

# **GNN for Water Cherenkov Detector**

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# Fixing Correlation Implementation

2. Say that

**This can be interesting?**

$$X_1 = Z_1 + Z_c$$

$$X_2 = Z_2 + Z_c$$

where  $Z_i \sim \mathcal{N}(0, e^{O_i})$  are "private information" and  $Z_c \sim \mathcal{N}(0, e^{O_3})$  is a "shared" gaussian information. Then we will have that

$$v_1 = e^{O_1} + e^{O_3}$$

$$v_2 = e^{O_2} + e^{O_3}$$

$$\rho = \frac{e^{O_3}}{\sqrt{v_1 \cdot v_2}}$$

**Last time showed:**

**But in this case, the correlation can only be positive!**

**Instead:**

$$f_{\mathbf{X}}(x_1, \dots, x_k) = \frac{\exp\left(-\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1}(\mathbf{x} - \boldsymbol{\mu})\right)}{\sqrt{(2\pi)^k |\boldsymbol{\Sigma}|}}$$
$$\boldsymbol{\Sigma}^{-1} = A^T A = \begin{pmatrix} \alpha_{11} & 0 & 0 & 0 \\ \alpha_{12} & \alpha_{22} & 0 & 0 \\ \alpha_{13} & \alpha_{23} & \alpha_{33} & 0 \\ \alpha_{14} & \alpha_{24} & \alpha_{34} & \alpha_{44} \end{pmatrix} \begin{pmatrix} \alpha_{11} & \alpha_{12} & \alpha_{13} & \alpha_{14} \\ 0 & \alpha_{22} & \alpha_{23} & \alpha_{24} \\ 0 & 0 & \alpha_{33} & \alpha_{34} \\ 0 & 0 & 0 & \alpha_{44} \end{pmatrix}$$

The matrix  $A^T A$  has to be positive definite, which imposes restriction on the network parameter output. Specifically:

1. The max value has to be on diagonal.
2. The average of any 2 diagonal elements must be larger than the real part of corresponding off-diagonal term  $a_{ii} + a_{jj} > |2\text{Re}(a_{ij})|$ .

But since we are only using 2-dimensional Hermitian matrices, this can be relatively simple to apply.

# Fixing Correlation Implementation

$$f_{\mathbf{X}}(x_1, \dots, x_k) = \frac{\exp\left(-\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1}(\mathbf{x} - \boldsymbol{\mu})\right)}{\sqrt{(2\pi)^k |\boldsymbol{\Sigma}|}} \quad \boldsymbol{\Sigma}^{-1} = \mathbf{A}^T \mathbf{A} = \begin{pmatrix} \alpha_{11} & 0 & 0 & 0 \\ \alpha_{12} & \alpha_{22} & 0 & 0 \\ \alpha_{13} & \alpha_{23} & \alpha_{33} & 0 \\ \alpha_{14} & \alpha_{24} & \alpha_{34} & \alpha_{44} \end{pmatrix} \begin{pmatrix} \alpha_{11} & \alpha_{12} & \alpha_{13} & \alpha_{14} \\ 0 & \alpha_{22} & \alpha_{23} & \alpha_{24} \\ 0 & 0 & \alpha_{33} & \alpha_{34} \\ 0 & 0 & 0 & \alpha_{44} \end{pmatrix}$$

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In NN, apply the constraint on off-diagonal terms by

$$- A_{12} = (\text{Sigmoid}(A_{12}) - 0.5) * (A_{11} + A_{22})$$

So that the resulted diagonal term is within  $(-0.5, 0.5) * (A_{11} + A_{22})$ , and hence satisfying the requirements.

