

# LG-Beam Amplitude Prediction

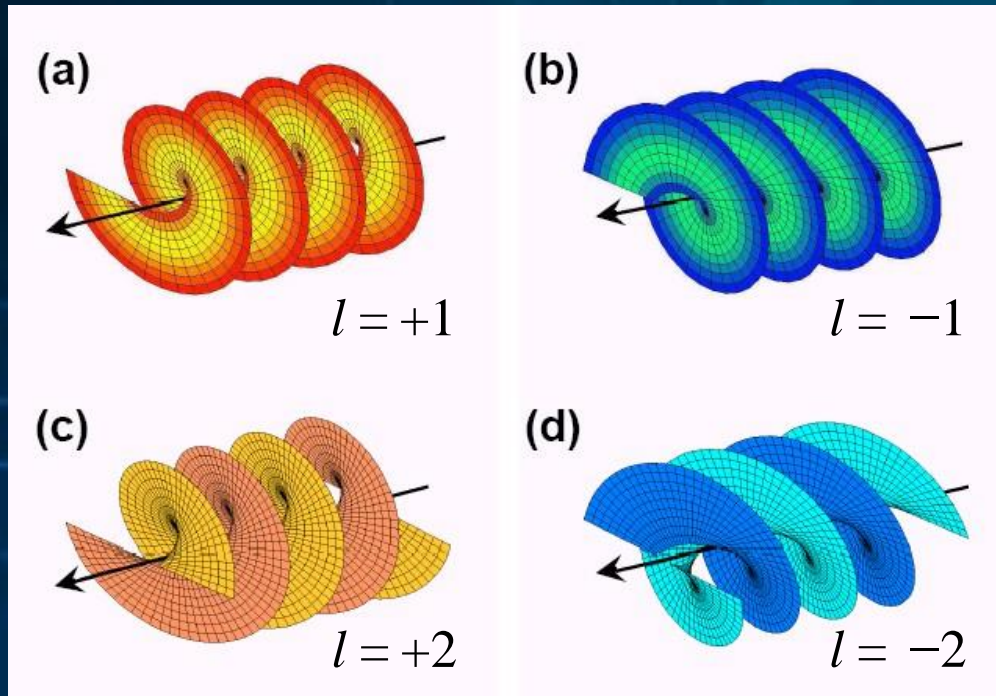
PARAG SHARMA

# Outline

- Orbital angular momentum of light
- The Mathematics of OAM
- Representation of L-G modes
- States in OAM Basis
- Superposition of LG Beams: Vortex Beams
- Machine learning for predicting complex amplitudes of superposition of two LG beams

# Orbital angular momentum of light ( OAM)

- The simplest example of a light beam carrying OAM is one with a phase in the transverse plane of  $\varphi(r, \varphi) = \exp(il\varphi)$
- helically phased beams carried an OAM equivalent to a value of  $l\hbar$  per photon



$$\mathbf{E}(\mathbf{r}, t) = \mathbf{E}_0(r, z) e^{im\varphi} e^{i(kz - \omega t)}$$

helical phase factor:

$$e^{il\varphi}$$

$$l = 0, \pm 1, \pm 2, \pm 3 \dots$$

Angular momentum (OAM):

$$L_z = l\hbar \quad \text{per photon}$$

**Helical modes:** (using cylindrical coordinates  $r, \varphi, z$ )

# The Mathematics of OAM

- Mathematically we can represent OAM with laguerre guassian modes
- Laguerre-Gaussian (LG) modes are solutions to paraxial Helmholtz Equation
- these modes have amplitude distributions, given by

