

Analysis of Turbine Wake Characteristics using Proper Orthogonal Decomposition and Triple Decomposition Methods

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- Motivation and Objectives
- Experiment setup
- Predicting velocity deficit in the wake
- Triple Decomposition (TD)
- Proper Orthogonal Decomposition (POD)
- Future work (DMD)



# **Motivation and Objectives**

- The study of wake behind a wind turbine is crucial in determining the wind farm layouts.
- The study focuses on understanding the momentum recovery process through quantitative and qualitative analysis.



Image courtesy:

http://nstergiannis.blogspot.com/



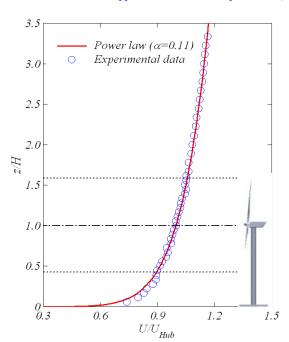
#### Simulation of Atmospheric Boundary Layer (ABL) Winds

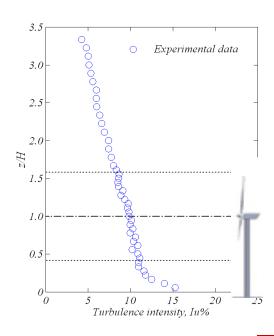


Terrain Category	Terrain description	Gradient height, Z <sub>G</sub> (m)	Roughness length, $Z_O$ (m)	Wind Speed exponent, α
1	Open sea, ice, tundra desert	250	0.001	0.11
2	Open country with low scrub or scattered trees	300	0.03	0.15
3	Suburban area, small towns, well wooded areas	400	0.3	0.25
4	Tall buildings, city centers, well developed industrial areas	500	3.0	0.36

#### Offshore wind farm (9.5% at hub height)

Architecture Institute of Japan

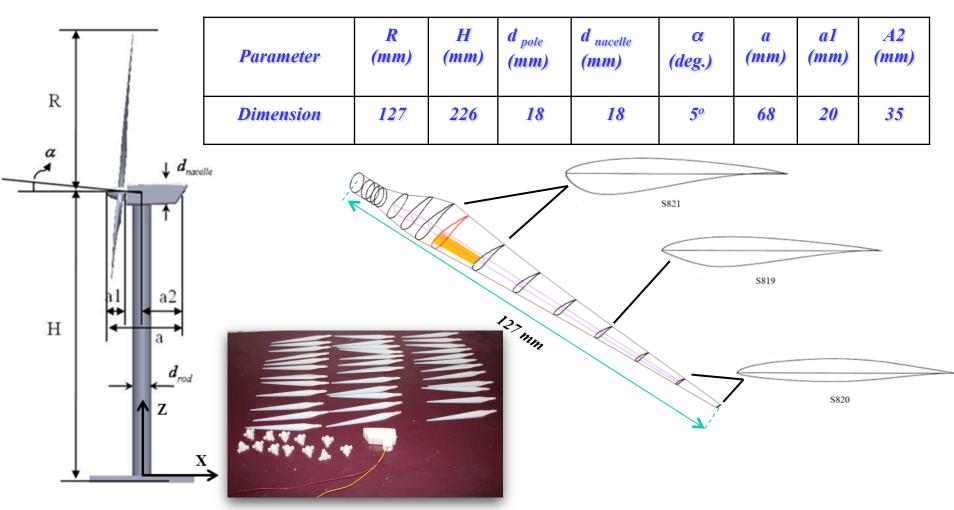






Hansen & Barthelmie, Wind Energ.2012;15:183–196

#### **Wind Turbine Models**



1:350 scaled model to simulate a 2MW wind turbine

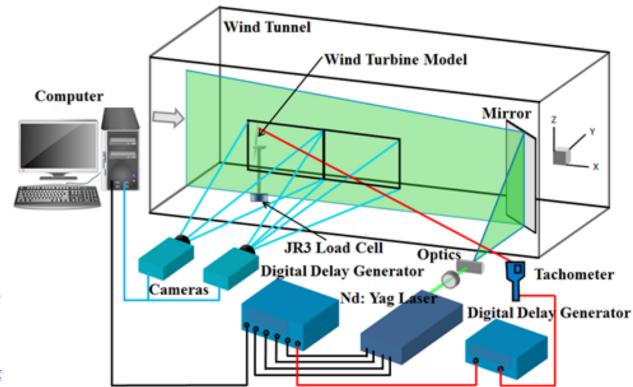
ERS-100 prototype of wind turbine blade developed by TPI