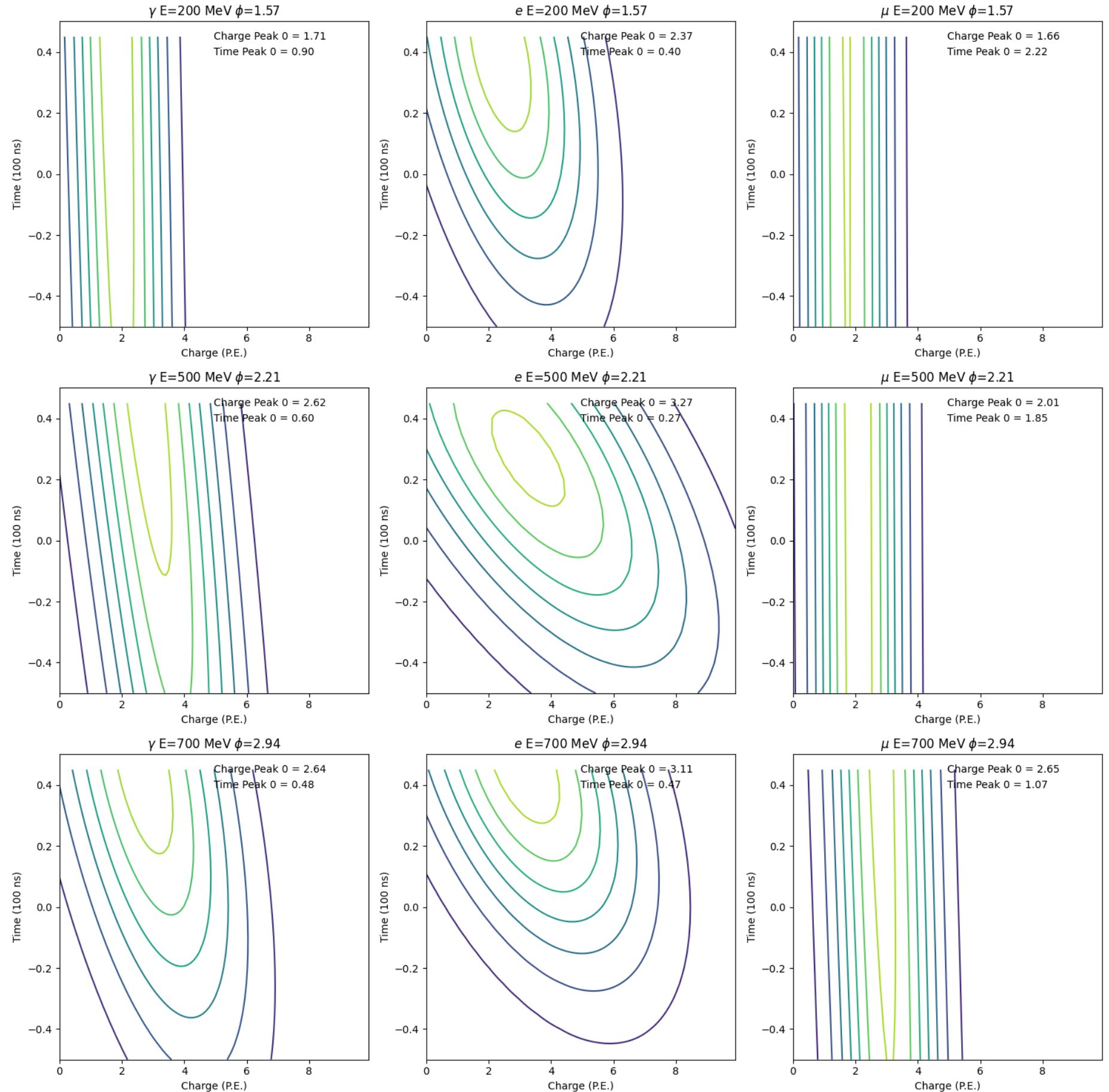
However, the more correct way of this should be done on the squared terms, i.e.  $A_{11}^2$ , etc.

But with the square, there is an extra complication to determine the sign of  $A_{12}$ . The current implementation can result in a wider range of  $A_{12}$  than what's mathematically allowed (since  $a^2 + b^2 < (a+b)^2$ ), but hopefully it will be a small effect...

