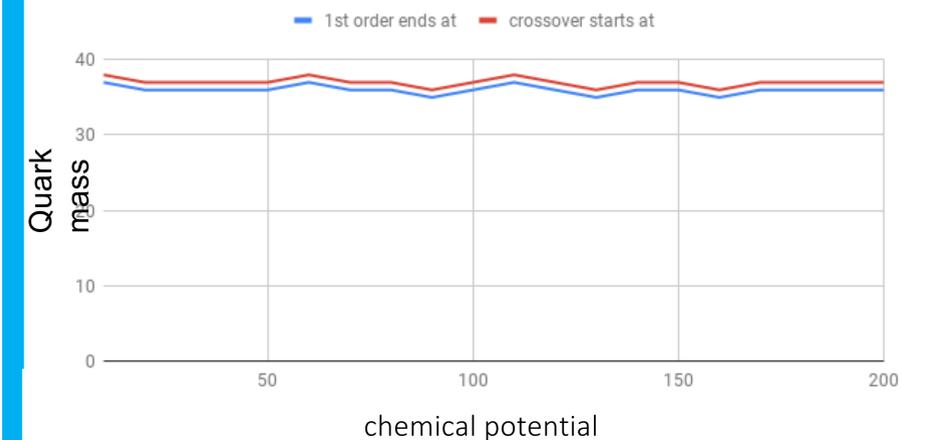
2 +1 flavor model:  
Quark masses can differ

Model	Training Data	Testing Data	Success Rate
3 flavor	80% randomly selected	Remaining 20%	100%
2+1 flavor	Chemical potential = 0	Chemical potential = 300	100%

## Application

We test the machine learning method on a physics model with known behavior

1<sup>st</sup> order ends at and crossover starts at



We will use machine learning to develop a physics model with realistic phase transitions.

## Acknowledgements

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