Pressure Flow Field and Inlet Flow Distortion Metrics Reconstruction from Velocity Data

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Complex engine intakes are susceptible to unsteady flow distortions that may compromise the propulsion system operability. Hence, the need for high spatial and temporal resolution flow information is essential to aid the development of distortion tolerant, closely coupled propulsion systems. Stereoscopic PIV methods have been successfully applied to these flows offering synchronous velocity datasets of high spatial resolution across the Aerodynamic Interface Plane. However, total pressure distortion measurements are still typically provided by low bandwidth, intrusive total pressure rakes of low spatial resolution which results in limited characterisation of the total pressure distortion. This limitation can potentially be addressed by pressure field reconstruction from non-intrusive, high resolution velocity data. A range of reconstruction methods are assessed based on representative data from steady and unsteady computational simulations of an S-duct configuration. In addition to the reconstructed total pressure field, the impact on the key distortion metrics is assessed. The effect of Mach number is considered. Overall the reconstruction methods show that the distortion metrics can be determined with sufficient accuracy to indicate that there is a potential benefit from exploiting high resolution velocity measurements in evaluating total pressure distortion.

Nomenclature

\[ A = \text{area}, \text{m}^2 \]
\[ c_p = \text{specific heat capacity at constant pressure}, \text{J/kgK} \]
\[ D = \text{diameter}, \text{m} \]
\[ H = \text{offset}, \text{m} \]
\[ I = \text{non-dimensional accuracy index} \]
\[ k = \text{turbulence kinetic energy}, \text{m}^2/\text{s}^2 \]
\[ L = \text{length}, \text{m} \]
\[ M = \text{Mach number} \]
\[ m_\theta = \text{total number of azimuthal grid nodes} \]
\[ n = \text{number of snapshots} \]

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\( n_r \) = total number of radial grid nodes
\( p \) = static pressure, Pa
\( p_0 \) = total pressure, Pa
\( P_0 \) = area-averaged total-pressure, Pa
\( q \) = dynamic head, Pa
\( r, \theta, z \) = cylindrical frame of reference, m, rad, m
\( R \) = gas constant, J/kgK
\( t \) = static temperature, K
\( t_0 \) = total temperature, K
\( t \) = time, s
\( t^* \) = acquisition time, s
\( t_{\text{conv}} \) = convective time, s
\( u_r, u_\theta, u_z \) = instantaneous velocities in cylindrical frame of reference, m/s
\( u \) = velocity vector, m/s
PPE = Poisson pressure equation
PR = Pressure ratio
DSI = direct spatial integration

**Subscripts**

- avg, mean = average
- \( i,j,k \) = indices
- in = inlet plane
- max = maximum
- out = exit plane
- rec = reconstructed
- ref = reference value at S-duct inlet
- rms = root mean square

**Greek symbols**

\( \gamma \) = heat capacity ratio
\[ \mu = \text{dynamic viscosity, Ns/m}^2 \]
\[ \rho = \text{density, kg/m}^3 \]
\[ \omega = \text{specific turbulence dissipation, 1/s} \]
\[ \Omega = \text{computational domain} \]

**Operators**

\[ \nabla = \text{divergence} \]

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**I. Introduction**

Advanced propulsion system installations for current and future aero vehicle architectures feature short and complex intake configurations where the coupling with the engine becomes critical [1,2]. Previous work showed that complex engine intakes are notably susceptible to flow separations and large unsteady perturbations, which can potentially compromise the operability of the entire propulsion system [3-6]. Examples of flow non-uniformities at the exit of a complex intake are presented in Fig. 1 which shows the computed time-averaged total pressure and swirl angle distortion distributions [7]. These are results from Reynolds Averaged Navier Stokes (RANS) based simulations for an S-duct with an area ratio \( A_{\text{out}}/A_{\text{in}}=1.52 \) and offset to inlet diameter ratio \( H/D_{\text{in}}=1.34 \) which was also previously studied experimentally by Wellborn et al [8]. These previous works indicate a flow separation within the first bend of the S-duct (Fig. 1a) which interacts with the associated secondary flows and results in two counter-rotating vortical structures located at the lower part of the AIP (Fig. 1b, c). These are further highlighted by the flow path lines superimposed on the swirl angle and pressure recovery contours of Fig. 1. The unsteady behaviour of these types of flow fields and the associated unsteady swirl and pressure distortions were further analysed by Chiereghin et al [7], MacManus et al [9] and Garnier et al [10,11]. These works highlight the need for spatially and temporally rich and synchronous datasets to enable the assessment of total pressure and swirl distortion for operability assessments.