# **Experimental Setup**

#### **Test conditions:**

- Velocity at hub height  $U_{Hub} = 4.5 \text{ m/s}$
- Chord Reynolds number, Re≈ 7,000
- Rotation speed of wind turbine:
- $\varpi = 0 \sim 2200 \text{ rpm}$
- Tip-speed-ratio,  $\lambda = 0 \sim 6.0$

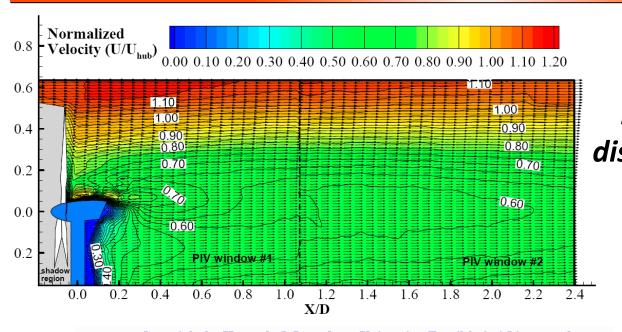


#### **Measured parameters:**

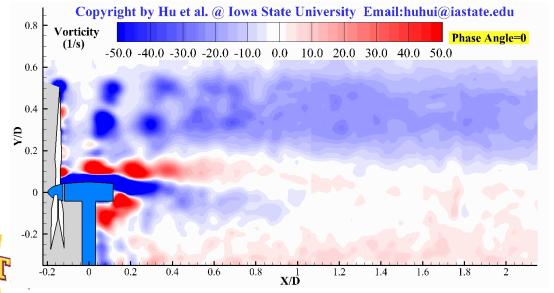
- Wake flow field distributions
- 1000 snapshots for free run case
- 300 snapshots for phase lock case



### **Velocity and Vorticity Distributions**



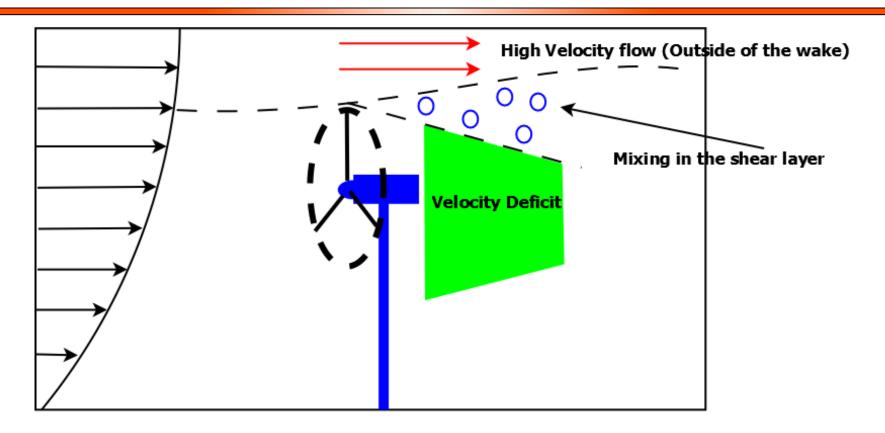
Normalized Velocity distribution of ensemble average



Vorticity distribution of Phase Locked

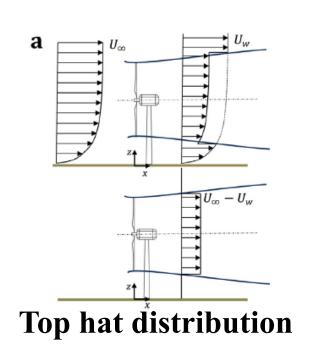


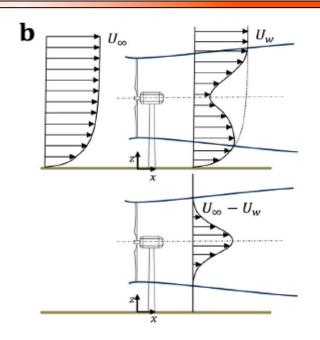
## Wake Aerodynamics



- Velocity deficit results in a shear layer
- Turbulent mixing can be observed
- High turbulence mixing shortens momentum recovery distance