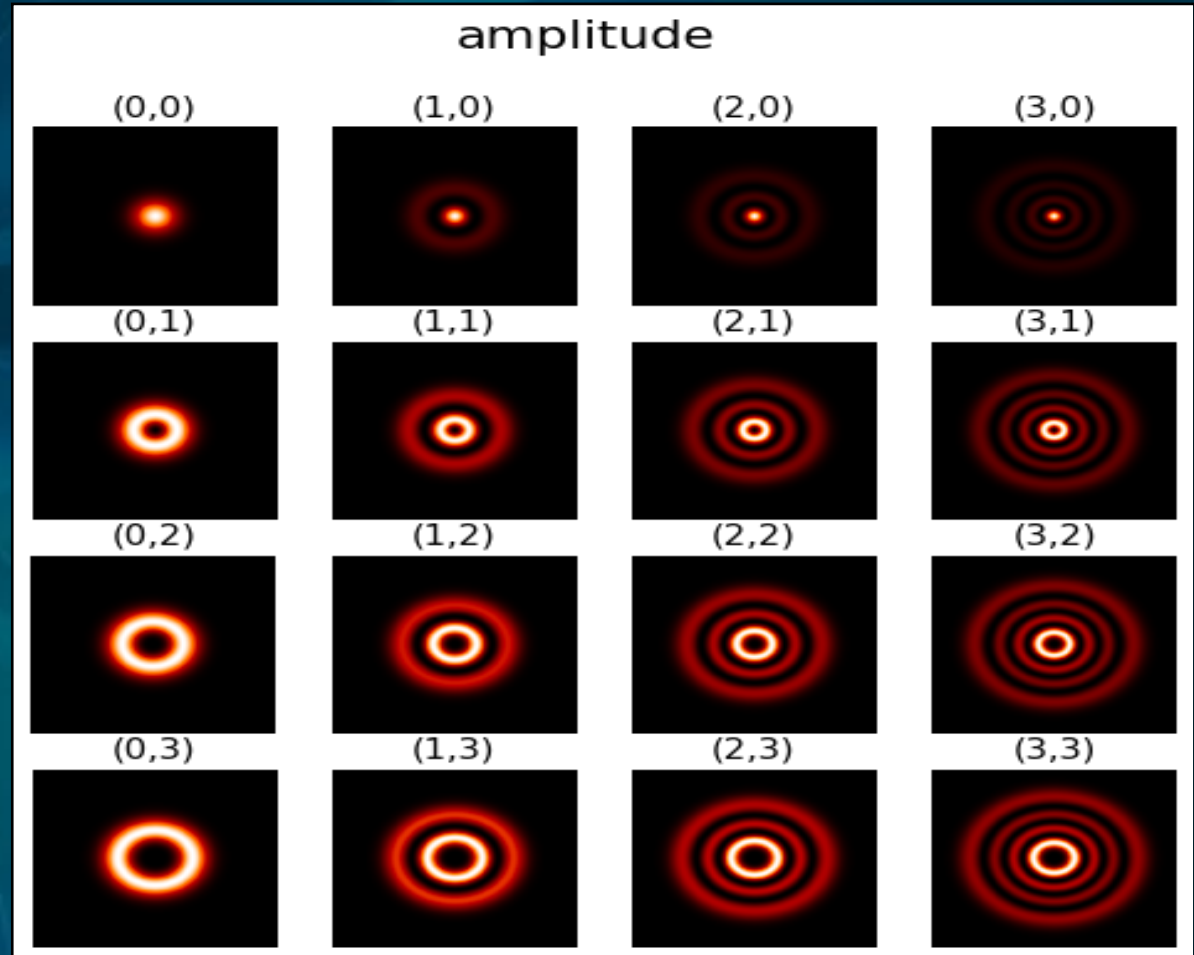
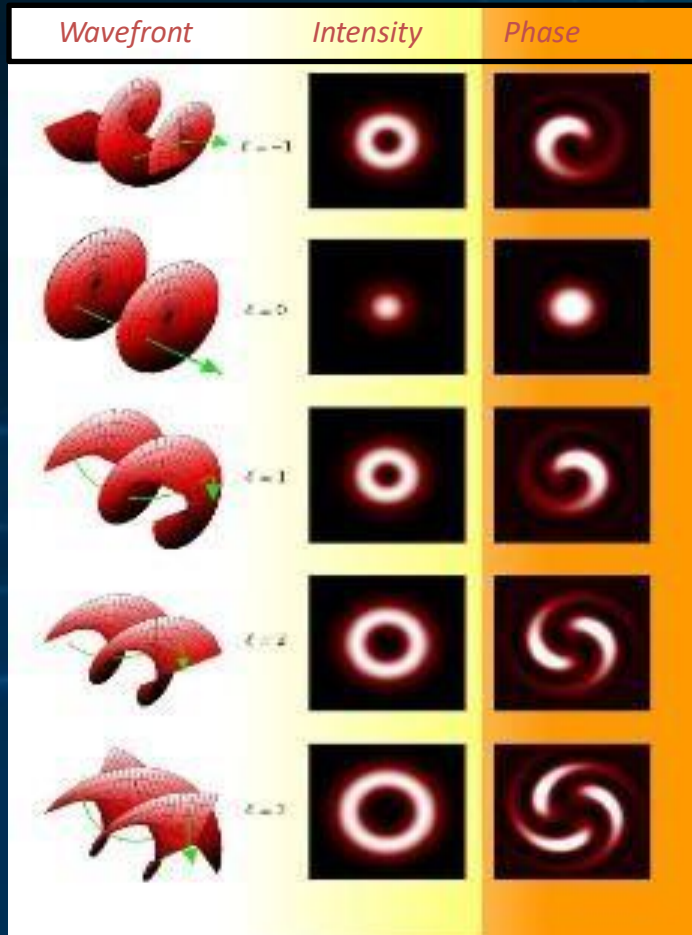
$$u_{pl}(r, \phi, z) = \frac{C}{(1 + z^2/z_R^2)^{1/2}} \left[ \frac{r\sqrt{2}}{w(z)} \right]^l L_p^l \left[ \frac{2r^2}{w^2(z)} \right] \exp \left[ \frac{-r^2}{w^2(z)} \right] \times \\ \times \exp \left[ \frac{-ikr^2 z}{2(z^2 + z_R^2)} \right] \exp(-il\phi) \exp \left[ i(2p + l + 1) \tan^{-1} \frac{z}{z_R} \right]$$