![Fig. 1 RANS predicted flow field. (a): surface streaklines (b): swirl angle (c): AIP total pressure ratio [7].](image)
Although current industry practice typically relies on low spatial resolution intrusive rakes for the measurement of the pressure and swirl distortion patterns these to be unable to capture the complex nature of the flows in both the spatial and the temporal domain [3]. This limitation of the intrusive swirl distortion measurement systems can be addressed by applying Stereoscopic Particle Image Velocimetry (S-PIV) techniques for the measurement of the velocity fields at the aerodynamic interface plane (AIP). This approach has the potential to provide 200-300 times higher spatial resolution than the conventional intrusive rakes across a typical AIP in a synchronous way and enables three component, planar velocity measurements. These can then be used to evaluate swirl distortion metrics at each snapshot as well as to evaluate the distortion statistics over a period of time. This method was successfully applied for AIP velocity field measurements at the exit of complex intakes across a range of inlet Mach numbers as reported by Zachos et al [12].

However, the need for total pressure distortion characterisation still remains as S-PIV systems only provide velocity measurements. A potentially attractive approach to establish a direct link between swirl distortion measurements and the associated pressure distortion could be through the reconstruction of the underpinning AIP static pressure field by using velocimetry data. This way the advantages of the S-PIV synchronous and spatially rich velocity fields could be further exploited for the derivation of the pressure distortion characteristics associated with them. Furthermore, apart from the inherent synchronisation of pressure and velocity information such an approach removes the need for additional intrusive instrumentation to measure pressure distortion.

Pressure field reconstruction from velocity data techniques are relatively well known and have been applied to determine integral aerodynamic forces and moments in fluid-structure interactions. Several of these methods have been proposed to enable the coupling of the pressure fields with the associated mechanical loads as reported by Lin and Rockwell [13], Unal et al [14], Noca et al [15,16] and Fujisawa et al [17]. These methods have been developed to support aerodynamic load determination in applications investigated using velocimetry techniques and to allow the synchronous estimation of flow and load information. For temporally resolved PIV data, these methods often permit flow acceleration terms to be determined with a sufficient level of accuracy as discussed by van Oudheusden et al in [18,19] and Hosokawa et al [20].

The aim of this work is to apply pressure field reconstruction methods in order to characterise the total pressure distortion generated by complex engine intakes using velocity information. Pressure fields are reconstructed by means of Direct Spatial Integration (DSI) of the momentum equation [14,19] as well as by integration of the Poisson Pressure
Equation (PPE) as proposed by Fujisawa et al [17] and Hosokawa et al [20]. Velocity data from unsteady Computational Fluid Dynamics (CFD) simulations are used for the AIP pressure field reconstruction for a typical S-duct configuration with an offset to inlet diameter ratio $H/D_i = 2.44$, area ratio $A_{out}/A_{in} = 1.52$ across a range of inlet Mach numbers $M_i$ between 0.27 and 0.6. The duct exit had a diameter of 150mm at the AIP while the inlet diameter $D_i$ was 121.6 mm. The Reynolds number based on the duct inlet diameter ranged from $0.7 \times 10^6$ to $1.4 \times 10^6$. The generated pressure fields are then used for the calculation of the total pressure distortion metrics and the results are compared against pressure field distributions and distortion metrics directly calculated from the CFD data. The effects of temporal and out-of-plane velocity gradients on reconstruction accuracy are assessed with the aim of determining the applicability of time-resolved, stereoscopic PIV systems for steady and unsteady pressure distortion characterisation of the flow at the exit of a complex intake.

II. Theoretical background and methods

A. Governing equations

For the calculation of the pressure field the incompressible momentum equation was used in conservative form as follows:

$$\nabla p = \rho \left( \frac{\partial u}{\partial t} + u \cdot \nabla u \right) + \mu \nabla^2 u$$

Equation (1) was solved in cylindrical coordinates on the AIP plane (Fig. 2) at the exit of the S-duct intake which had a diameter $D_{AIP} = 150$ mm. A polar grid was used to discretise the computational domain (Fig. 2) with a total of 9,000 nodes arranged in 50 equi-spaced rakes and 180 equi-spaced circumferential rings. This spatial resolution is similar to that achieved by S-PIV measurements of the same S-duct configuration ($H/D_i = 2.44, D_{AIP} = 150$ mm) reported by Zachos et al [12]. Considering a two dimensional, incompressible and inviscid flow [19] at the AIP, the spatial pressure gradients can be estimated in cylindrical coordinates as follows:

$$\frac{\partial p}{\partial r} = \rho \left( \frac{\partial u_r}{\partial t} + u_r \frac{\partial u_r}{\partial r} + u_\theta \frac{\partial u_r}{\partial \theta} + u_z \frac{\partial u_r}{\partial z} \right)$$

$$\frac{1}{r} \frac{\partial p}{\partial \theta} = \rho \left( \frac{\partial u_\theta}{\partial t} + u_r \frac{\partial u_\theta}{\partial r} + u_\theta \frac{\partial u_\theta}{\partial \theta} + u_z \frac{\partial u_\theta}{\partial z} \right)$$
Equations (2) and (3) allow the derivation of the AIP spatial pressure gradients using only spatial and temporal velocity gradients, after which the pressure can be obtained from integration. A circumferential static pressure distribution along the outer boundary of the domain is required as a boundary condition to allow the calculation of the pressure field across the entire domain of interest.