### **Analytical Wake Models**





Gaussian distribution

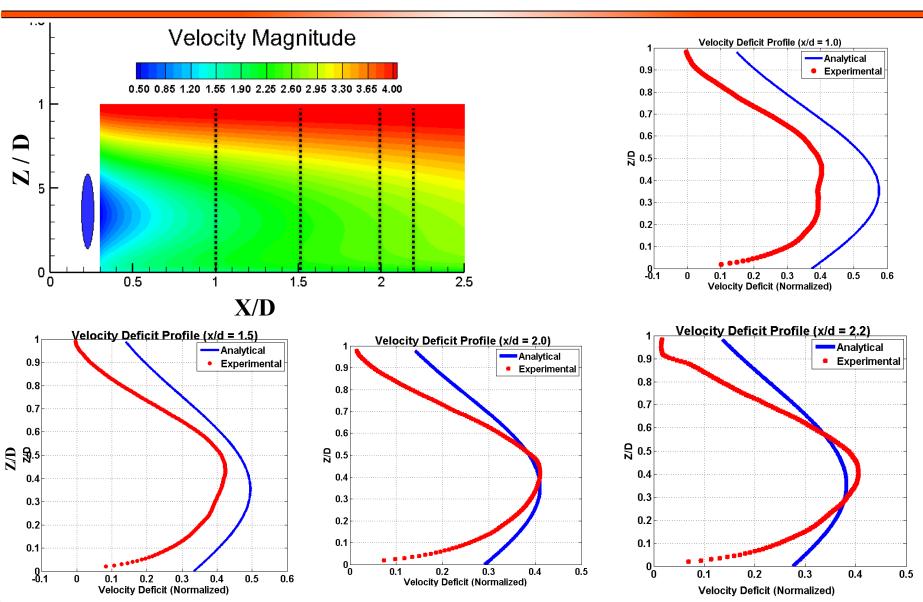
$$\Delta U/U \downarrow \infty = (1 - \sqrt{1 - C \downarrow T} / 8 (k \uparrow * x/d \downarrow 0)$$

$$+0.2 \sqrt{\beta}) \uparrow 2 \quad ) \times \exp(-1/2 (k \uparrow * x/d \downarrow 0)$$

$$+0.2 \sqrt{\beta}) \uparrow 2 \quad \{(z-z \downarrow h/d \downarrow 0) \uparrow 2 + (y/d \downarrow 0) \uparrow 2 \}\}$$



# **Predicting Velocity Deficit**

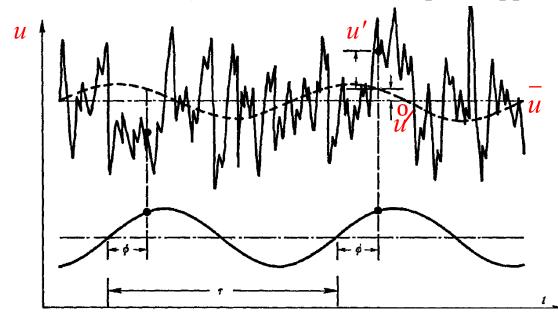




#### Phase-locked PIV measurement results in turbine wake

#### **Triple decomposition method:**

Hussain (J. Fluid Mech. vol. 41, part 2, pp. 241-258, 1970)



$$u = u + u' + u'$$

*u* is instantaneous velocity, the right terms correspond to mean value, contribution of organized wave and random fluctuating respectively.

Then the phase-averaged velocity is obtained:

$$\langle u \rangle = u + u = u - u$$

 $\langle u \rangle = \overline{u} + \overline{u} = u - u'$ The phase averaging process is equivalent to a low-pass filter which keeps the mean flow and the organized wave at a single frequency and removes the other higher frequency fluctuations.



