

Evaluation of Fluid Thermal Systems by  
Dynamic Data Driven Application Systems

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# Outline

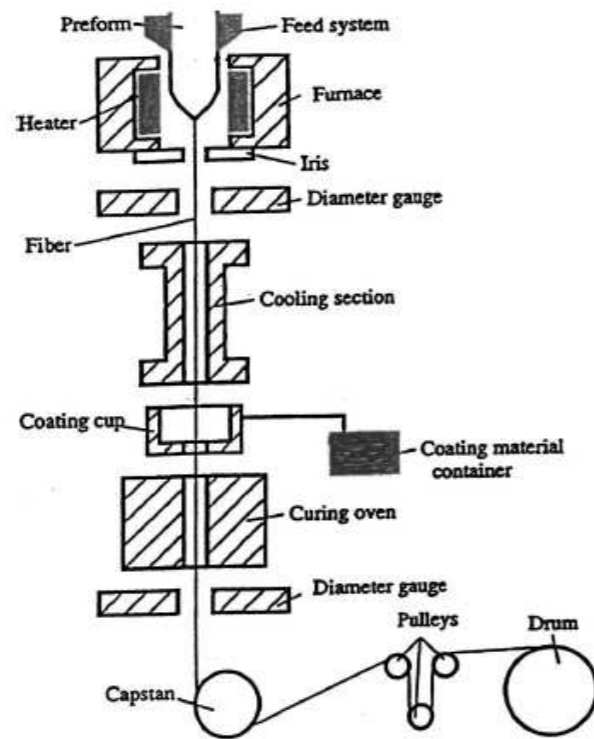
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- Introduction
- Problem Definition
- Dynamic Data Driven Applications System Methodology
- Results
- Current Focus of Research
- Conclusions

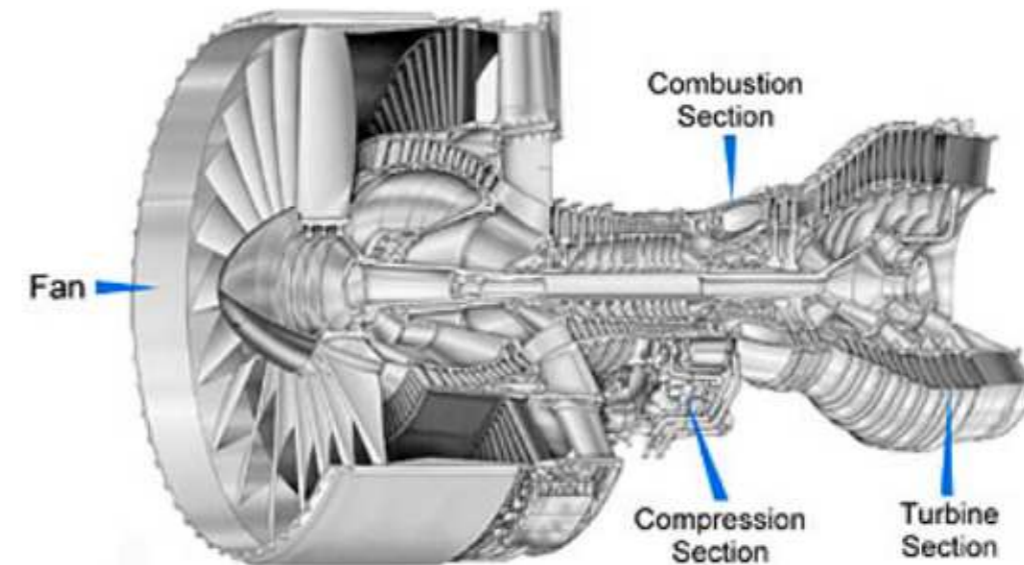
# Introduction

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- In many engineering applications involving fluid-thermal systems, detailed quantitative information on the flow, temperature and species concentration is needed for system optimization



Optical fibre furnace

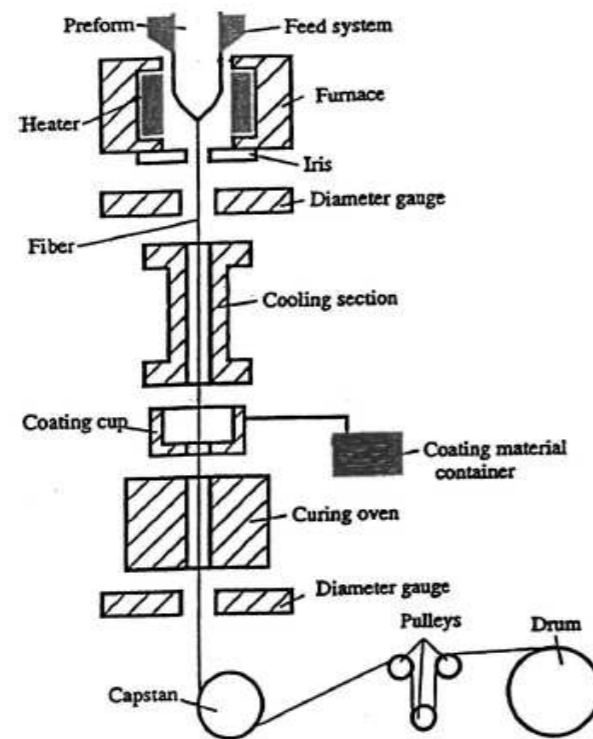


Turbofan engine

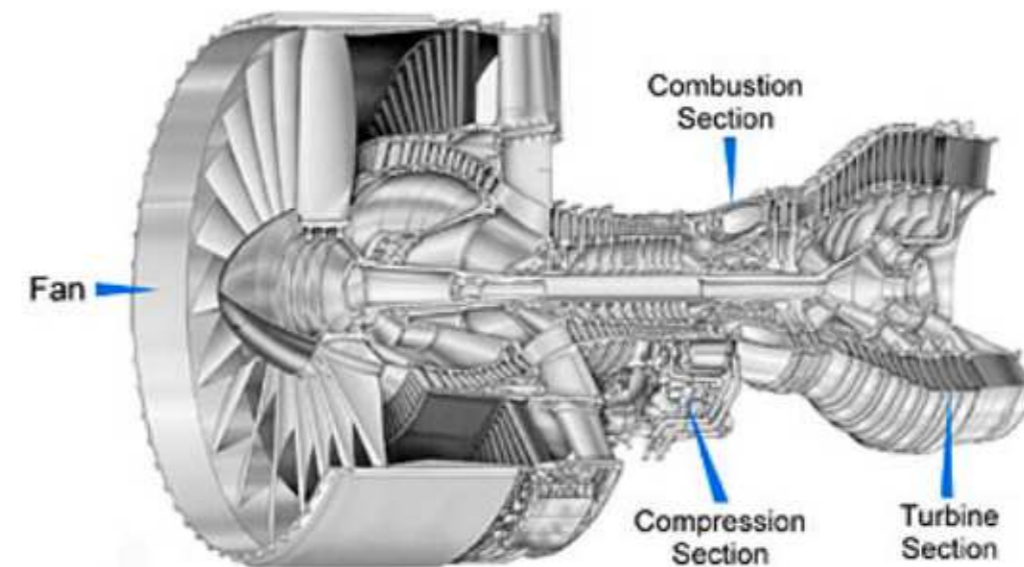
# Introduction

- Numerical simulation can obtain the desired information and thus optimize the system