# Example of The New Implementation

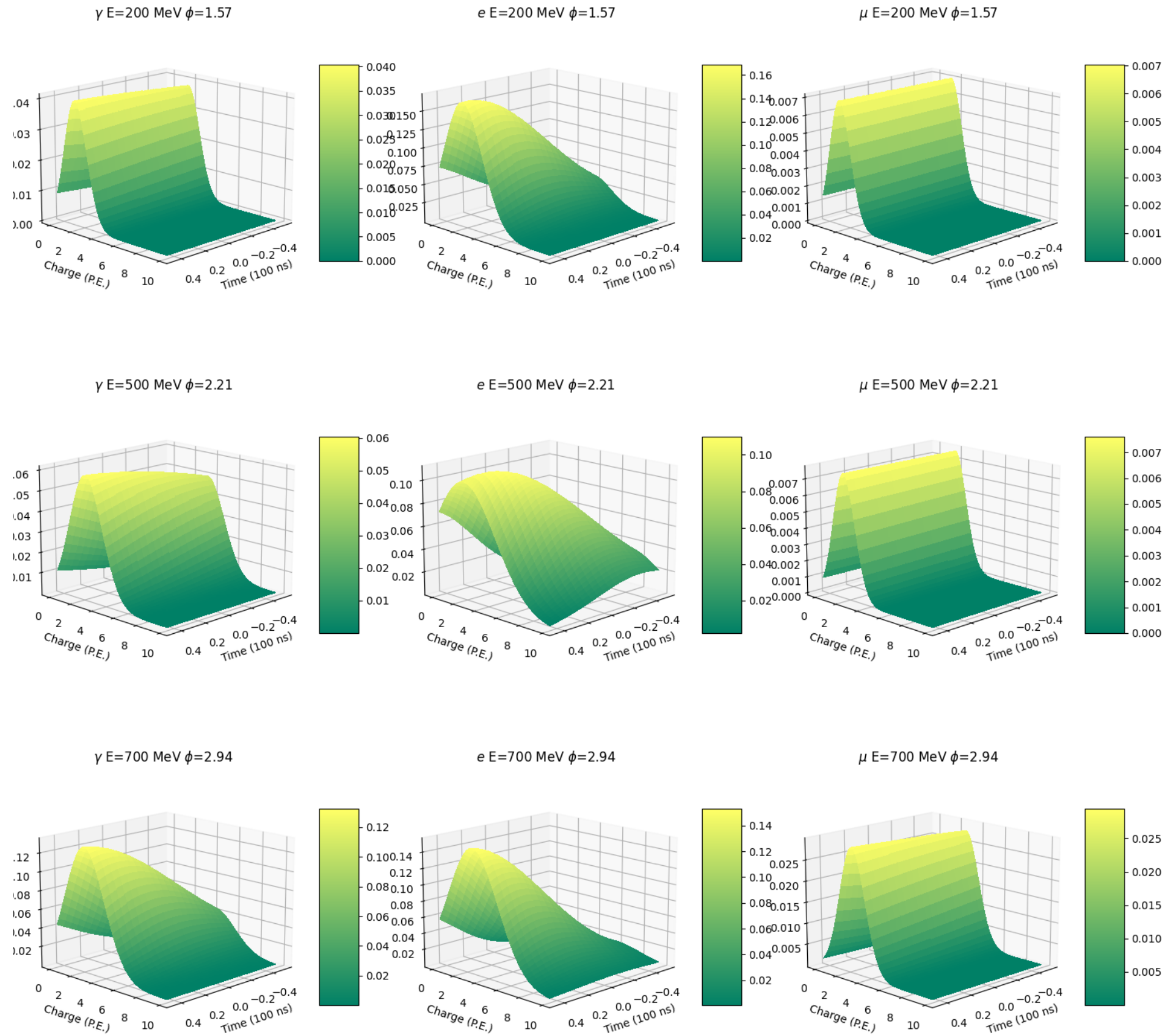
1 Gauss

Sorry time unit  
should be 10 ns,  
not 100 ns



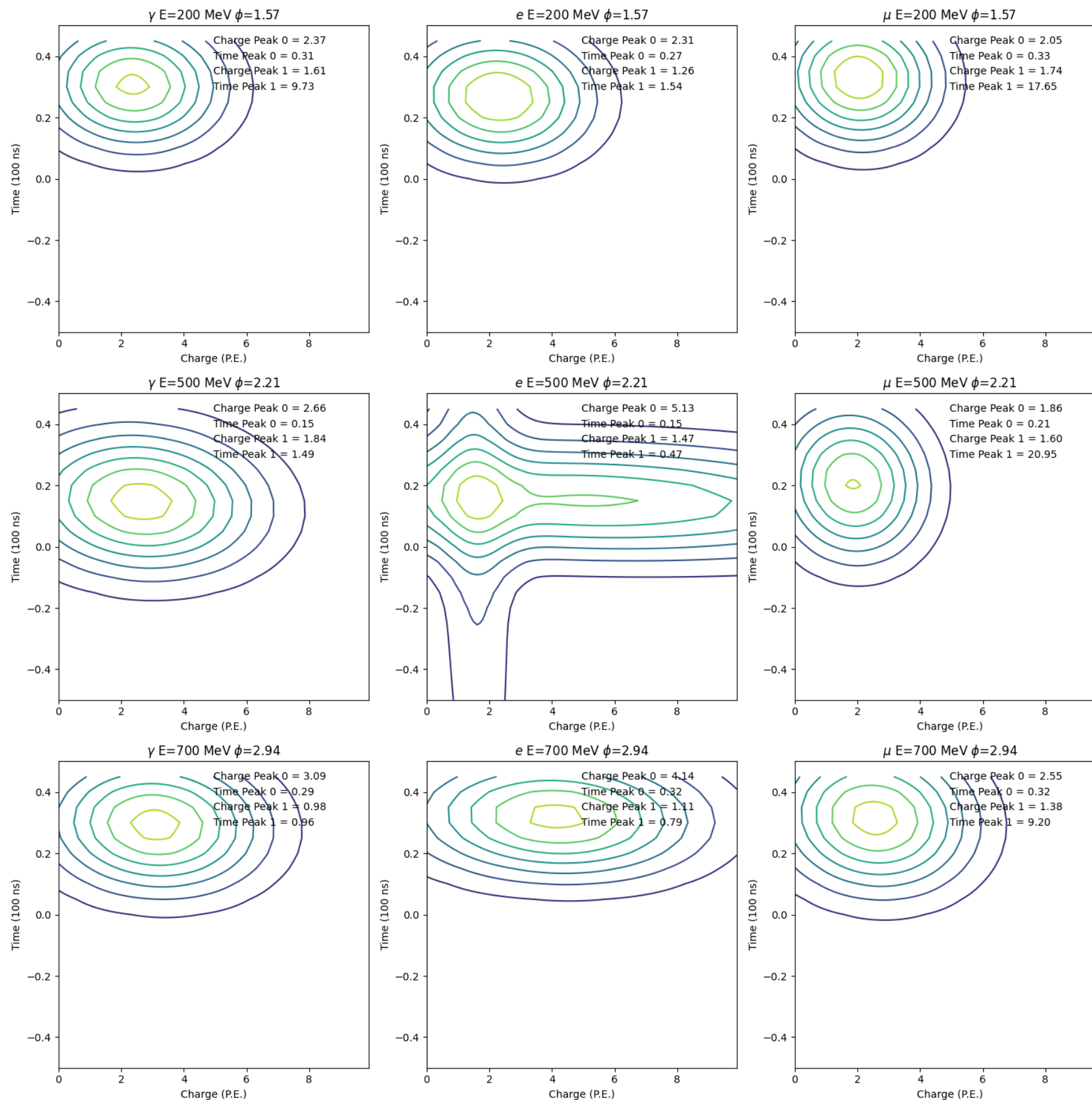
# Example of The New Implementation

## 1 Gauss



# Example of The New Implementation

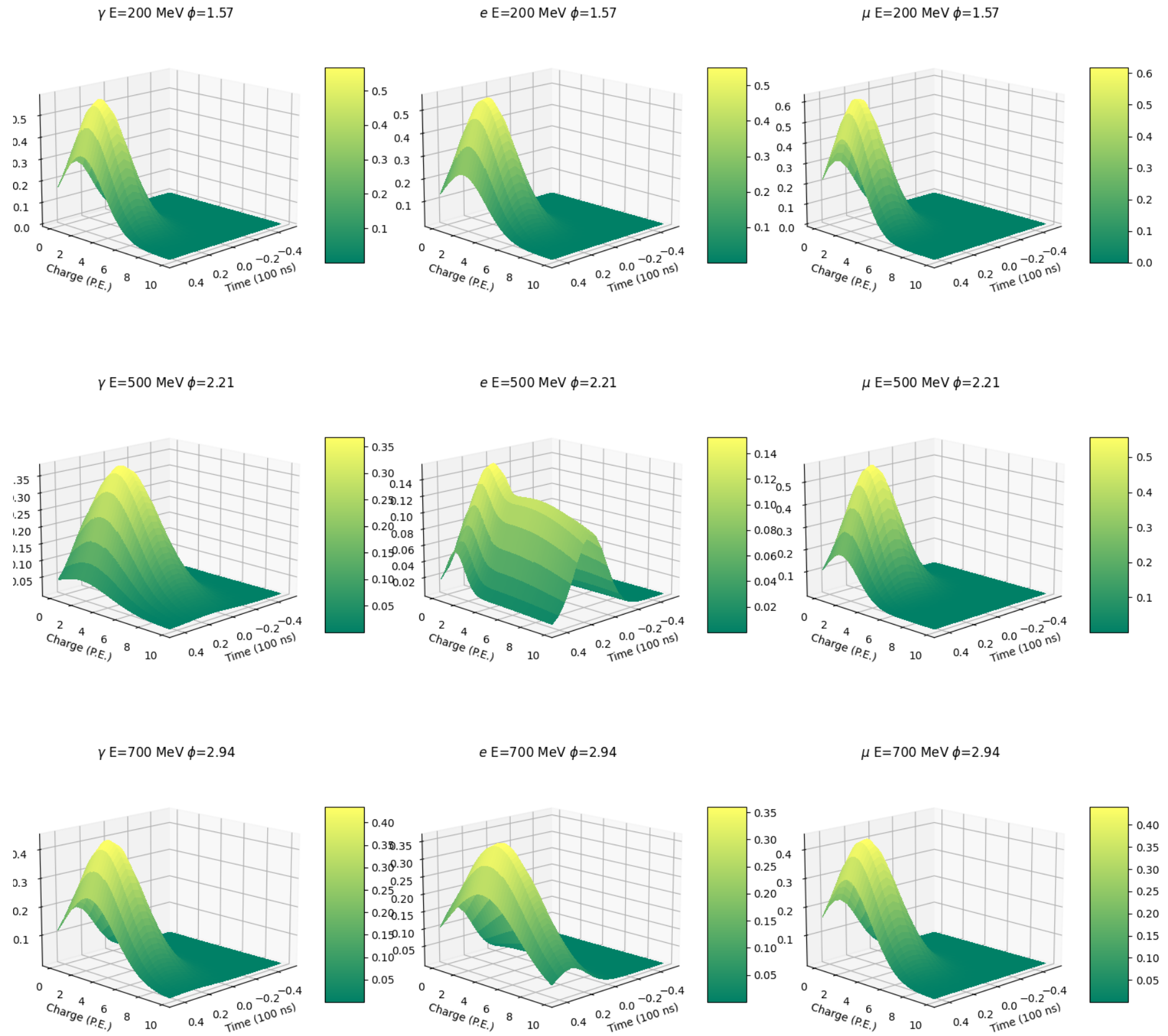
