

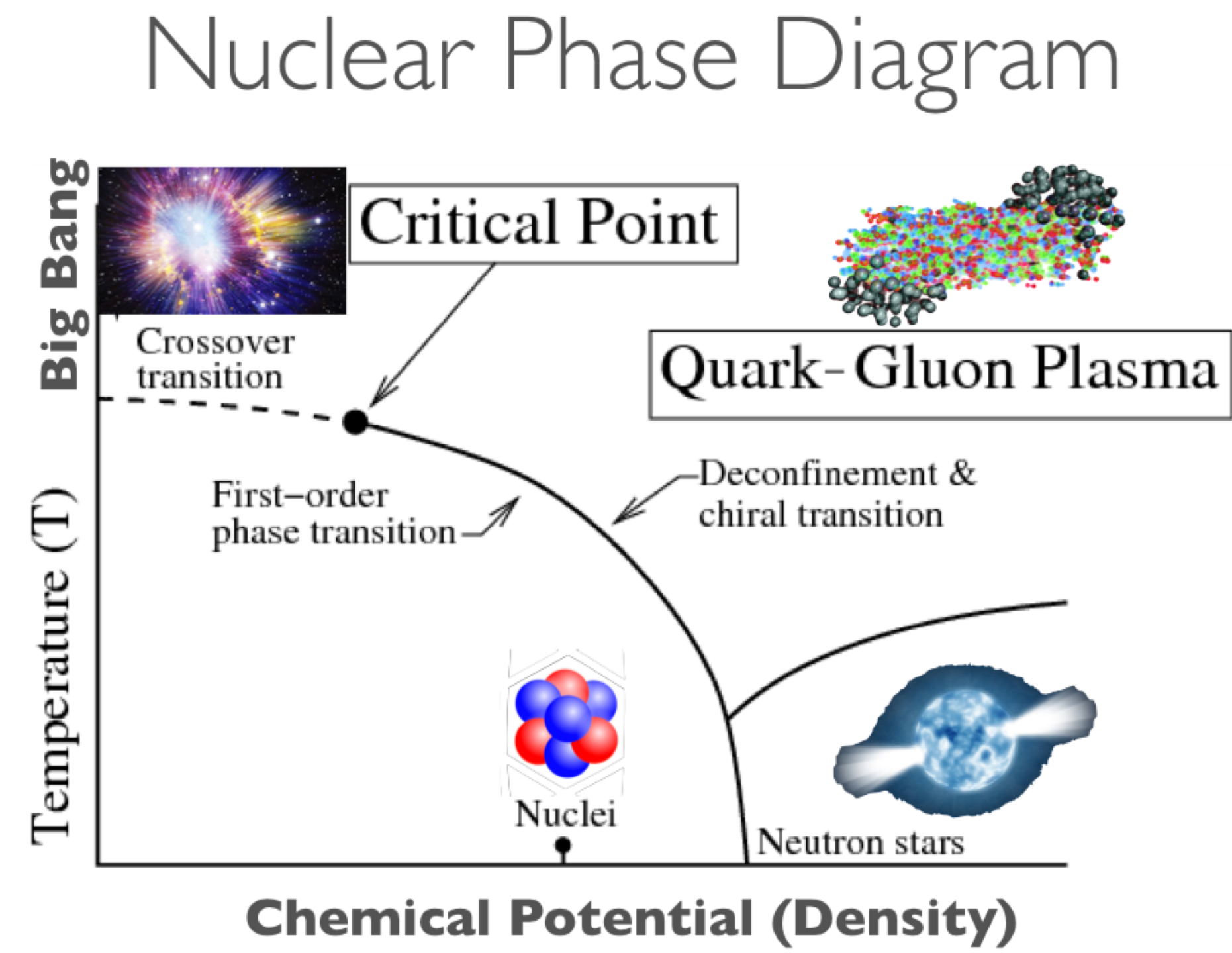
Machine learning for classifying the chiral phase transition in AdS/QCD