However, this approach requires well-defined boundary and operating conditions which may not be completely known due to limited access for experimental measurements



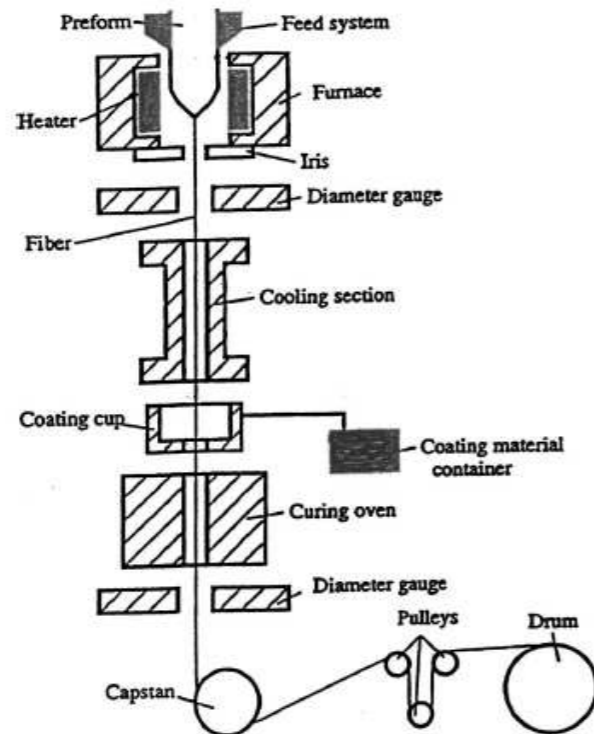
Optical fibre furnace



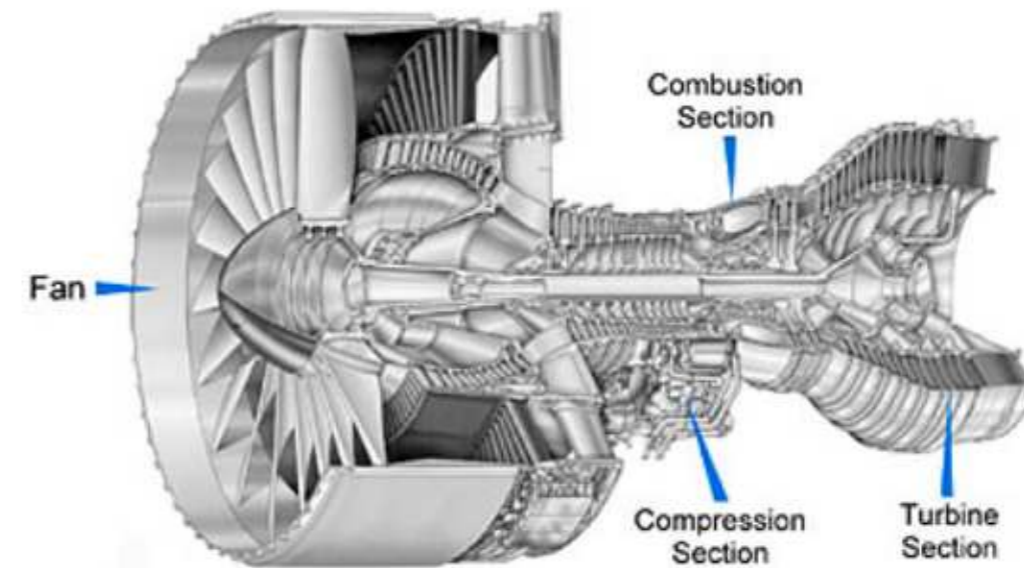
Turbofan engine

# Introduction

- The objective of our research is to develop a Dynamic Data Driven Applications System approach that synergizes experiment and simulation to determine the boundary and operating conditions, thereby achieving a full simulation capability



