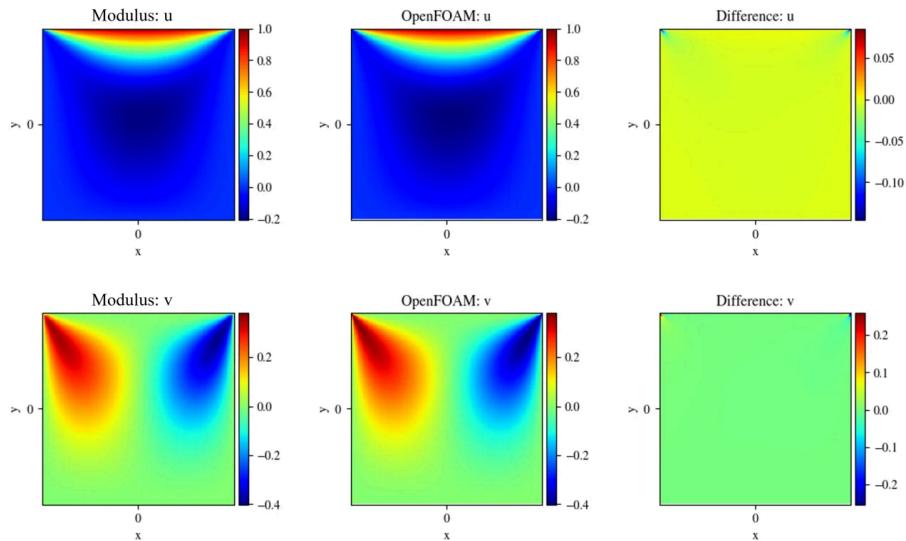


# NVIDIA Modulus Part 3: Turbulent physics: Zero Equation Turbulence Model

## Part 2: Introducing Lid Driven Cavity (LDC) Example

[https://docs.nvidia.com/deeplearning/modulus/modulus-sym/user\\_guide/basics/lid\\_driven\\_cavity\\_flow.html](https://docs.nvidia.com/deeplearning/modulus/modulus-sym/user_guide/basics/lid_driven_cavity_flow.html)