Beixi Hao

Faculty Advisor: Dr. Sean Bartz

Indiana State University

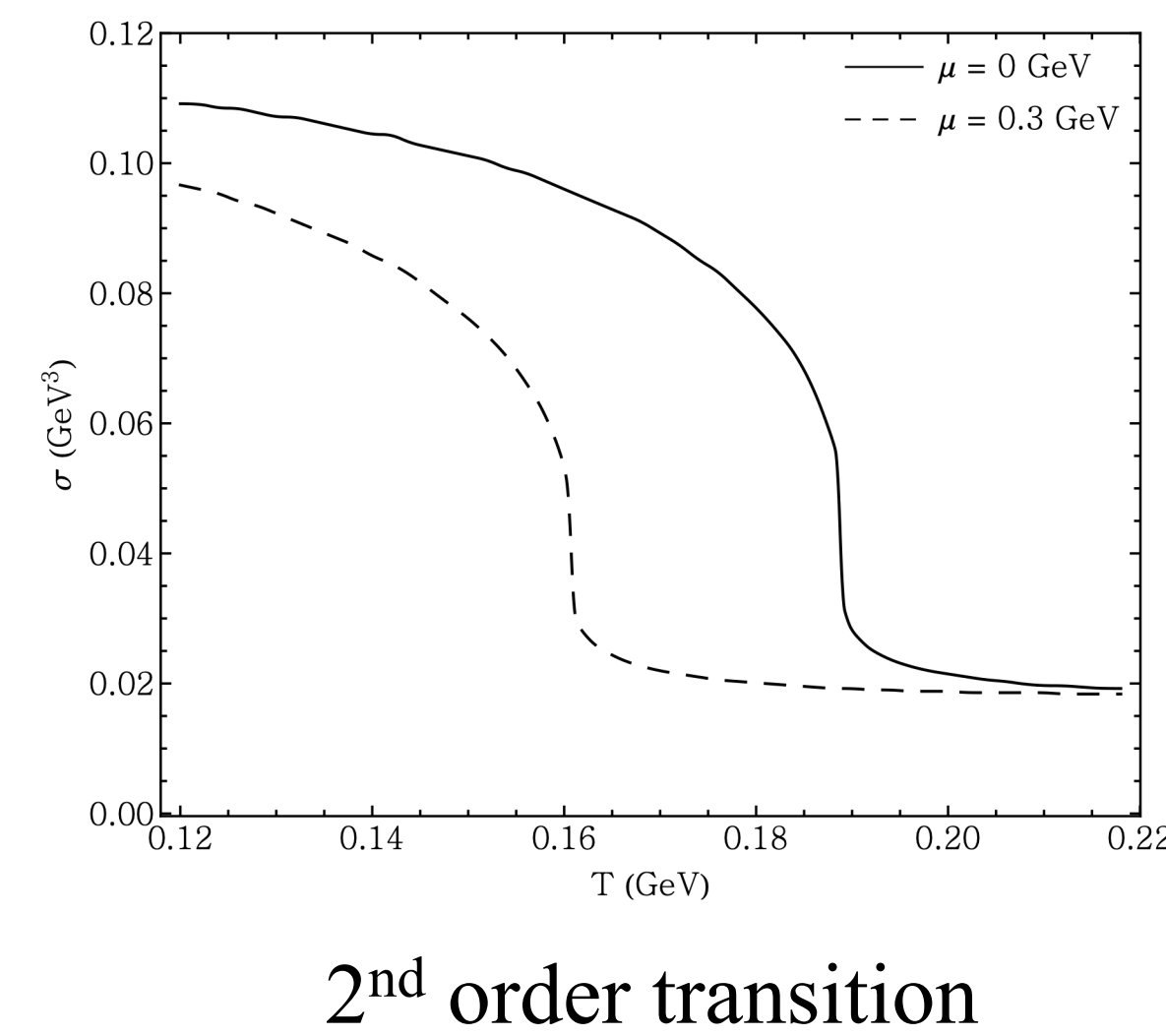