Optical fibre furnace



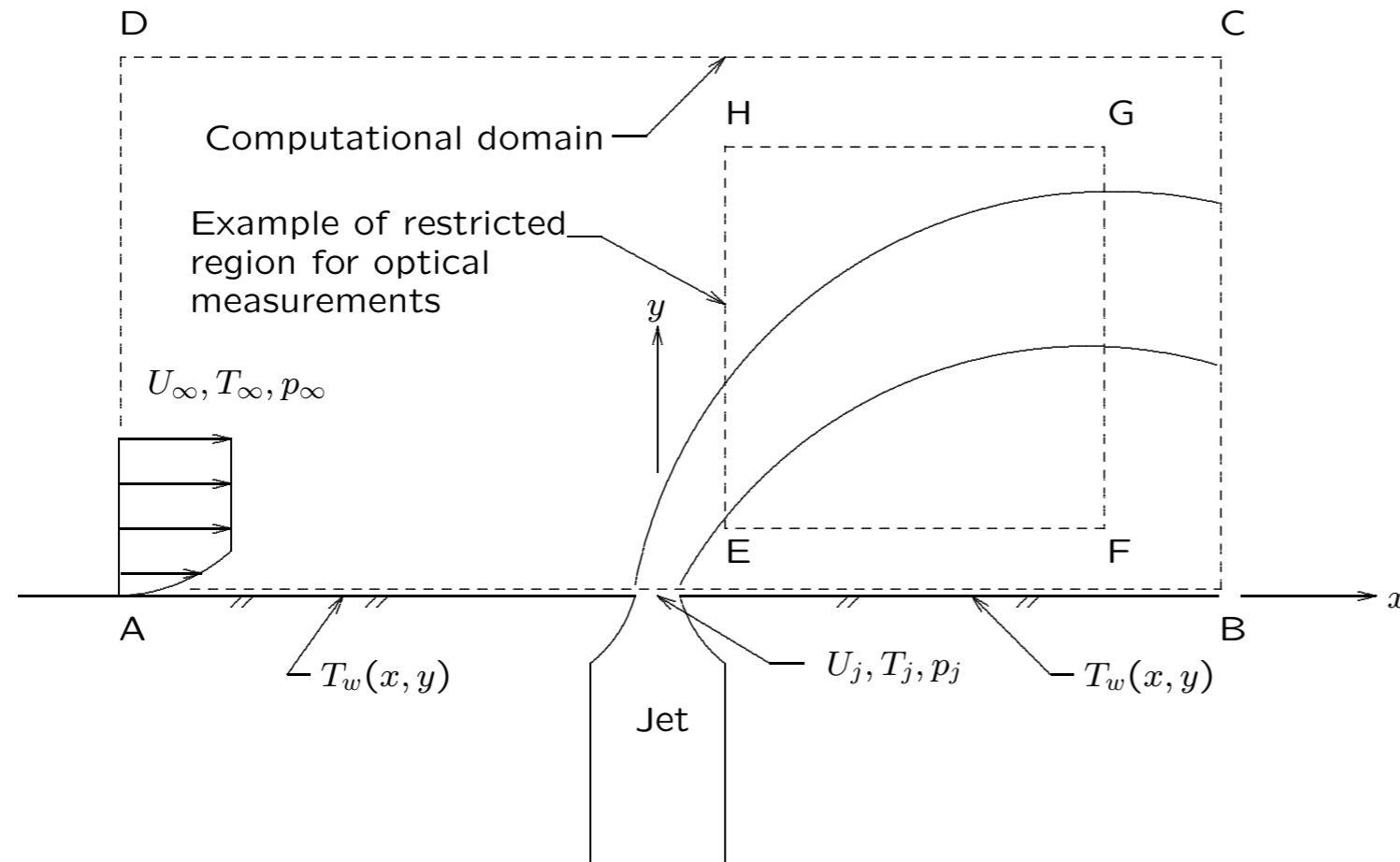
Turbofan engine

# Problem Definition

## Jet in Crossflow

- Heated wall jet in crossflow

The objective is to determine the jet inflow conditions ( $U_j$ ,  $T_j$  and  $p_j$ ) using a Dynamic Data Driven Applications Systems method that synergizes experiment and simulation



Item	Parameters	
	Known	Unknown
$U_\infty$	✓	
$T_\infty$	✓	
$p_\infty$	✓	
$U_j$		✓
$T_j$		✓
$p_j$		✓

# Problem Definition

## Jet in Crossflow

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- Experiment

Rutgers Low Speed Wind Tunnel

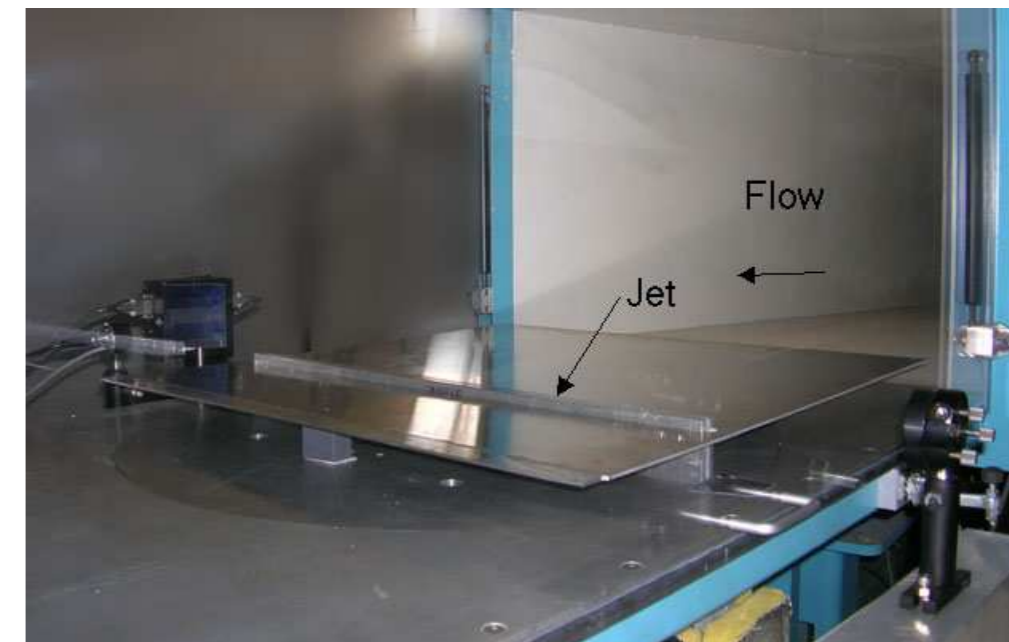
Non-intrusive laser diode measurement

Measure absorbance vs time at fixed  $(x, y)$

Static temperature  $T$  vs time from absorbance

Limited region for absorbance measurement

Each  $(x, y)$  measurement requires  $\approx 1$  hr



Experimental configuration

# Problem Definition

## Jet in Crossflow

- Laser diode absorbance

Instantaneous absorbance

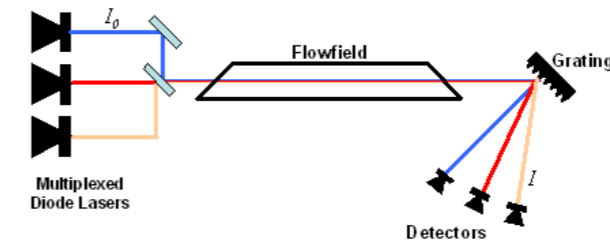
$$A(x, y) = \frac{(I_o - I(x, y, t))}{I_o}$$

where  $I_o$  is incident intensity at  $(x, y, z_1)$  and  $I(x, y, t)$  is transmitted intensity at  $(x, y, z_2)$