- The mathematical representation in terms of Laguerre - Gauss modes contains two integer numbers: l is the azimuthal index giving an OAM of  $l\hbar$  per photon, and p is the number of radial nodes in the intensity distribution



# Graphical representation of L-G modes

- The wavefront has a *helical shape* composed by  $\ell$  lobes disposed around the propagation axis  $z$ .



# States in OAM Basis

- it is useful to express most beams in a complete basis set of orthogonal modes. For OAM carrying beams this is most usually the Laguerre–Gaussian (LG) mode set
- Any image can be synthesized from the appropriate superposition of modes that form a complete basis set
- Laguerre-Gaussian (LG) modes form a complete set
- A photon that is in an LG eigenmode can be represented as  $|l, p\rangle$ . We can prepare an OAM state in the space spanned by  $|l, p\rangle$  basis and measure it using the polar coordinate basis

$$\sum_{p=0}^{\infty} \sum_{l=-\infty}^{\infty} \rho LG_p^{*l}(\rho, \phi, z) LG_p^l(\rho', \phi', z) = \delta(\phi - \phi') \delta(\rho - \rho')$$

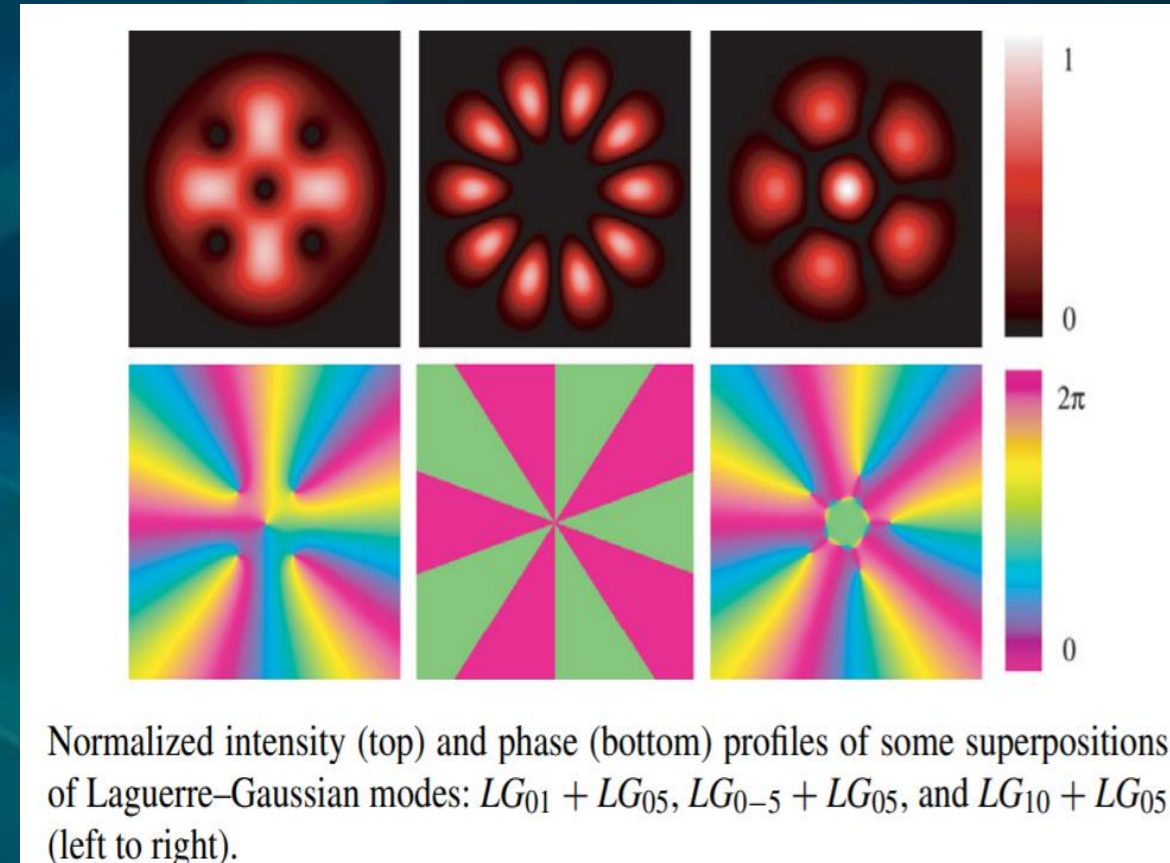
$$\int_0^{\infty} \int_0^{2\pi} \rho d\rho d\phi LG_p^{*l}(\rho, \phi, z) LG_{p'}^{l'}(\rho, \phi, z) = \delta_{ll'} \delta_{pp'}$$

$$LG_p^l(\rho, \phi, z) \rightarrow \psi_l(\phi) = \frac{1}{\sqrt{2\pi}} e^{-il\phi} \equiv \langle \phi | l \rangle$$

$$\sum_{l=-\infty}^{\infty} |l\rangle \langle l'| = \delta(\phi - \phi') \quad \text{with} \quad \langle l | l' \rangle = \delta_{ll'}$$

# Superposition of LG Beams: Vortex Beams

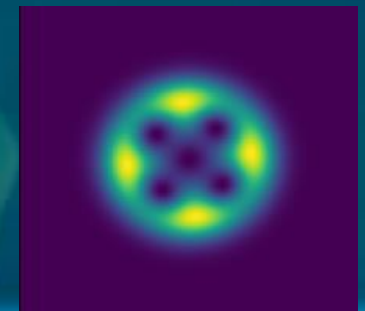
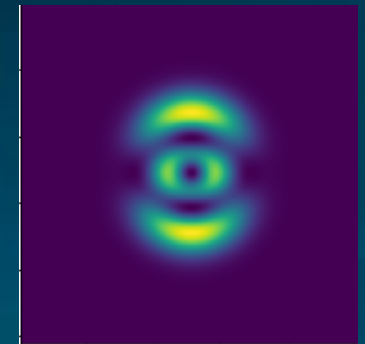
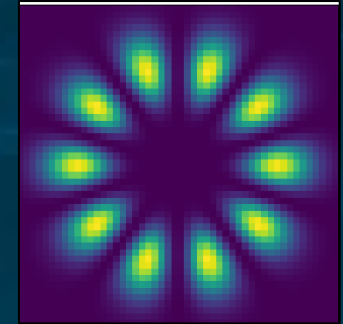
- When Laguerre–Gaussian beams interfere they produce vortex structures resulting in complex patterns of bright and dark regions
- By superposing two co-propagating LG beams with OAM  $l_1$  and  $l_2$  at  $z = 0$ , the beams interfere constructively at  $|l_2 - l_1|$  azimuthal positions, separated by regions of destructive interference and constructive interference, resulting in a transverse intensity profile of  $|l_2 - l_1|$  dark petals (zeroth intensity regions) or bright petals (high intensity regions)