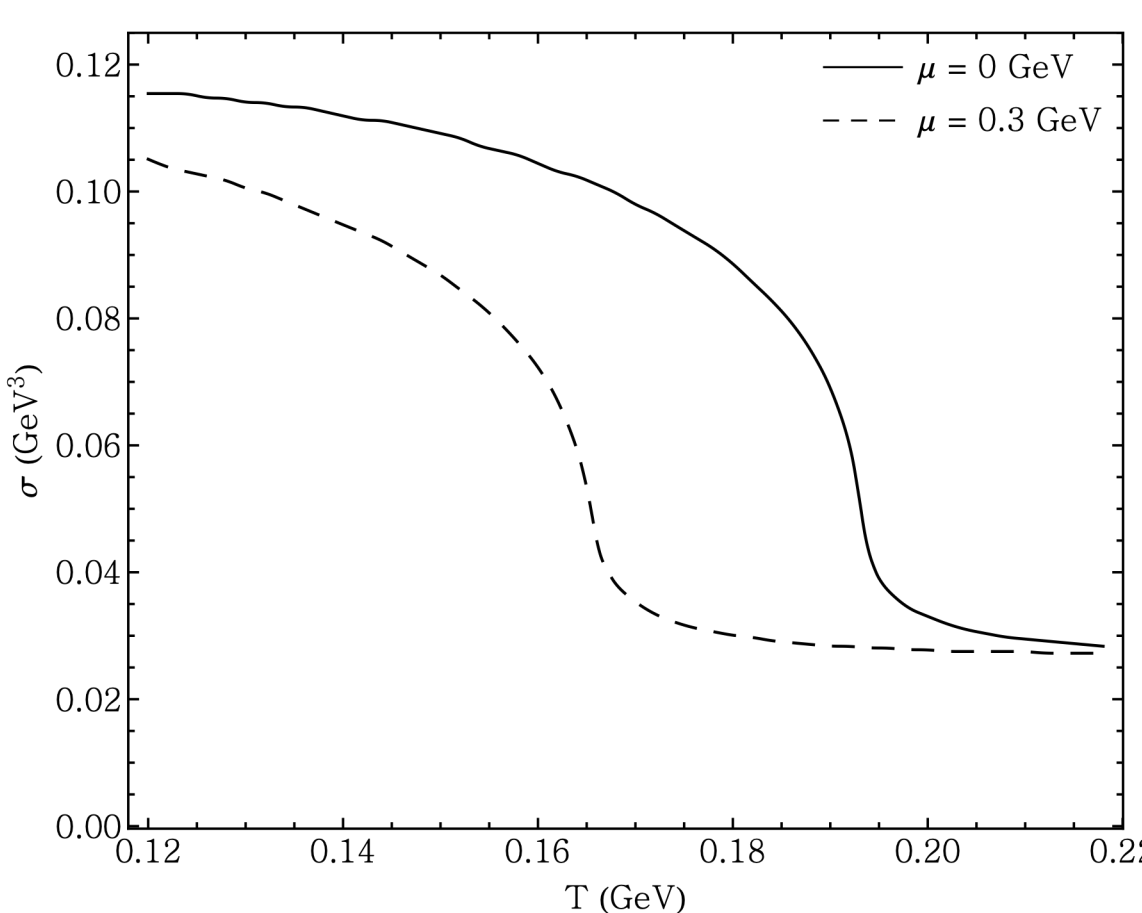
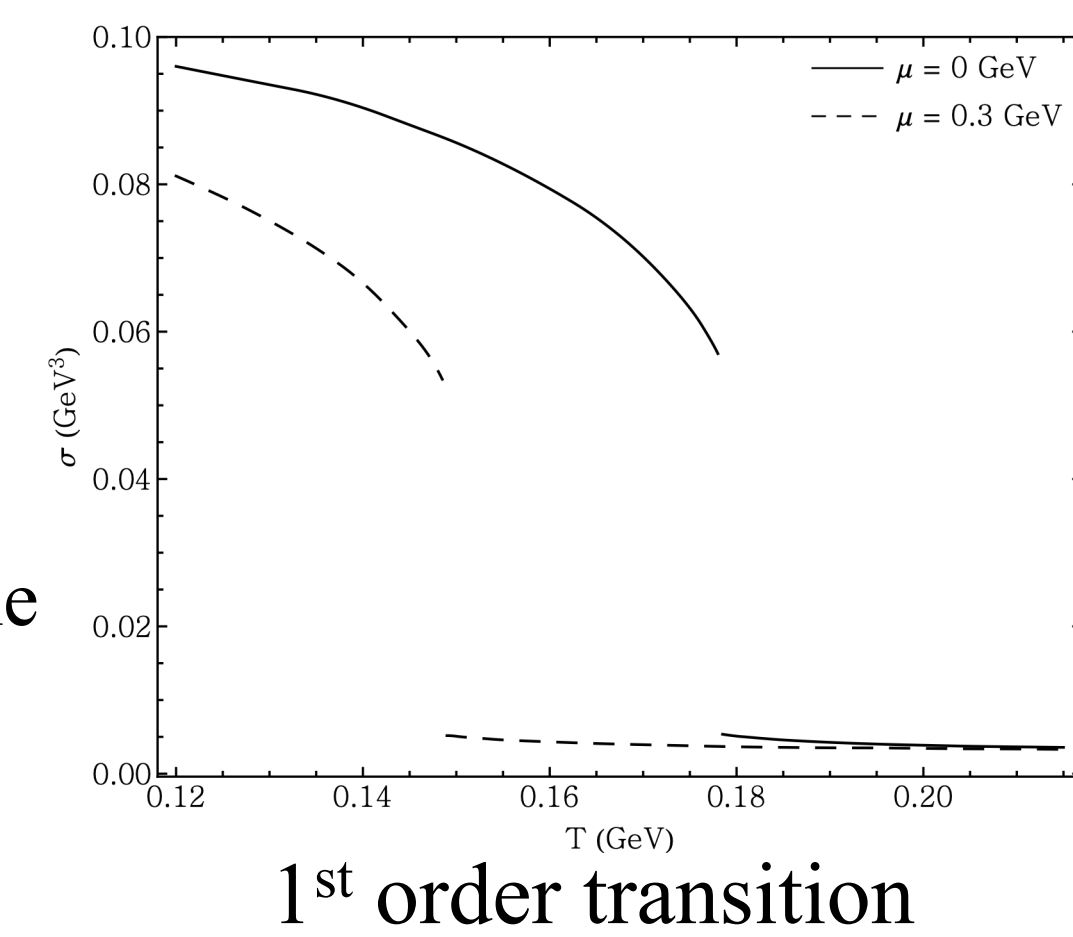
Introduction – Physics Background