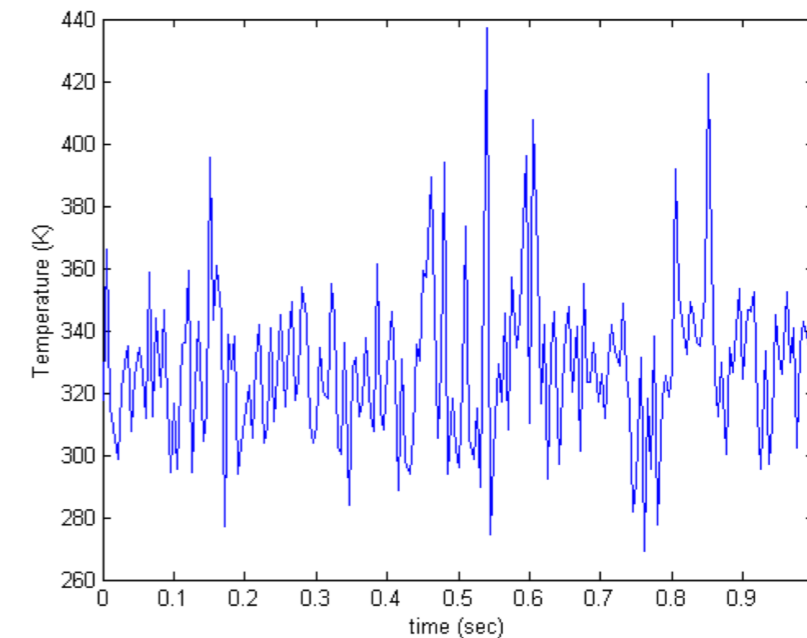
Absorbance per cm of the  $^Q R_2(6)$  line of the oxygen transition  $b_1 \Sigma_g^+ \nu' = 0 \leftarrow X^3 \Sigma_g^- \nu'' = 0$  at 761.139 nm is

$$\frac{dA}{dz} = 0.083 T^{-1} - 2.26 \cdot 10^{-5}$$

where  $T(x, y, z, t)$  is the static temperature in K



Laser diode arrangement



Typical  $T$  vs time



# Problem Definition

## Jet in Crossflow

- Simulation

Laminar Navier-Stokes equations

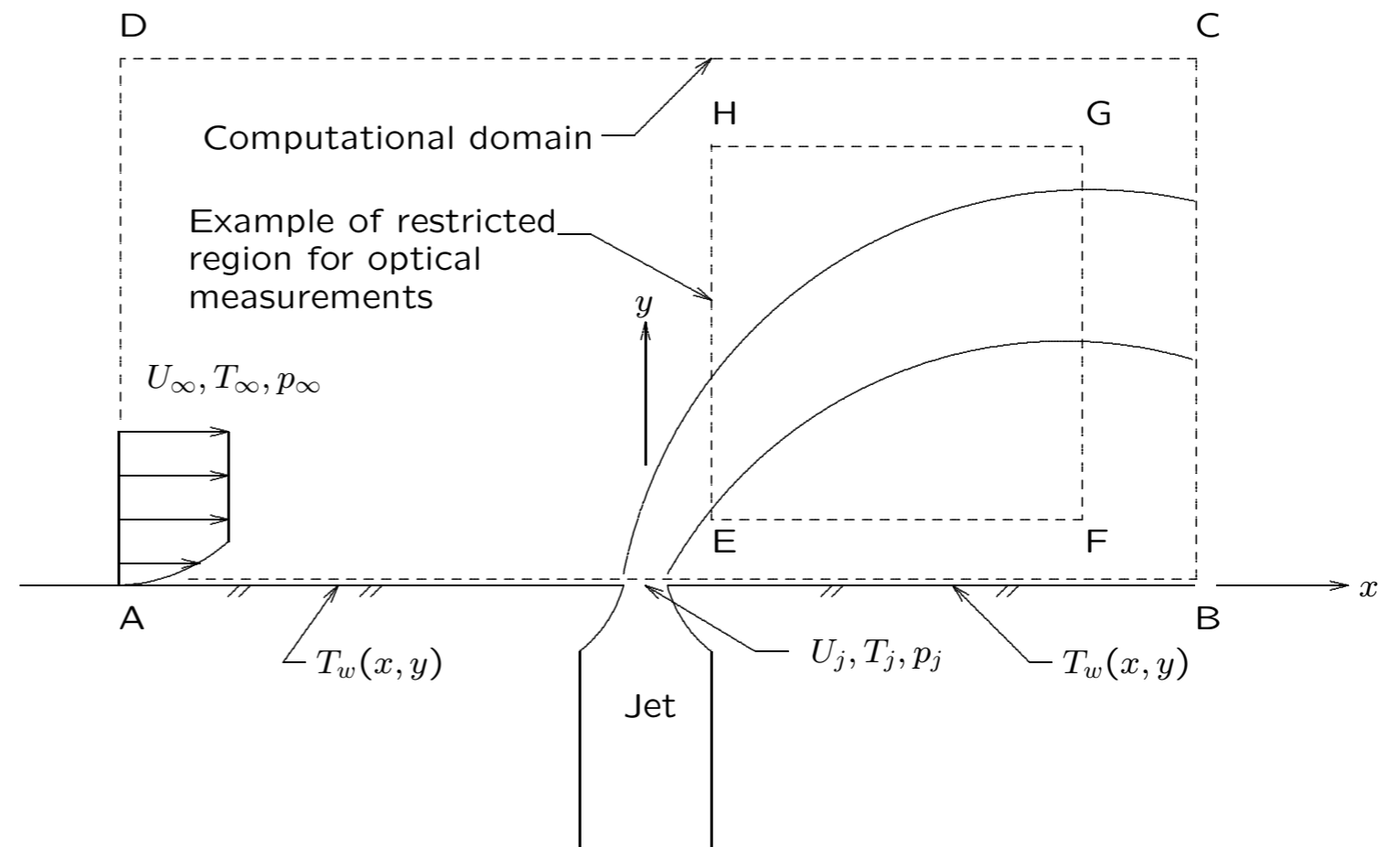
Incompressible, ideal gas

Unsteady, time-dependent

Sutherland viscosity law

Fluent<sup>©</sup>

Parallel (8 processors)



# Problem Definition

## Jet in Crossflow

- Assumptions

Large set  $\mathcal{D}$  of discrete data locations defined ( $\leq$  no. of grid cells in simulation)