# Machine learning for predicting complex amplitudes of superposition of two LG beams

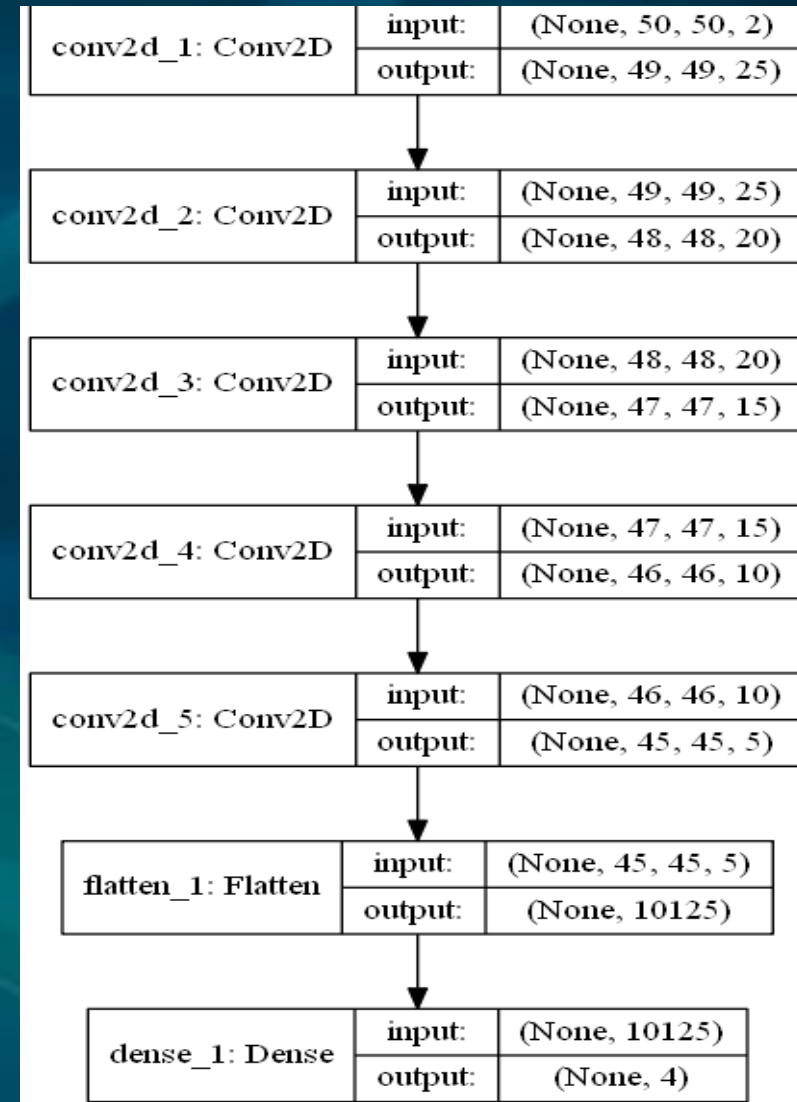
- We want to find the values of the complex coefficients for the constituent LG-beam from the superposed value  $a |l_1, p_1\rangle + b |l_2, p_2\rangle$ , here  $a$  and  $b$  can take any arbitrary complex value and  $l_1$  and  $l_2$  are the OAM values of two superposed LG beams
- This is a regression problem in which we need to find values of  $a$  and  $b$  using the matrix for superposed amplitude for two LG beams
- We can use deep convolutional neural network for this purpose
- First step is to generate data by the simulation of LG beams





# Model and training parameters

- To train the model we generate around 3000 superposed matrices with random complex coefficients a and b whose real and imaginary parts are chosen randomly from uniform distribution from -1 to 1
- The input consist of the matrix with dimensions 50x50x2 since each point is represented by a complex number
- The model consists of 5 2d convolution layers with filter size 25,20,15,10 and 5, and 1 densely connected layer with 4 neurons as the output
- The loss function is the mean squared error
- The model is trained on 70% data with learning rate of 0.01 for 30 epochs
- Different combinations for l and p were used to generate different dataset for training



# Training Results for different combinations of l and p values