Quark-gluon plasma (QGP) occurs at high temperature and density. The quarks and gluons move freely, and the plasma behaves like a perfect liquid. We analyze the phase transition from ordinary nuclear matter.

Transition Types

- Chiral parameter σ 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

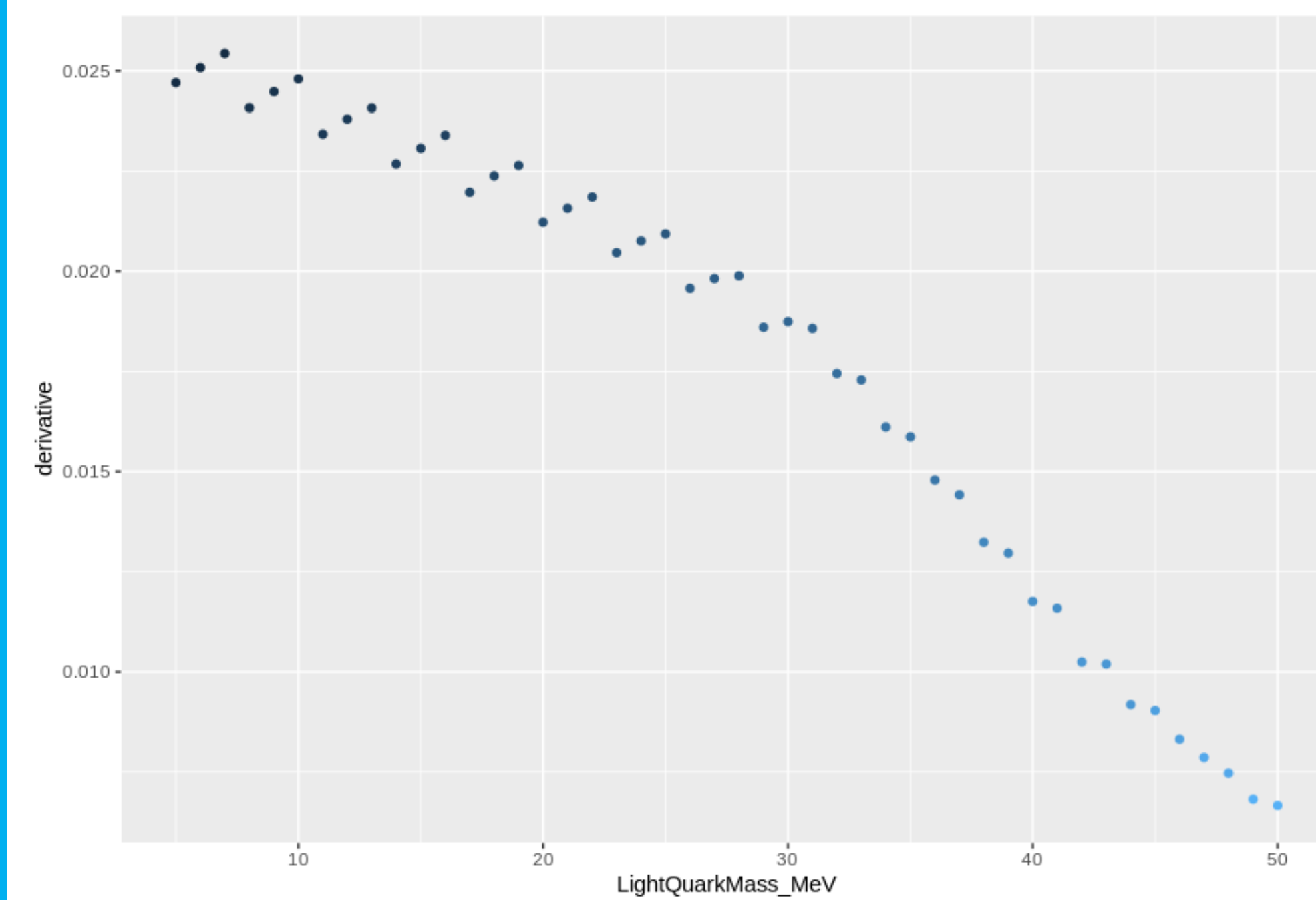
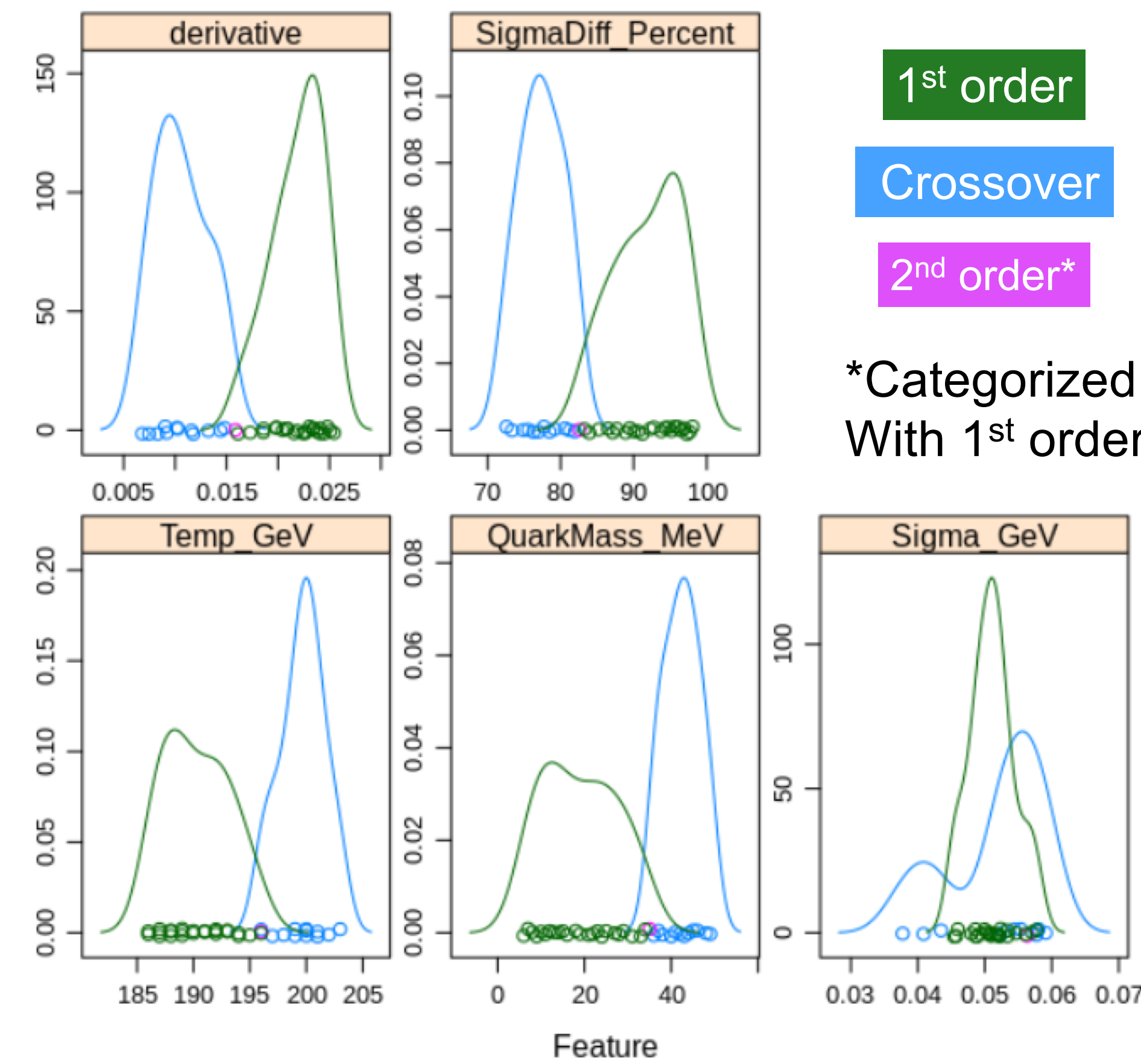