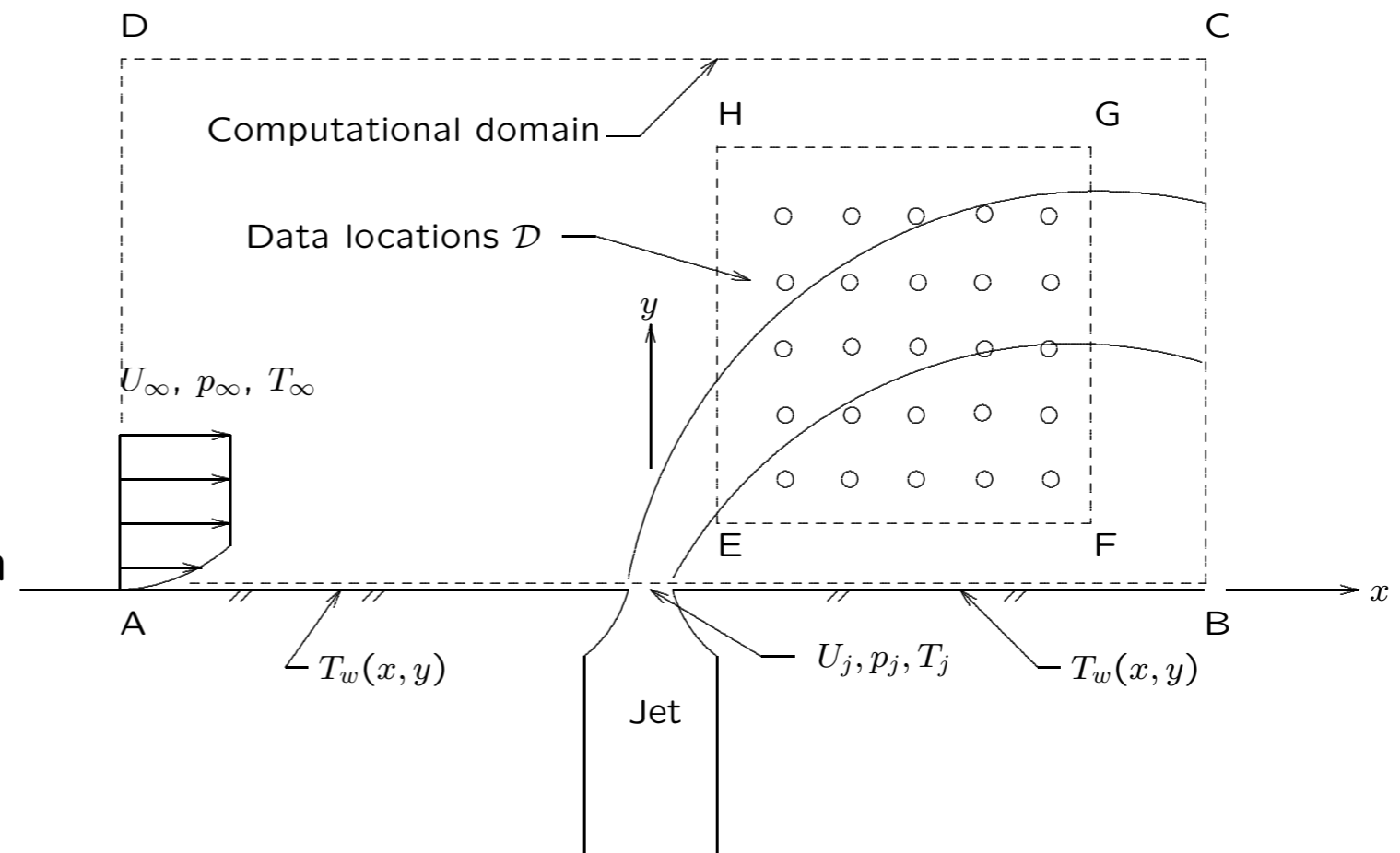
For each experiment  $\mathcal{E}_\nu$ , time series data obtained for small subset  $\mathcal{D}_\nu$  of locations

For each simulation  $\mathcal{S}_\nu$ , time series data obtained for entire set  $\mathcal{D}$  for  $U_{j\nu}, p_{j\nu}, T_{j\nu}$

- The quantity for comparison between experiment and simulation is the mean square temperature fluctuations  $\overline{T'^2}$

- Problem

What is the optimum sequence of simulations and experiments to determine  $U_j, T_j, p_j$  ?



# Dynamic Data Driven Applications System Methodology

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- Algorithm

Build initial Response Surface models  $\mathcal{M}_1$  for  $\overline{T'^2}$  on  $\mathcal{D}$  using simulation assumed values for  $U_j, p_j, T_j$

Select  $\mathcal{D}_1$  randomly from  $\mathcal{D}$

loop in  $\nu$  beginning with  $\nu = 1$  until converged

{

    Perform experiment  $\mathcal{E}_\nu$  and obtain  $\overline{T'^2}$  on  $\mathcal{D}_\nu$

    Use Response Surface models  $\mathcal{M}_\nu$  for  $\overline{T'^2}$  on  $\mathcal{D}_\nu$  to predict  $U_{j_{\nu+1}}, p_{j_{\nu+1}}, T_{j_{\nu+1}}$  based on experiment  $\mathcal{E}_\nu$

    Perform simulation  $\mathcal{S}_{\nu+1}$  using  $U_{j_{\nu+1}}, p_{j_{\nu+1}}, T_{j_{\nu+1}}$  and update Response Surface Models  $\mathcal{M}_{\nu+1}$

    Select  $\mathcal{D}_{\nu+1}$  based upon largest values of  $\overline{T'^2}$  on  $\mathcal{D}$  from Response Surface Models  $\mathcal{M}_{\nu+1}$

}

# Results

## Flowfield Structure

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- The flowfield is inherently unsteady

<i>Item</i>	<i>Value</i>
$U_{\infty}$ (m/s)	4
$T_{\infty}$ (K)	290
$p_{\infty}$ (atm)	1
$U_j$ (m/s)	8
$T_j$ (K)	370
$p_j$ (atm)	1