#### Phase-locked PIV measurement results-Reynolds shear stress

The Reynolds stress contribute a major part of the momentum transfer in turbine wake flow

Triple decomposition method to Reynolds stress:

$$r \downarrow ij /-\rho = (u + u \uparrow')(v + v') = u v + u'v + u v' + u \uparrow' v'$$

Where: 
$$u' = \langle u \rangle - \overline{u}$$
  $u' = u - \langle u \rangle$ 

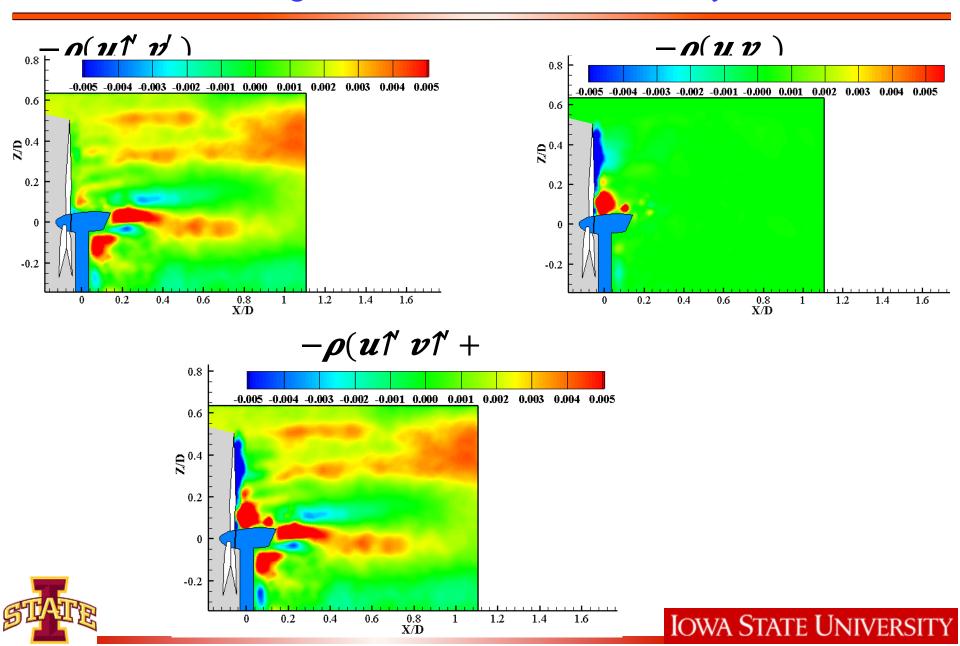
Additionally, coherent structures are uncorrelated with random parts:

$$u'v = 0 uvt' = 0$$
So: 
$$r lij / -\rho = uv + ut' v'$$

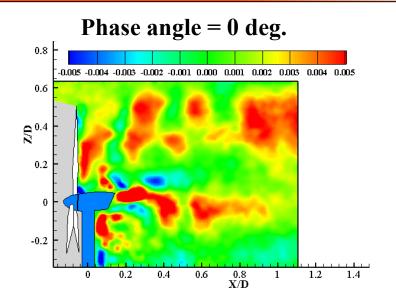
Therefore the overall Reynolds stress consists of the contributions from the coherent structures uv and random fluctuating uv

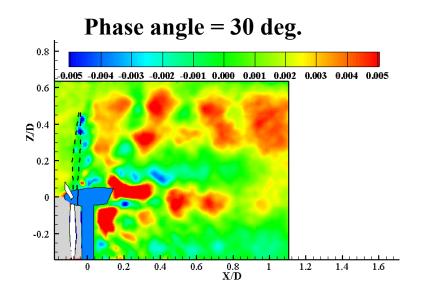


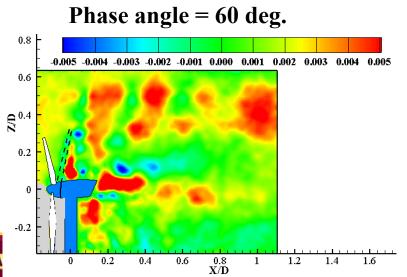
#### Ensemble Averaged PIV measurement results-Reynolds stress

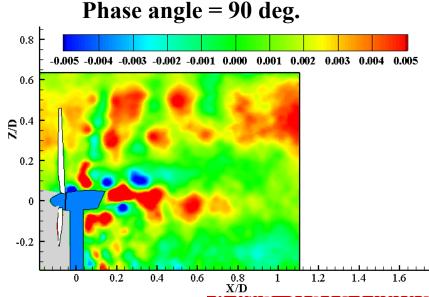


### Decomposed Reynolds Stress $(-\rho < u \land v \land >)$









# **POD** (Proper Orthogonal Decomposition)

- POD is used to extract dominant large scale structures buried in a flow.
- The flow is decomposed to modes using linear decomposition and reconstructed using singular value decomposition (SVD).
- The magnitude of the eigenvalues represent the kinetic energy which the modes are ranked based on.
- Large scale structures can be represented with the first few modes.