Machine Learning Model

Types of input data for machine learning

- Maximum derivative of chiral parameter σ**
- Percent change of chiral parameter**
- Temperature at maximum derivative
- σ 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
- 1st 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

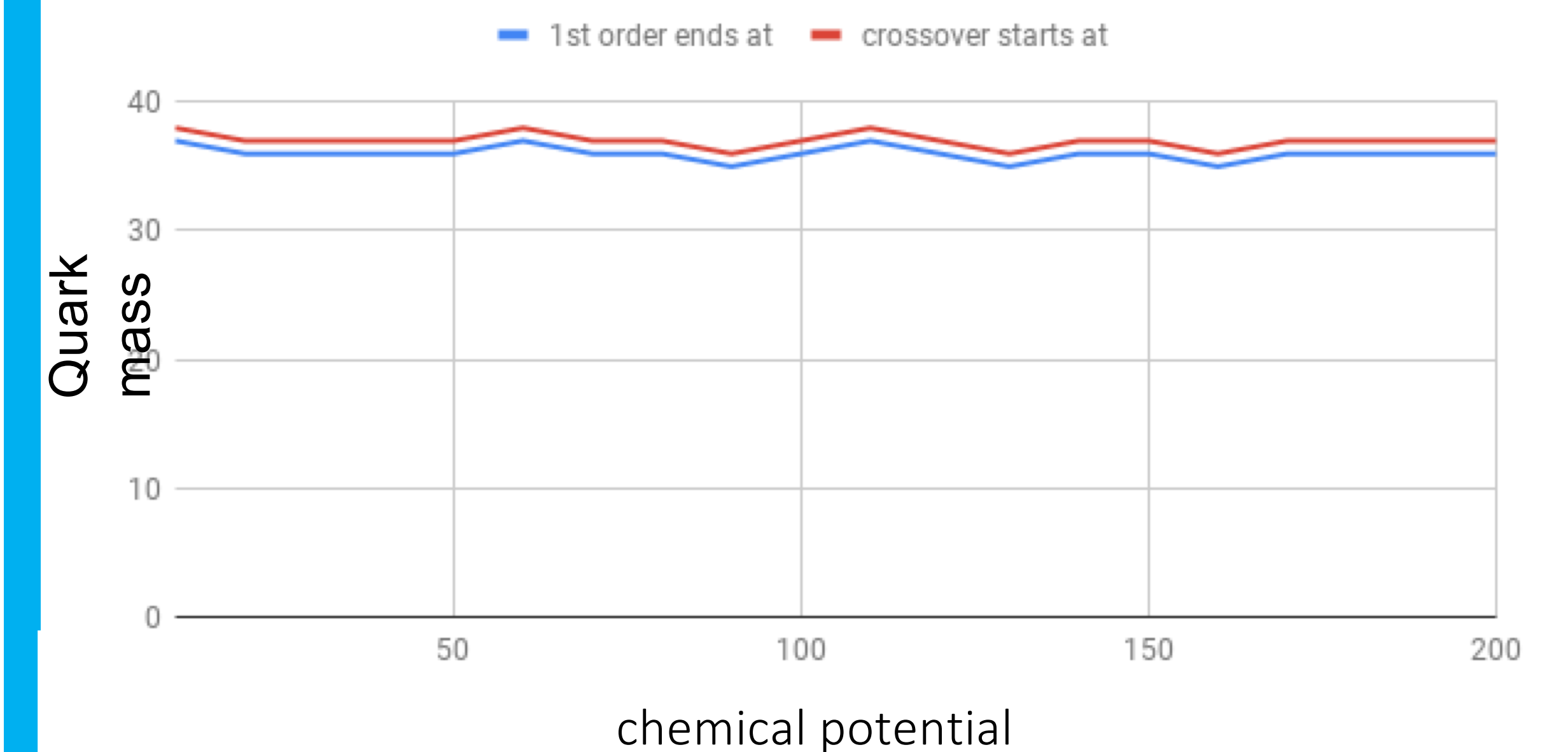
2 +1 flavor model:
Quark masses can differ

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

1st order ends at and crossover starts at



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

Acknowledgements

Indiana State University SURE Program

Indiana State University Physics and Chemistry Department