Temperature contours

# Results

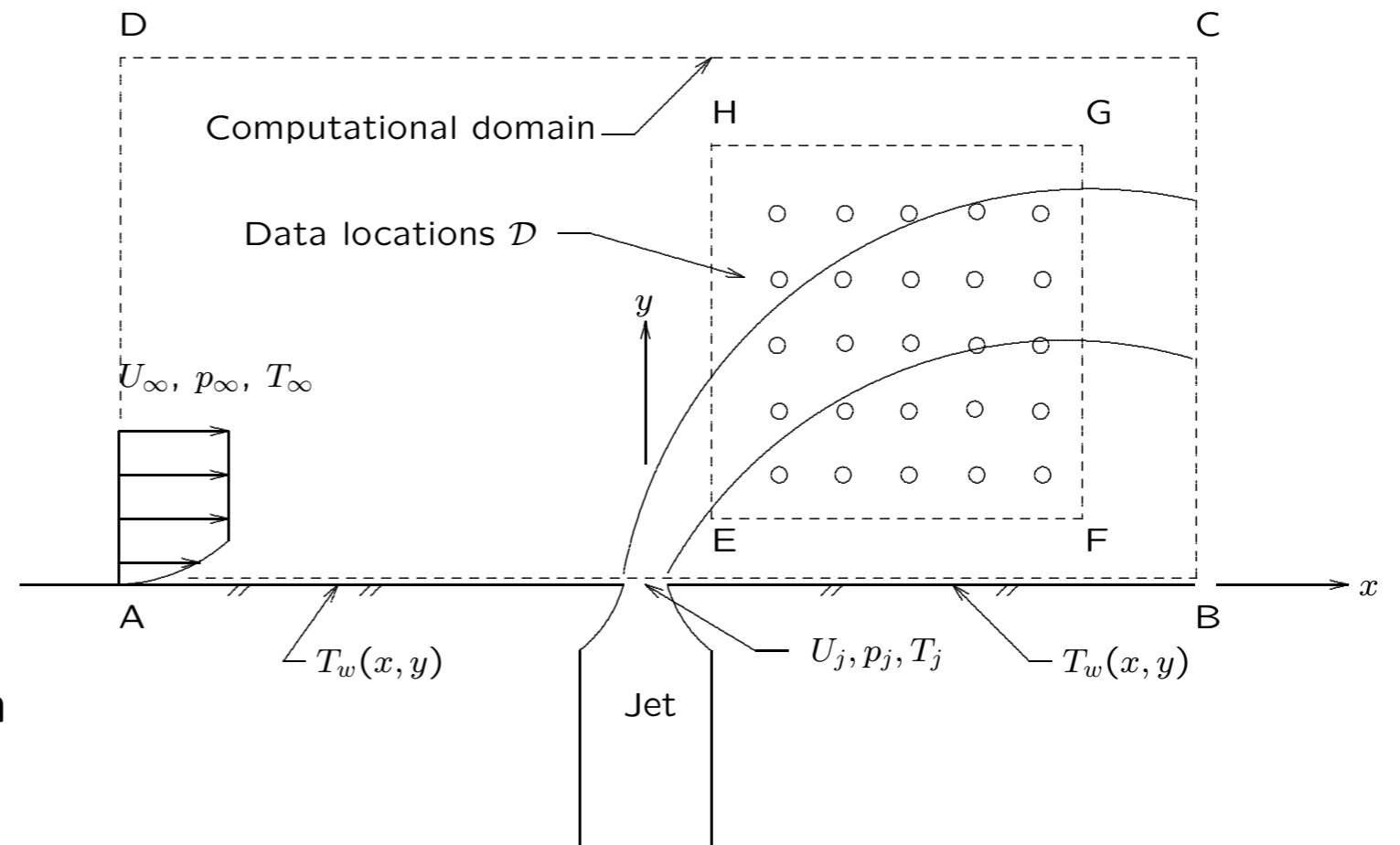
## Simple Test of Algorithm

- Pseudo-experiment performed to generate  $T$  vs time at 18 locations

Parameters	
Item	Value
$U_\infty$ (m/s)	4
$T_\infty$ (K)	290
$p_\infty$ (atm)	1
$U_j$ (m/s)	8
$T_j$ (K)	"unknown"
$p_j$ (atm)	1

- Questions

- What is the effect of the signal level on accuracy of Response Surface model ?
- What is the required duration for the time series ?



# Results

## Simple Test of Algorithm

- Step No. 1

The initial values of  $T_j$  for Response Surface selected at random from  $290 < T_j < 500$

The initial locations for “measurement” selected at random from matrix of 18 locations (3 horizontal rows of 6 points each)

Case	$T_j$ (K)	$\overline{T'^2}$ (K <sup>2</sup> )			
		Location 7	Location 10	Location 12	Location 13
Experiment No. 1	“unknown”	392.6	286.8	207.4	120.1
1	405	762.0	543.7	358.5	112.7
2	356	305.2	142.2	104.7	115.6
$T_j _{\text{predicted}}$		365.3	373.6	375.8	→ 281.1
Average $T_j _{\text{predicted}}$	349 K				
Std Dev $T_j _{\text{predicted}}$	45.4 K				

# Results

## Simple Test of Algorithm

- Step No. 1 (cont'd)

Case	$T_j$ (K)	$\overline{T'^2}$ (K <sup>2</sup> )			
		Location 7	Location 10	Location 12	Location 13
Experiment No. 1	"unknown"	392.6	286.8	207.4	120.1
1	405	762.0	543.7	358.5	112.7
2	356	305.2	142.2	104.7	115.6
$T_j _{\text{predicted}}$		365.3	373.6	375.8	→ 281.1

Average  $T_j|_{\text{predicted}}$  349 K

Std Dev  $T_j|_{\text{predicted}}$  45.4 K

Revised  $T_j|_{\text{predicted}}$  = 372 discarding values that are more than one standard deviation from mean

The revised locations for experimental measurement based upon the highest values of  $\overline{T'^2}$  from Cases 1 and 2 are Locations 1, 2, 7 and 8