## 2 Gauss





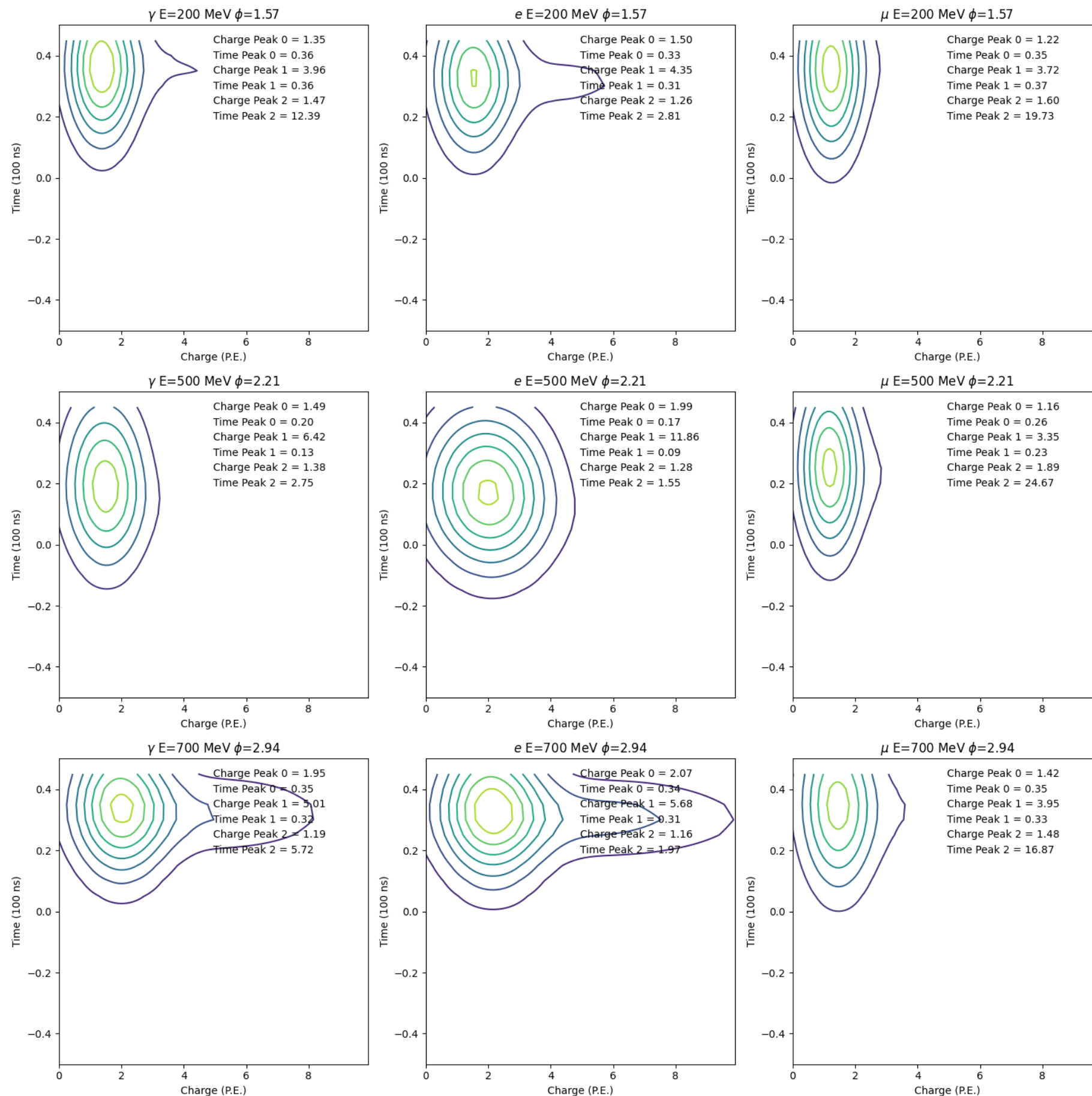
# Example of The New Implementation

## 2 Gauss



# Example of The New Implementation

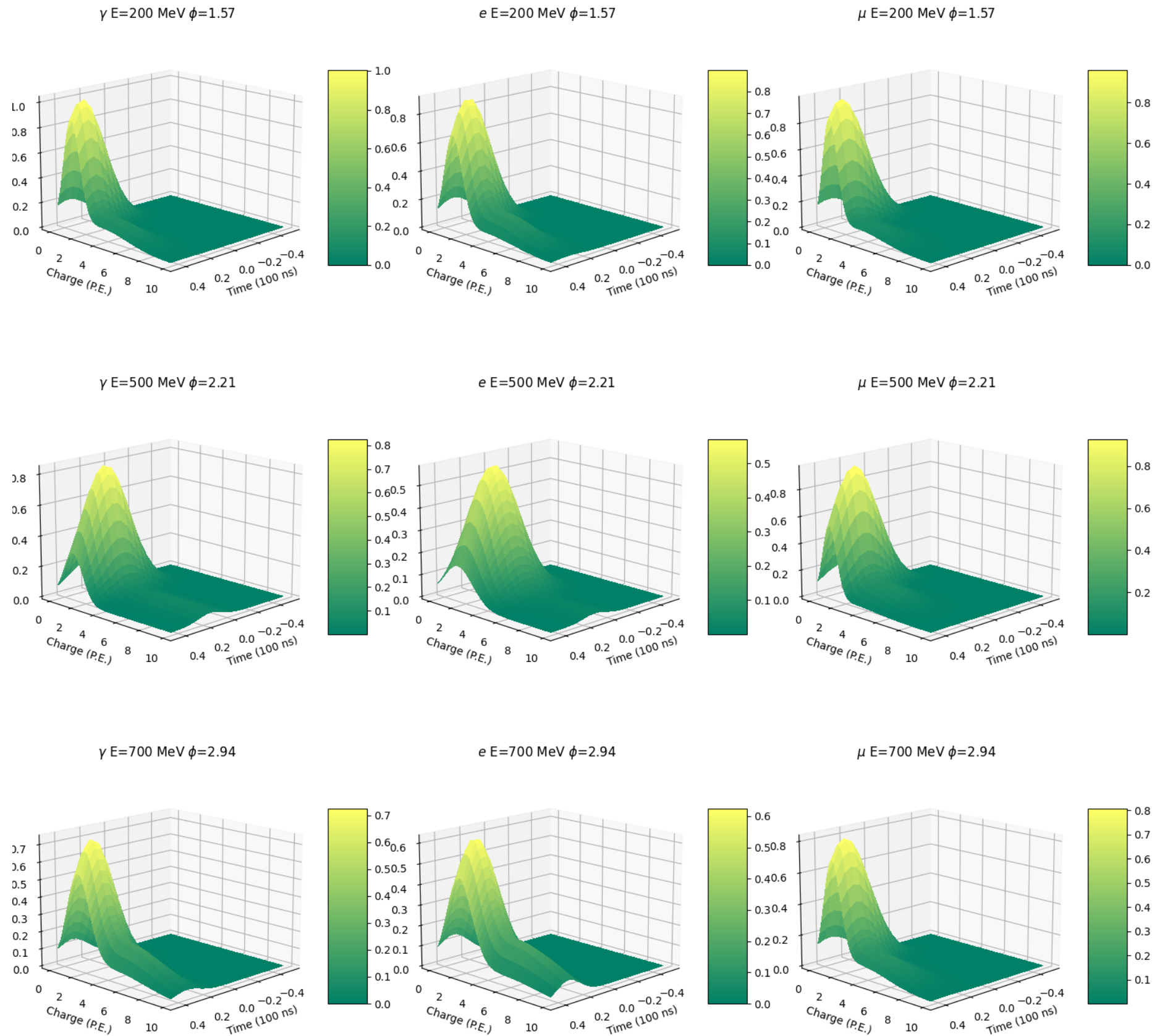
3 Gauss





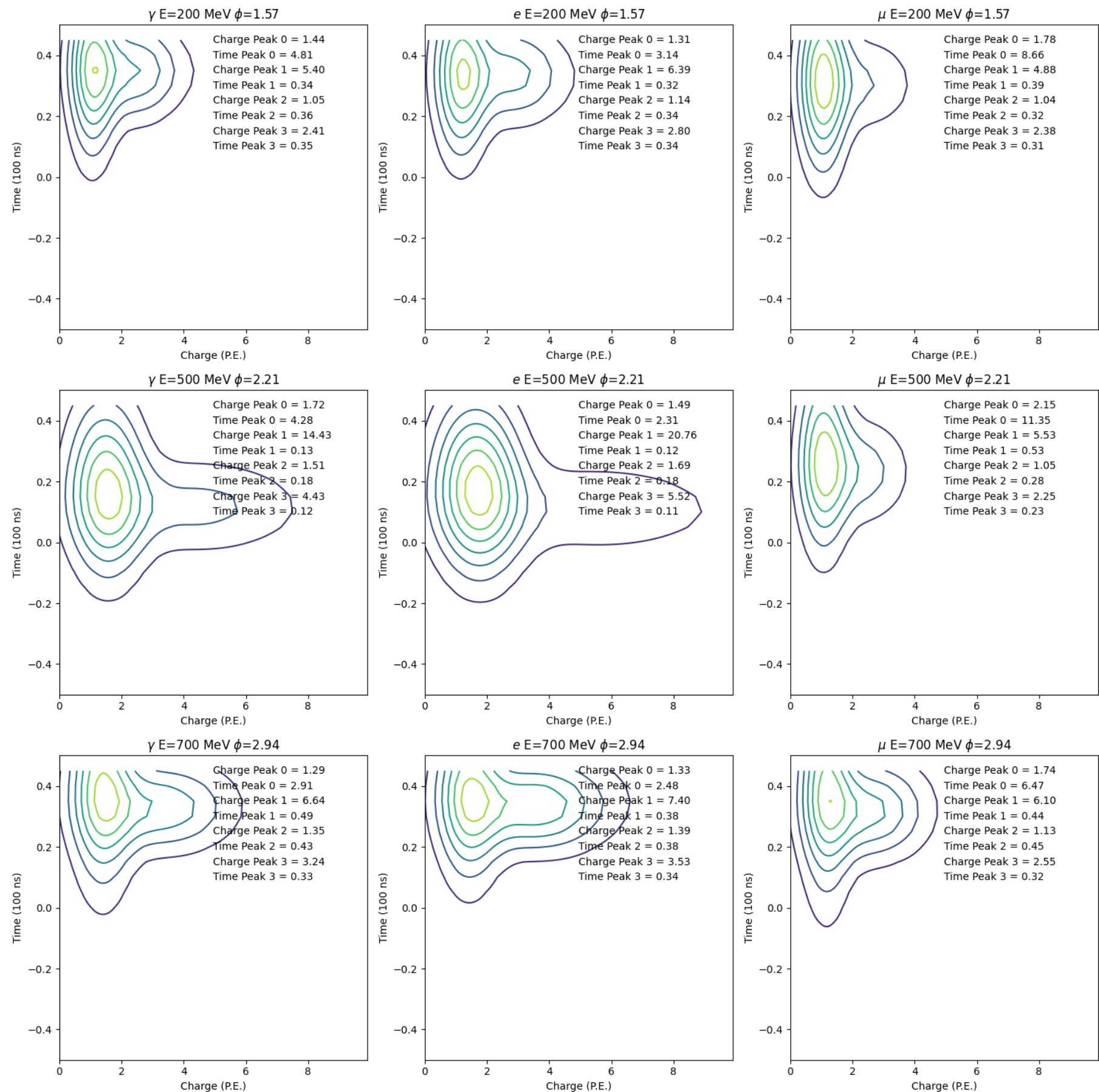
# Example of The New Implementation

## 3 Gauss



# Example of The New Implementation

4 Gauss