## **Method of Snapshots**

(1) Arrange all of fluctuating velocity components in a matrix U as:

$$\mathbf{U} = \begin{bmatrix} \mathbf{u}^{1} & \mathbf{u}^{2} & \cdots & \mathbf{u}^{N}_{1} \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{u}^{1}_{M} & \mathbf{u}^{2}_{M} & \cdots & \mathbf{u}^{N}_{M} \\ v_{1}^{1} & v_{1}^{2} & \cdots & v_{1}^{N} \\ \vdots & \vdots & \ddots & \vdots \\ v_{M}^{1} & v_{M}^{2} & \cdots & v_{M}^{N} \end{bmatrix}$$

where M is the number of spatial discrete points and N is the number of the PIV snapshots, which represent the spatial and temporal resolutions of the PIV data respectively.

(2) The eigenvalues and eigenvectors of the auto-covariance matrix are calculated as:

$$\tilde{\mathbf{C}} \cdot \mathbf{A}^{i} = \lambda^{i} \cdot \mathbf{A}^{i}$$
, with  $\tilde{\mathbf{C}} = \mathbf{U}^{T} \cdot \mathbf{U}$   
 $\lambda_{1} > \lambda_{2} > \dots > \lambda_{N} = 0$ 



Ma, Geisler, Agocs & Schroeder, "Investigation of coherent structures in active flow control over a backward-facing step by PIV", 2014. 16<sup>th</sup> International symposium on Flow Visualization, Okinawa, Japan.

#### Free run PIV results analysis-Proper orthogonal decomposition (POD)

(3) **Each mode** is obtained by projecting matrix U onto each eigenvector and then normalized by its norm as:

$$\phi^{i} = \frac{\sum_{n=1}^{N} (A_{n}^{i} \cdot \mathbf{u}^{n})}{\|\sum_{n=1}^{N} (A_{n}^{i} \cdot \mathbf{u}^{n})\|}, i = 1, \dots, N$$

So: 
$$\Phi = [\phi^1 \quad \phi^2 \quad \cdots \quad \phi^N]$$

#### **Reconstruct the snapshot:**

(4) The coefficients of each mode can be obtained:

$$\mathbf{a}^{n} = \mathbf{\Phi}^{T} \cdot \mathbf{u}^{n}$$

(5) A snapshot can be reconstructed as:

$$\mathbf{u}^{n} = \sum_{i=1}^{N} a_{i}^{n} \phi^{i} = \mathbf{\Phi} \cdot \mathbf{a}^{n}$$

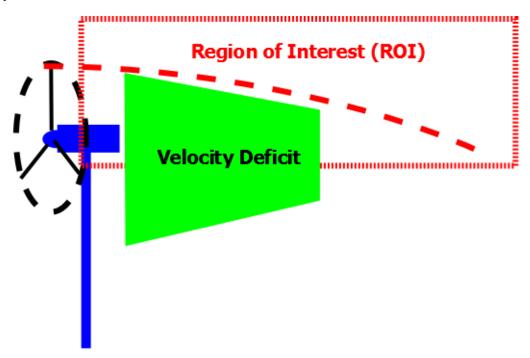
(6) An instantanous velocity fields can be reconstructed using the first Lth modes as:



$$U^n = \overline{U} + u^n = \overline{U} + \sum_{i=1}^{L} a_i^n \phi_i$$
 where  $\overline{U}$  is the ensemble-averaged velocity

# **Selecting a Region of Interest (ROI)**

- A region in the wake was selected for POD analysis to minimize the outliers and measurement noise.
- The region encompasses includes the shear layer, located between the wake and high momentum flow outside the wake.