# Results

## Simple Test of Algorithm

- Step No. 2

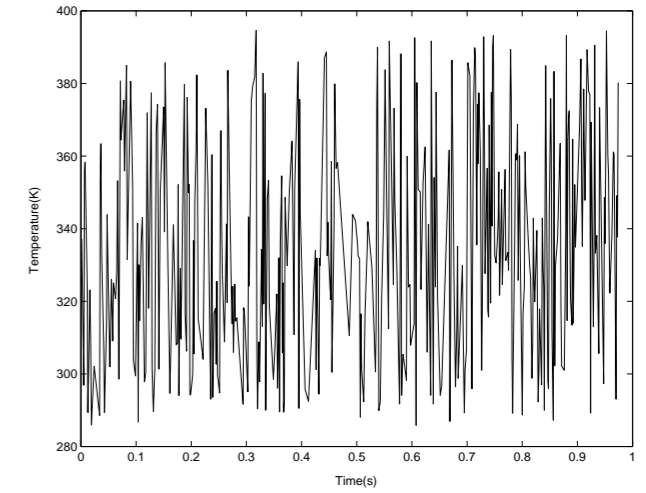
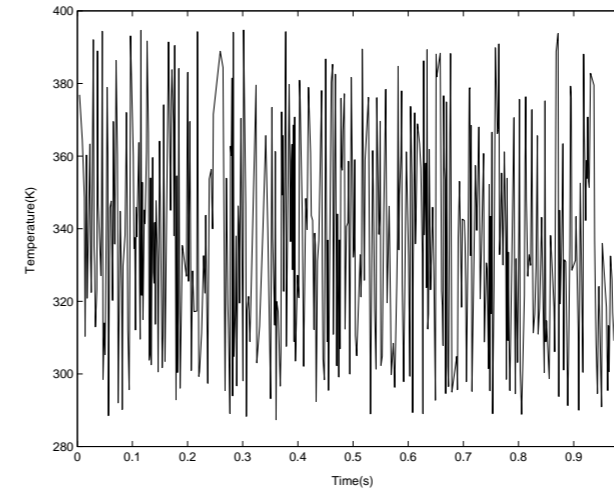
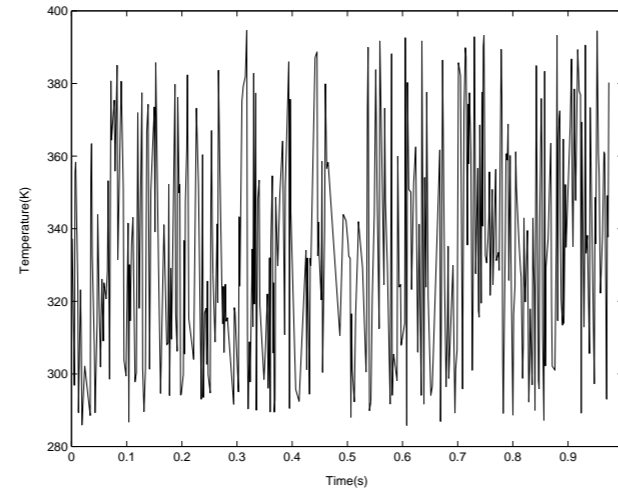
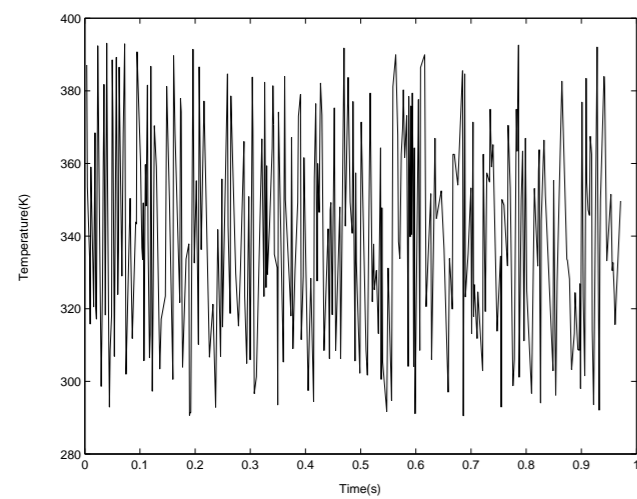
Case	$T_j$ (K)	$\overline{T'^2}$ (K <sup>2</sup> )			
		Location 1	Location 2	Location 7	Location 8
Experiment No. 2	"unknown"	408.4	272.1	392.6	381.7
1	405	1035.2	641.8	762.0	617.1
2	356	303.6	224.3	305.2	224.0
3	367	358.4	253.1	390.8	303.9
$T_j _{\text{predicted}}$		376.0	367.2	373.7	377.2
Average $T_j _{\text{predicted}}$	373.5 K				
Std Dev $T_j _{\text{predicted}}$	4.4 K				
Experiment	370 K				



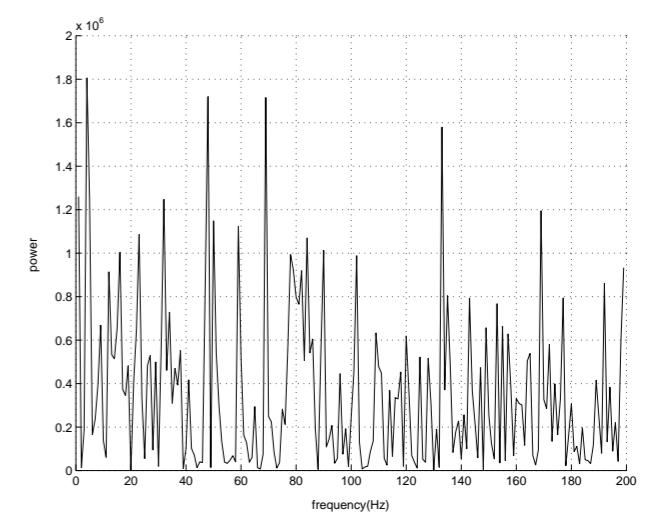
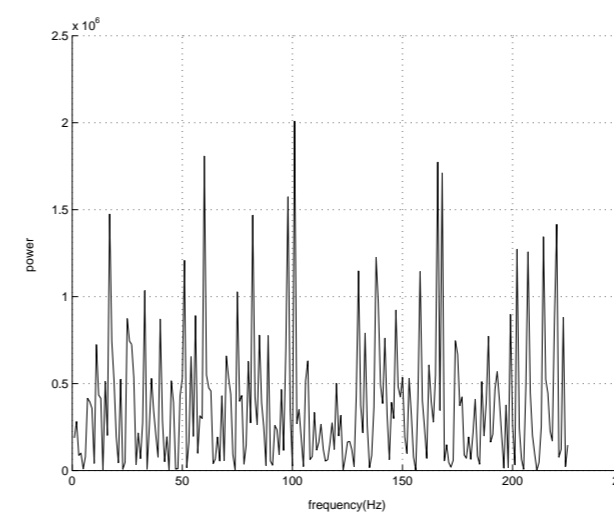
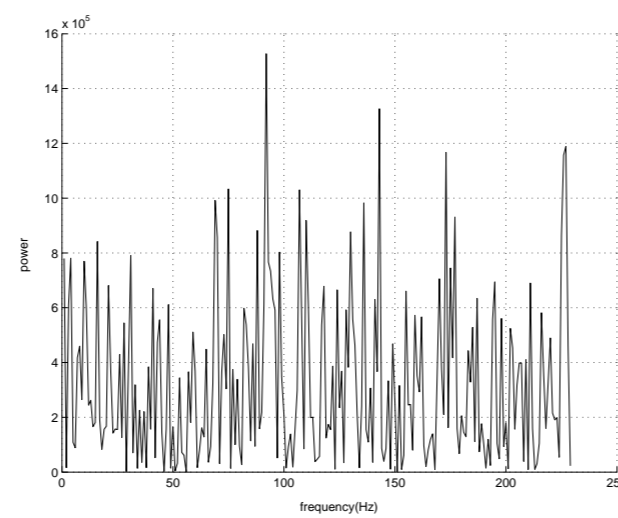
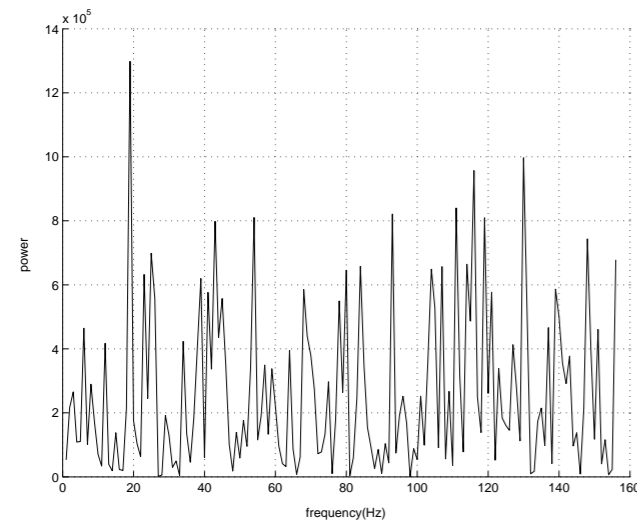
# Results