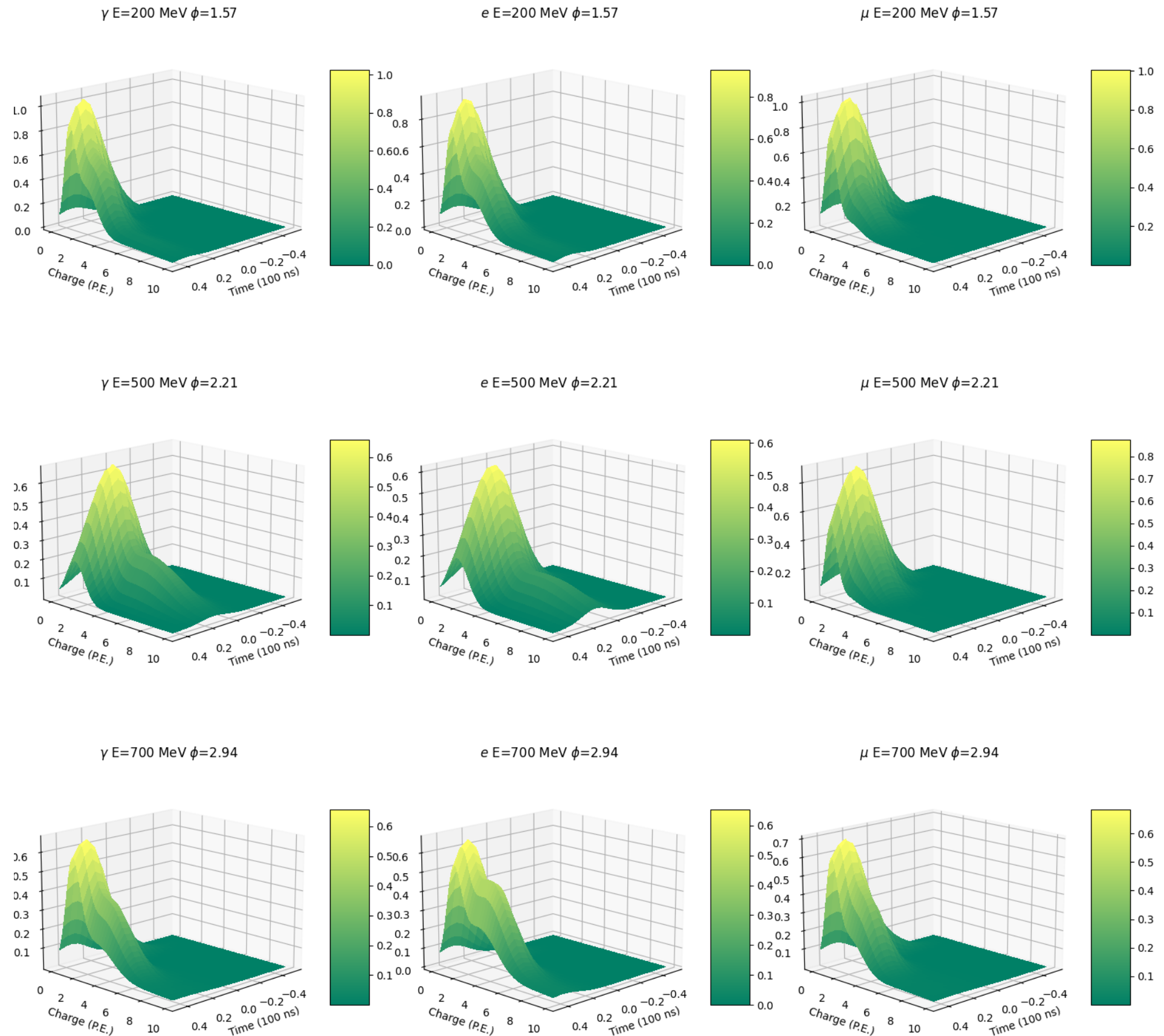


# Example of The New Implementation

## 4 Gauss

- 5 Gauss also runs fine. But the job lasted longer than 12 hrs and got terminated, 87% done.

- Still have intermediate outputs.



# Summary

$$f_{\mathbf{X}}(x_1, \dots, x_k) = \frac{\exp\left(-\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^T \boldsymbol{\Sigma}^{-1}(\mathbf{x} - \boldsymbol{\mu})\right)}{\sqrt{(2\pi)^k |\boldsymbol{\Sigma}|}} \quad \Sigma^{-1} = A^T A = \begin{pmatrix} \alpha_{11} & 0 & 0 & 0 \\ \alpha_{12} & \alpha_{22} & 0 & 0 \\ \alpha_{13} & \alpha_{23} & \alpha_{33} & 0 \\ \alpha_{14} & \alpha_{24} & \alpha_{34} & \alpha_{44} \end{pmatrix} \begin{pmatrix} \alpha_{11} & \alpha_{12} & \alpha_{13} & \alpha_{14} \\ 0 & \alpha_{22} & \alpha_{23} & \alpha_{24} \\ 0 & 0 & \alpha_{33} & \alpha_{34} \\ 0 & 0 & 0 & \alpha_{44} \end{pmatrix}$$

Implemented this “matrix method” in the NN, with changed definition of network outputs.  
The network now outputs  $a_{11}$ ,  $a_{12}$ ,  $a_{22}$ , which in terms of sigma and correlation is

$$a_{11} = \frac{1}{\sigma_t \sqrt{1 - \rho^2}}$$

$$a_{22} = \frac{1}{\sigma_q}$$

$$a_{12} = \frac{-\rho}{\sigma_q \sqrt{1 - \rho^2}}$$

Working on the maths for the LogNorm case.

# Back Ups