# Free run PIV results analysis-Proper Orthogonal Decomposition (POD)

- 1. Each eigenvalue is proportional to its total kinetic energy of velocity fluctuations
- 2. The most important modes containing higher energy are in the first few modes, which correspond to large-scale flow structures

3. Other further modes containing lower energy are less dominant, sometime

**Energy %** 

1.0

Mode

Number **Modal Energy** 10° 41.7 1 7.4 4.0 3 Energy 101 3.4 2.3 5 2.2 1.6 8 1.3 1.2

Mode Number

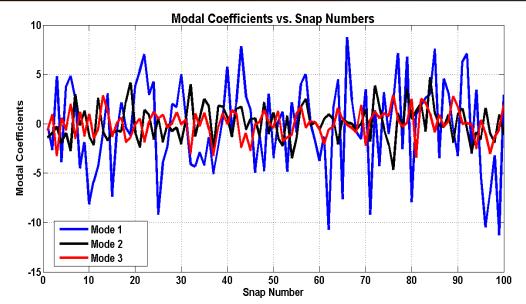
7 8 9 10

10

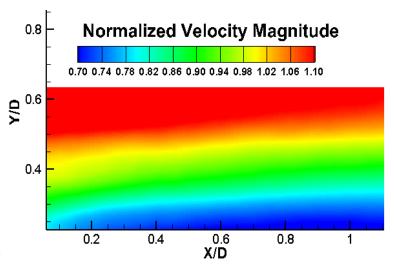
are merged into measurement noise.

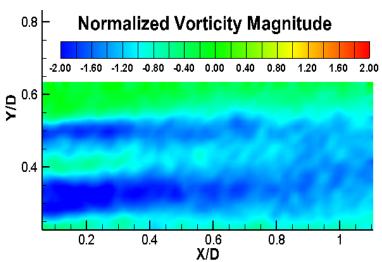
IOWA STATE UNIVERSITY

#### **Proper Orthogonal Decomposition (POD)**



Mode	Std. Dev	
1	4.31	
2	1.72	
3	1.37	

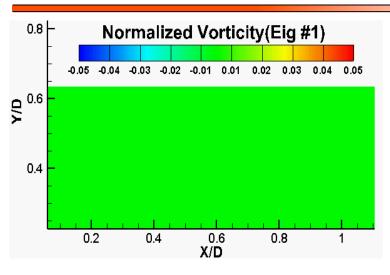




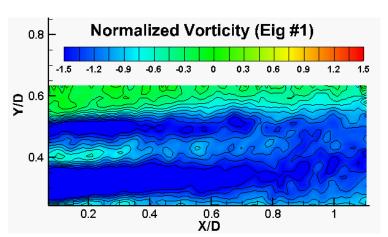
**Time Averaged Solutions (Mode 0)** 

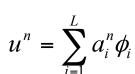
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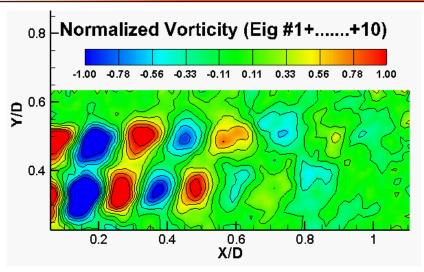
# Fluctuations and Velocity Reconstructions



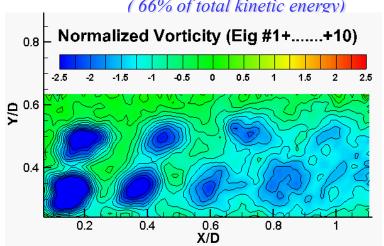
Reconstructed by using the first 1 mode (41% of total kinetic energy)

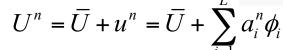






Reconstructed by using the first 5 mode (66% of total kinetic energy)