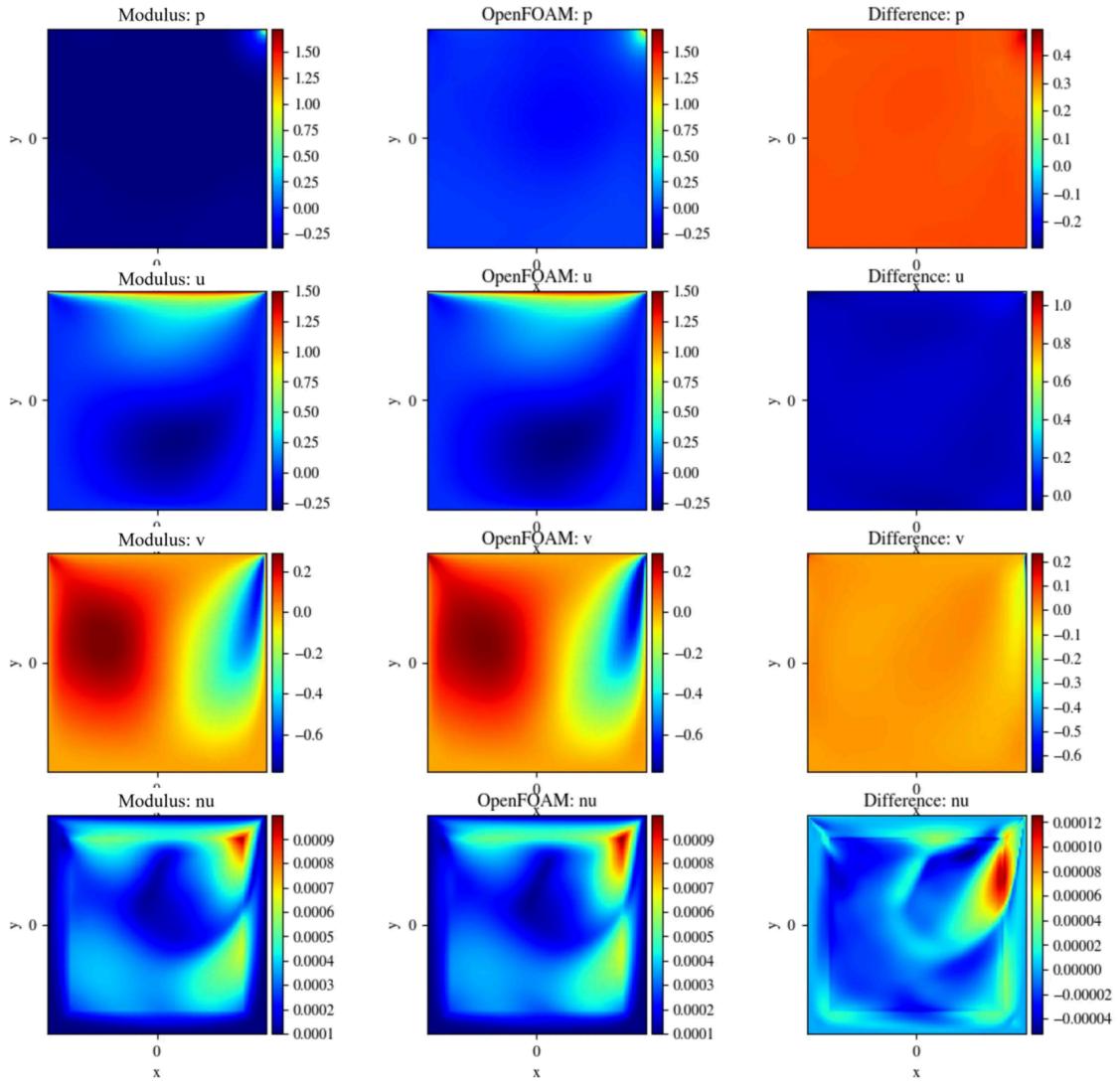


## Part 3: Turbulent physics: Zero Equation Turbulence Model

- Integrating the **Zero Equation turbulence model** into a Lid-Driven Cavity (LDC) flow
- Code walkthrough
- Visualize outputs

<https://github.com/AI-ME-Ben/NVIDIA-Modulus-Reproduce/tree/main>

[https://catalog.ngc.nvidia.com/orgs/nvidia/teams/modulus/resources/modulus\\_sym\\_examples\\_supplemental\\_materials](https://catalog.ngc.nvidia.com/orgs/nvidia/teams/modulus/resources/modulus_sym_examples_supplemental_materials)



## Part 1: Integrating the Zero Equation turbulence model into a Lid-Driven Cavity (LDC) flow

- The previous LDC case had  $Re = 10$ , but here it's increased to **1000**.
- The model computes **effective viscosity** using the mean strain rate tensor.

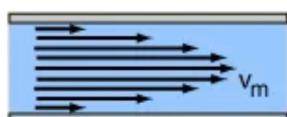
### 1. Reynolds Number Formula

$$Re = \frac{UL}{\nu}$$

where:

- $U$  = characteristic velocity
- $L$  = characteristic length
- $\nu$  = molecular (kinematic) viscosity

This equation shows that  $Re$  is directly affected by molecular viscosity ( $\nu$ ), but it does not include turbulent viscosity ( $\nu_t$ ).



Laminar flow, Reynolds low



Turbulence, Reynolds high

## 1. Standard 2D Steady-State Navier-Stokes Equations (Incompressible)

For an incompressible flow ( $\nabla \cdot \mathbf{u} = 0$ ):

$$u \frac{\partial u}{\partial x} + v \frac{\partial u}{\partial y} = -\frac{\partial p}{\partial x} + \nu \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right)$$
$$u \frac{\partial v}{\partial x} + v \frac{\partial v}{\partial y} = -\frac{\partial p}{\partial y} + \nu \left( \frac{\partial^2 v}{\partial x^2} + \frac{\partial^2 v}{\partial y^2} \right)$$

## 2. Modified Navier-Stokes Equations with Effective Viscosity

In turbulence models, we replace molecular viscosity  $\nu$  with effective viscosity  $\nu_{\text{eff}}$ :

$$\nu_{\text{eff}} = \nu + \nu_t$$

where:

- $\nu$  = molecular kinematic viscosity
- $\nu_t$  = turbulent viscosity (depends on turbulence model)

Now, the modified equations become:

$$u \frac{\partial u}{\partial x} + v \frac{\partial u}{\partial y} = -\frac{\partial p}{\partial x} + \nu_{\text{eff}} \left( \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \right)$$
$$u \frac{\partial v}{\partial x} + v \frac{\partial v}{\partial y} = -\frac{\partial p}{\partial y} + \nu_{\text{eff}} \left( \frac{\partial^2 v}{\partial x^2} + \frac{\partial^2 v}{\partial y^2} \right)$$

where  $\nu_{\text{eff}}$  accounts for both laminar and turbulent diffusion effects.