## Typical Experimental Results

### Temperature vs time



### Power spectrum of temperature fluctuations



$$(x, y) = (0.218\text{m}, 0.03\text{m})$$

$$(x, y) = (0.228\text{m}, 0.03\text{m})$$

$$(x, y) = (0.238\text{m}, 0.03\text{m})$$

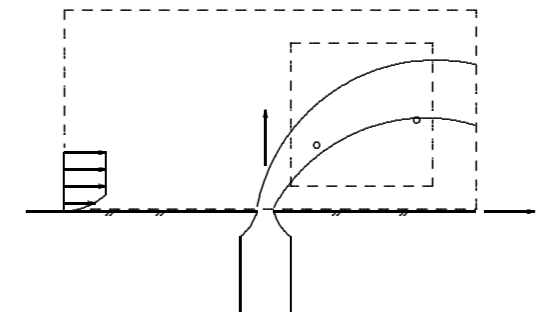
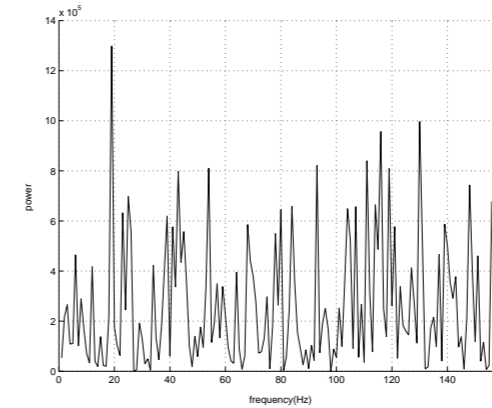
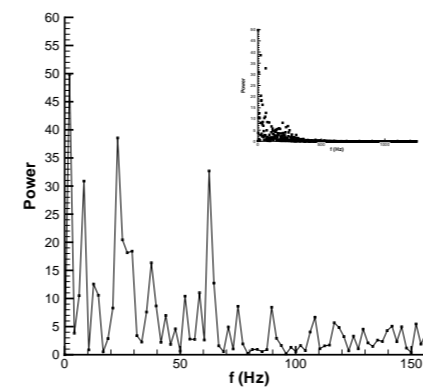
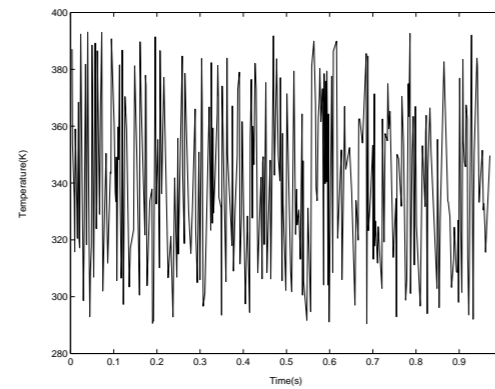
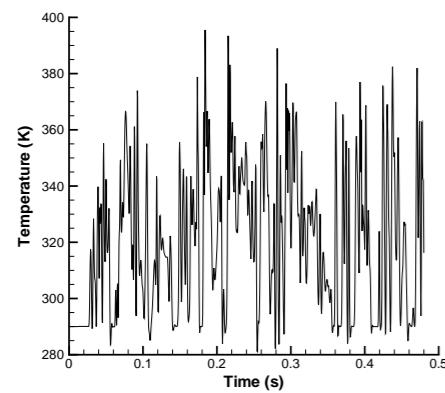
$$(x, y) = (0.238\text{m}, 0.04\text{m})$$

# Results

## Comparison of Computation and Experiment

Temperature vs time

Power spectrum



Computation

Experiment

Computation

Experiment

- Difference in spectra at higher frequencies due to three-dimensional effects in experiment

Jet “undulating” in spanwise direction

- What is the effect of the simulation model accuracy on accuracy of Response Surface model ?

## Current Focus of Research

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- Two issues need to be addressed in systematic manner regarding the accuracy of the Response Surface model
  - Signal level  
Currently using simple statistical model to discard unrealistic predicted values for  $U_j, p_j, T_j$
  - Accuracy of simulation model  
Need to incorporate the uncertainty of the simulation model into the DDDAS methodology  
The uncertainty in the simulation model affects:
    - \* The uncertainty in the predicted boundary conditions
    - \* The convergence criteria for the DDDAS methodology
- The jet in crossflow is a good configuration for evaluating method addressing these issues

## Conclusions

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- Developing DDDAS methodology for evaluation of fluid thermal systems
  - Examples are optical fibre furnace and turbofan combustor
  - Need for complete flowfield simulation to optimize system performance
  - Boundary conditions for flowfield simulation are not completely known *a priori*
  - Non-intrusive optical measurements (*e.g.*, laser diode absorbance) feasible in limited region
  - DDDAS method to determine complete boundary conditions by synergizing experiment and simulation