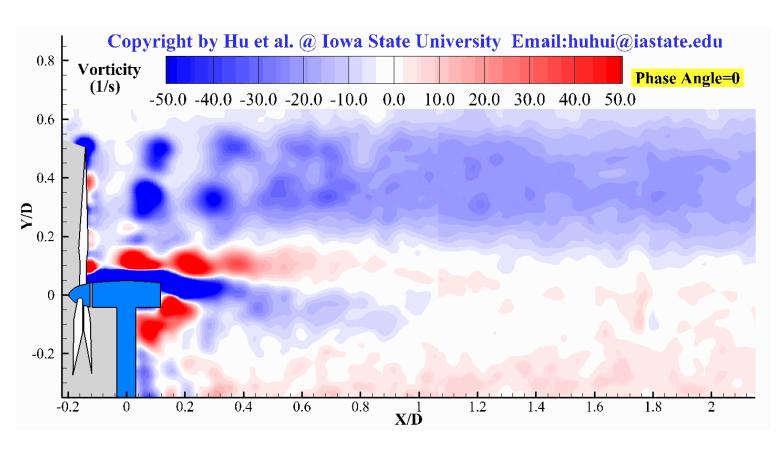


### **Conclusions**

- Proper Orthogonal Decomposition (POD) can be successfully used to identify principal components of the flow.
- The principal components carry most of the kinetic energy and the low order reconstructions aid the design process of blades, rotor assemblies and vortex generators
- Reynolds stress obtained from TD indicates vertical momentum transfer in the recovering wake.
- The wind farm designer can use this information to determine the distance between turbines for max. performance.



# Thank you for your attention!





# **APPENDIX**



# **Future Work - DMD**

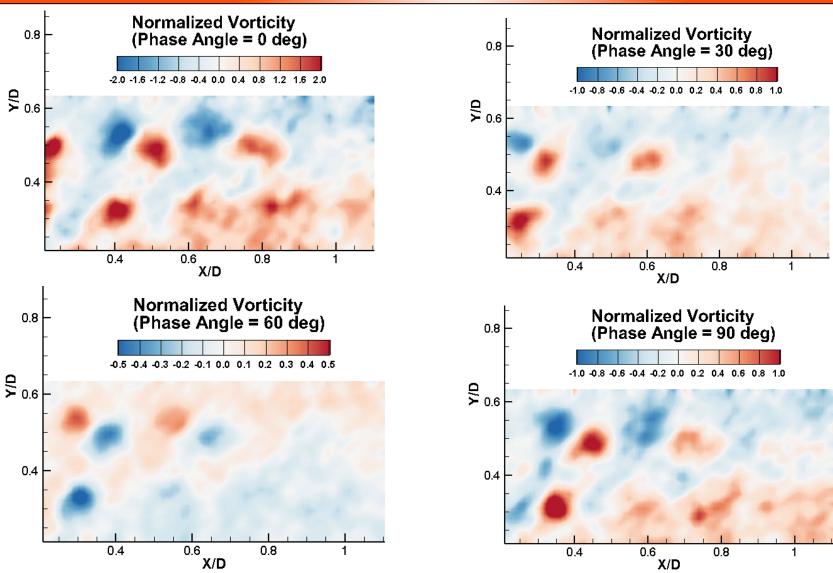
• Dynamic Mode Decomposition (DMD) is a new tool in dynamical systems that are used to investigate flow features of unsteady flows.

• DMD assigns frequencies to the large scale structures while POD modes are ranked based on kinetic energy of the flow.

• DMD is based on a temporal orthogonal framework while POD is based on a spatial orthogonal one.