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### Understanding Effective Viscosity in the Zero Equation Turbulence Model

The Zero Equation Turbulence Model (also called the Algebraic Turbulence Model) estimates turbulence effects using an effective (turbulent) viscosity  $\nu_{\text{eff}}$ . Instead of solving additional transport equations (like in k-epsilon or k-omega models), it computes turbulent viscosity directly using local velocity gradients and wall distances.

## 1. Definition of Effective Viscosity

In the `ZeroEquation` class, the effective viscosity  $\nu_{\text{eff}}$  is computed as:

$$\nu_{\text{eff}} = \nu + \rho \cdot l_m^2 \cdot \sqrt{G}$$

where:

- $\nu$  = kinematic viscosity (molecular viscosity of the fluid)
- $\rho$  = density of the fluid
- $l_m$  = mixing length (describes turbulence scale, dependent on wall distance)
- $G$  = strain rate magnitude, which represents turbulence production

## 2. Breakdown of Key Components

### (A) Mixing Length $l_m$

The mixing length model is a simplified approach to turbulence modeling. It assumes that the turbulent viscosity is proportional to the velocity gradient and a characteristic length scale. The mixing length is given by:

$$l_m = \min(\kappa d, \beta d_{\max})$$

where:

- $\kappa = 0.419$  (Von Kármán constant, empirical value)
- $d$  = normal distance to the nearest wall (function `"sdf"` in code)
- $\beta = 0.09$  (scaling factor for maximum distance)
- $d_{\max}$  = maximum reference distance in the flow field

This equation ensures that the mixing length scales correctly near walls and does not grow indefinitely in free shear flows.



- Near walls:  $l_m$  is small due to the wall distance constraint, leading to lower turbulent viscosity.
- Away from walls:  $l_m$  can grow, allowing higher turbulent viscosity, capturing the effects of larger eddies.
- In free shear flows (e.g., mixing layers, jets): The model adapts dynamically using velocity gradients.

## (B) Strain Rate Magnitude $G$

The strain rate magnitude represents how fast the velocity gradients change in a given region, and it plays a crucial role in turbulence generation.

$$G = 2 \left( \frac{\partial u}{\partial x} \right)^2 + 2 \left( \frac{\partial v}{\partial y} \right)^2 + 2 \left( \frac{\partial w}{\partial z} \right)^2 \\ + \left( \frac{\partial u}{\partial y} + \frac{\partial v}{\partial x} \right)^2 + \left( \frac{\partial u}{\partial z} + \frac{\partial w}{\partial x} \right)^2 + \left( \frac{\partial v}{\partial z} + \frac{\partial w}{\partial y} \right)^2$$

This formula sums the squares of all velocity gradients, effectively capturing how much "shearing" is present in the flow.

- If velocity gradients are large,  $G$  increases, leading to a higher turbulent viscosity.
- If velocity gradients are small,  $G$  is low, and the flow behaves more like a laminar flow.

## 3. Physical Interpretation

The effective viscosity  $\nu_{\text{eff}}$  is a combination of:

1. Molecular viscosity  $\nu$ : Represents standard fluid resistance (important in laminar flows).
2. Turbulent viscosity  $\nu_t = \rho l_m^2 \sqrt{G}$ : Accounts for additional momentum transfer due to turbulence.

## 4. Advantages of the Zero Equation Model

- ✓ No additional transport equations: Unlike  $k-\epsilon$  or  $k-\omega$ , this model avoids solving extra PDEs.
- ✓ Fast computation: Since it relies only on algebraic calculations, it's suitable for real-time simulations.
- ✓ Good for simple flows: Works well for boundary layer flows and shear-driven turbulence (e.g., lid-driven cavity).
- ✗ Limited accuracy: Does not capture complex turbulence effects like separation or wake flows.

## 5. How Modulus Sym Uses This Model

In Modulus Sym, the `ZeroEquation` model is implemented as an additional PDE constraint in the computational graph:

```
python
ze = ZeroEquation(nu=1e-4, max_distance=height / 2, dim=2, time=False)
```

- `nu=1e-4`: Specifies the base kinematic viscosity.
- `max_distance=height / 2`: Sets the reference maximum wall distance.
- `dim=2`: Configures for 2D turbulence.
- `time=False`: Uses a steady-state formulation.