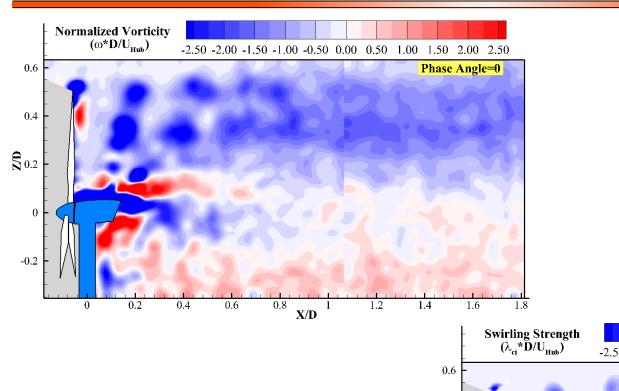


# POD reconstructions on phase locked data



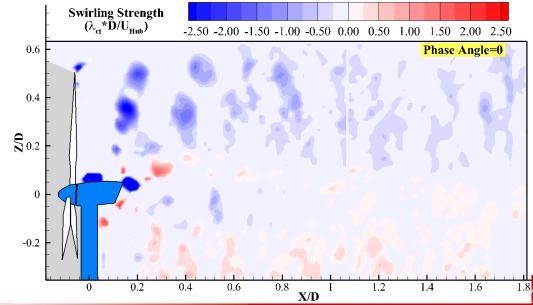


### Swriling strength distributions in the turbine wake



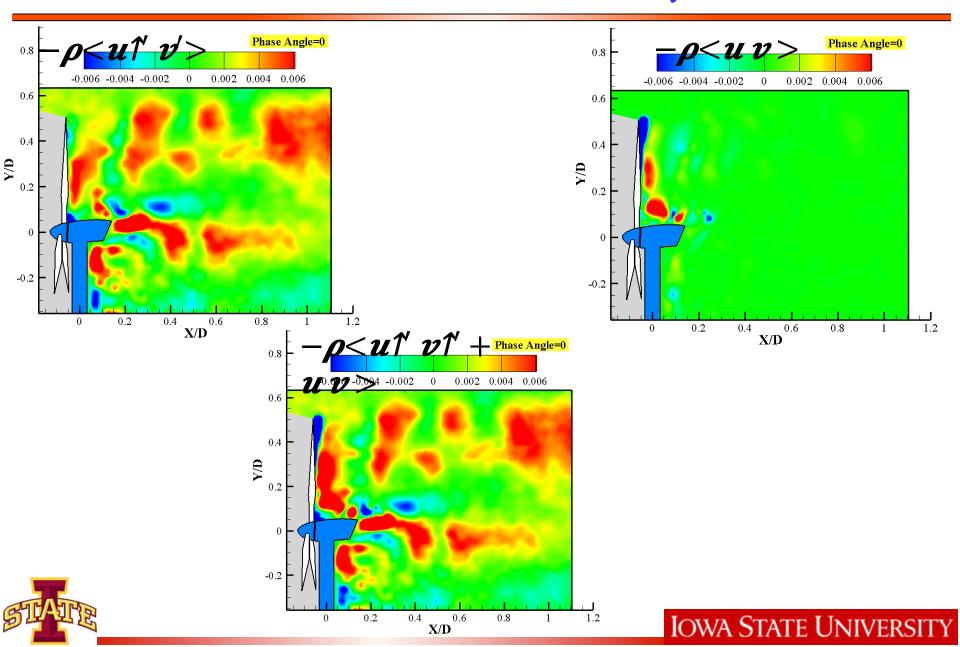
Low turbulence inflow

Phase angle=0

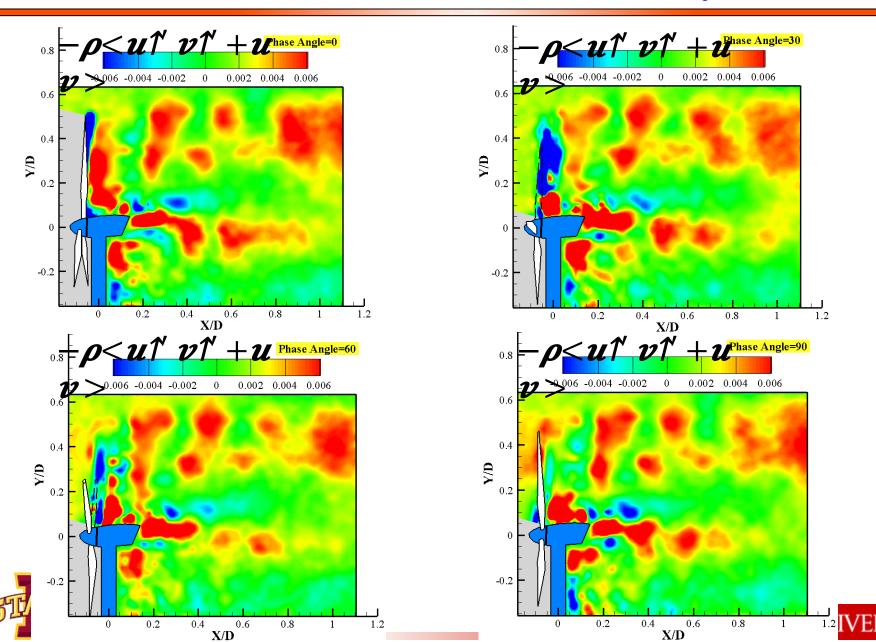




#### Phase-locked PIV measurement results-Reynolds stress



### Phase-locked PIV measurement results-Reynolds stress



# **Observations**

- Reynolds stresses are distributed within the shear layer.
- The shed vortices also contain turbulent mixing components.
- Presence of a tower and a nacelle also contributes to T.K.E and Reynolds Stress in downstream wake.