Once added, the solver ensures that the effective viscosity is computed at each point dynamically.

```
from modulus.sym.eq.pdes.turbulence_zero_eq import ZeroEquation
```

<https://help.autodesk.com/view/SCDSE/2019/ENU/?guid=GUID-BBA4E008-8346-465B-9FD3-D193CF108AF0>

<https://help.autodesk.com/view/SCDSE/2019/ENU/?guid=GUID-4148B3AE-8B8E-49AA-A604-1DAE92CD8D02>

<https://turbmodels.larc.nasa.gov/>

## Part 2: Code Walkthrough

### Geometry and Governing Equations

- Governing Equations:
  - **Navier-Stokes Equations** (with turbulence model viscosity)
  - **Zero Equation Model** (for effective viscosity calculation)

### Importing Required Packages

```
import os
import warnings
from sympy import Symbol, Eq, Abs
import torch
import modulus.sym
from modulus.sym.hydra import to_absolute_path, instantiate_arch, ModulusConfig
from modulus.sym.utils.io import csv_to_dict
from modulus.sym.solver import Solver
from modulus.sym.domain import Domain
from modulus.sym.geometry.primitives_2d import Rectangle
from modulus.sym.domain.constraint import PointwiseBoundaryConstraint, PointwiseInteriorConstraint
from modulus.sym.domain.validator import PointwiseValidator
from modulus.sym.eq.pdes.navier_stokes import NavierStokes
from modulus.sym.eq.pdes.turbulence_zero_eq import ZeroEquation
```

### Defining Geometry

```
# Define cavity geometry
height, width = 0.1, 0.1
x, y = Symbol("x"), Symbol("y")
rec = Rectangle((-width / 2, -height / 2), (width / 2, height / 2))
```

### Adding Turbulence Model

- The **Zero Equation Model** is instantiated to calculate the **effective viscosity ( $\nu_t$ )**.
- The kinematic viscosity for Navier-Stokes equations is replaced with the one computed by **ZeroEquation**.

```
# Define Zero Equation turbulence model
ze = ZeroEquation(nu=1e-4, dim=2, time=False, max_distance=height / 2)

# Define Navier-Stokes equations with turbulence model viscosity
ns = NavierStokes(nu=ze.equations["nu"], rho=1.0, dim=2, time=False)
```

### Boundary Conditions

- **Top Wall:** Moving with constant velocity  **$u = 1.5, v = 0$** .
- **Other Walls:** No-slip condition  **$u = 0, v = 0$** .

```
# Create computational domain
ldc_domain = Domain()

# Top moving wall (Lid)
top_wall = PointwiseBoundaryConstraint(
    nodes=nodes,
    geometry=rec,
```

```

        outvar={"u": 1.5, "v": 0},
        batch_size=cfg.batch_size.TopWall,
        lambda_weighting={"u": 1.0 - 20 * Abs(x), "v": 1.0}, # Reduce weight at edges
        criteria=Eq(y, height / 2),
    )
    ldc_domain.add_constraint(top_wall, "top_wall")

    # No-slip walls
    no_slip = PointwiseBoundaryConstraint(
        nodes=nodes,
        geometry=rec,
        outvar={"u": 0, "v": 0},
        batch_size=cfg.batch_size.NoSlip,
        criteria=y < height / 2,
    )
    ldc_domain.add_constraint(no_slip, "no_slip")

```

## Interior Constraints (Enforcing PDEs)

- The **continuity** and **momentum equations** must be satisfied in the domain.
- **Signed Distance Field (SDF)** is used for turbulence viscosity derivatives.

```

interior = PointwiseInteriorConstraint(
    nodes=nodes,
    geometry=rec,
    outvar={"continuity": 0, "momentum_x": 0, "momentum_y": 0},
    batch_size=cfg.batch_size.Interior,
    compute_sdf_derivatives=True, # Required for turbulence viscosity gradients
    lambda_weighting={
        "continuity": Symbol("sdf"),
        "momentum_x": Symbol("sdf"),
        "momentum_y": Symbol("sdf"),
    },
)
ldc_domain.add_constraint(interior, "interior")

```

## Validation Against OpenFOAM Data

[https://catalog.ngc.nvidia.com/orgs/nvidia/teams/modulus/resources/modulus\\_sym\\_examples\\_supplemental\\_materials](https://catalog.ngc.nvidia.com/orgs/nvidia/teams/modulus/resources/modulus_sym_examples_supplemental_materials)

```

file_path = "openfoam/cavity_uniformVel_zeroEqn_refined.csv"

if os.path.exists(to_absolute_path(file_path)):
    mapping = {
        "Points:0": "x",
        "Points:1": "y",
        "U:0": "u",
        "U:1": "v",
        "p": "p",
        "d": "sdf",
        "nuT": "nu",
    }
    openfoam_var = csv_to_dict(to_absolute_path(file_path), mapping)
    openfoam_var["nu"] += 1e-4 # Effective viscosity correction

    # Create validator

```

```
openfoam_validator = PointwiseValidator(  
    nodes=nodes,  
    invar={k: v for k, v in openfoam_var.items() if k in ["x", "y", "sdf"]},  
    true_outvar={k: v for k, v in openfoam_var.items() if k in ["u", "v", "nu"]},  
    batch_size=1024,  
)  
ldc_domain.add_validator(openfoam_validator)
```

## Running the Solver and Visualizing Results

### Solver Setup

```
# Create solver  
slv = Solver(cfg, ldc_domain)  
  
# Run solver  
slv.solve()
```

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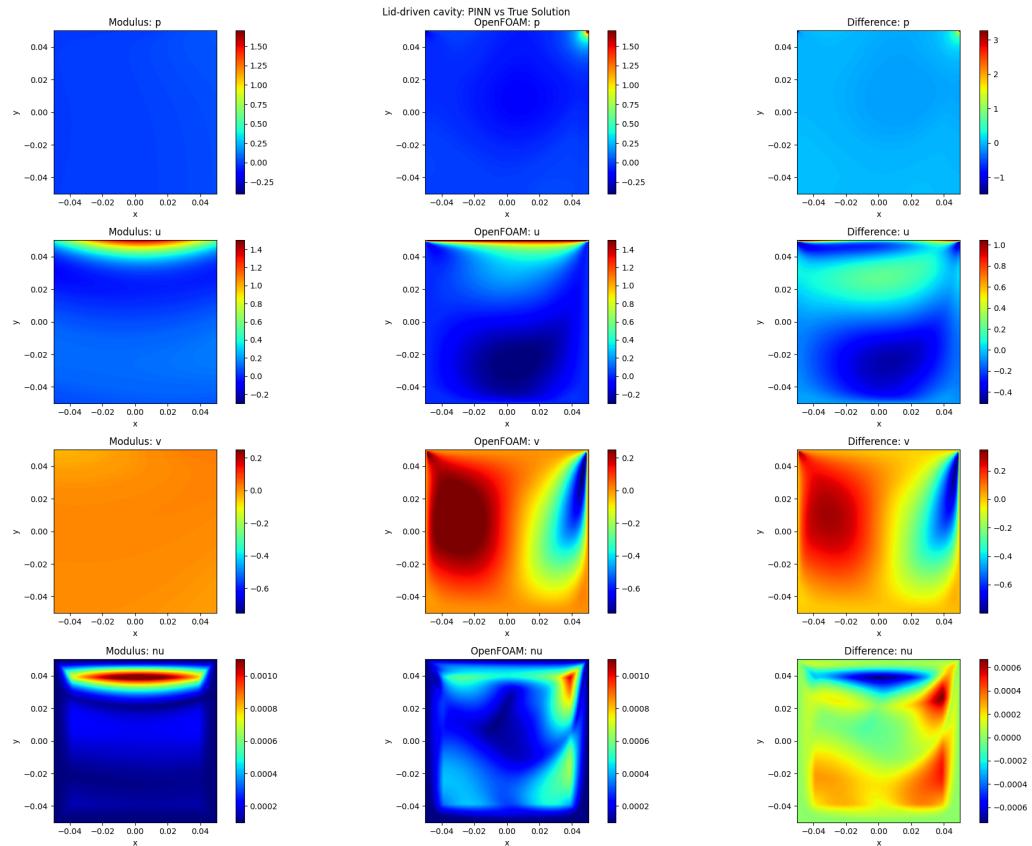
### Part 3: Results Visualization

<https://github.com/AI-ME-Ben/NVIDIA-Modulus-Reproduce/tree/main>

- **Velocity and Pressure fields** are visualized from the inference domain.
- **Comparison with OpenFOAM** simulation results.

[https://catalog.ngc.nvidia.com/orgs/nvidia/teams/modulus/resources/modulus\\_sym\\_examples\\_supplemental\\_materials](https://catalog.ngc.nvidia.com/orgs/nvidia/teams/modulus/resources/modulus_sym_examples_supplemental_materials)

rec\_validation\_freq: 100



## Summary

- Defined the Zero Equation Turbulence Model in Modulus Sym.
- Set up the Lid-Driven Cavity Flow problem with  $\text{Re} = 1000$ .
- Implemented boundary conditions, domain constraints, and validation.
- Ran the solver and compared with OpenFOAM results.