Alternatively the spatial pressure gradient may also be obtained through solving the Poisson Pressure Equation which is derived from the inviscid form of Equation (1) by applying the divergence operation as follows [17]:

\[
\nabla^2 p = \rho \nabla \left[ \frac{du}{dt} + u \cdot \nabla u \right] = \frac{1}{r} \frac{\partial}{\partial r} \left( r \frac{\partial p}{\partial r} \right) + \frac{1}{r^2} \frac{\partial^2 p}{\partial \theta^2} + \frac{\partial^2 p}{\partial z^2} =
\]

\[
\rho \left[ \frac{1}{r} \frac{\partial u_r}{\partial r} + \frac{\partial^2 u_r}{\partial \theta^2} + \frac{1}{r} \frac{\partial^2 u_\theta}{\partial \theta \partial t} + \frac{1}{r} \frac{\partial^2 u_z}{\partial z \partial t} \right]
\]

\[
\rho \left[ \left( \frac{\partial u_r}{\partial r} \right)^2 + \frac{1}{r^2} \left( \frac{\partial u_\theta}{\partial \theta} \right)^2 + \left( \frac{\partial u_z}{\partial z} \right)^2 + \frac{2}{r} \frac{\partial u_\theta}{\partial \theta} \right]
\]

\[
+ 2 \frac{\partial u_z}{\partial r} \frac{\partial u_r}{\partial z} + 2 \frac{\partial u_z}{\partial \theta} \frac{\partial u_\theta}{\partial z} + \left( \frac{u_r}{r} \right)^2 \frac{2}{r} \frac{u_\theta}{r} \frac{\partial u_\theta}{\partial r}
\]

\[
+ 2 \frac{u_r}{r} \frac{\partial u_\theta}{\partial \theta}
\]

To solve Equation (4) in the polar domain (Fig. 2), a prescribed static pressure distribution along the outer boundary of the domain would normally suffice. This would be similar to the one used to solve Equations (2) and (3). However, in order to avoid singularities in the numerical integration scheme, the centre of the computational domain \((r=0)\) is removed. This yields to the necessity of a virtual boundary within the domain which requires a second boundary condition. The virtual boundary however cannot be treated with a Dirichlet boundary condition as the static pressure is unknown in this region. Hence, a Neumann type boundary condition is required whereby the divergence of the static pressure along the virtual boundary must be prescribed as:

\[
\nabla p \cdot n = g \quad \text{at} \quad \partial \Omega
\]

where \(\Omega\) is the computational domain, \(\partial \Omega\) the domain virtual boundary (Fig. 2), \(n\) the outward unit vector normal to \(\partial \Omega\) and \(g\) a known function along the domain boundary. Equation (5) finally reduces to \(\frac{\partial p}{\partial r} = g\) which can then be calculated by Equation (2) based on velocity data.
To introduce compressibility effects a possible solution would be to use the continuity equation to directly derive density at each node. However, as reported by van Oudheusden et al. [19] this method introduces errors in the pressure field reconstruction. A simpler method to estimate flow density was used herein based on the adiabatic flow condition and the ideal gas law [19]. The gas law is employed to replace the density in the momentum equations where the static temperature is derived from the assumption of constant total enthalpy using the velocity magnitude at each node as follows:

$$\rho_{i,j} = \frac{p_{i,j}}{RT}; \quad t = t_0 \frac{1}{2} \left( \frac{u_{i,j}}{c_p} \right)^2$$  \hspace{1cm} (6)

![Domain discretization at a given axial position k.](image)

**Fig. 2** Domain discretization at a given axial position $k$.

**B. Reconstruction numerical methods**

The radial and circumferential velocity gradients involved in the DSI and PPE total pressure reconstruction methods (Eq. (2), (3) and (4)) were approximated using a central finite difference scheme as follows:

$$\left( \frac{\partial u_i}{\partial r} \right)_{i,j,k} = \frac{u_{i+1,j,k} - u_{i-1,j,k}}{2 \Delta r} + O(r^2) \approx \frac{u_{i+1,j,k} - u_{i-1,j,k}}{2 \Delta r}$$  \hspace{1cm} (7)

$$\left( \frac{\partial u_i}{\partial \theta} \right)_{i,j,k} = \frac{u_{i,j+1,k} - u_{i,j-1,k}}{2 \Delta \theta} + O(\theta^2) \approx \frac{u_{i,j+1,k} - u_{i,j-1,k}}{2 \Delta \theta}$$  \hspace{1cm} (8)
where \( u_i \) refers to a generic velocity component and \( \Delta r \), \( \Delta \theta \) and \( \Delta z \) are the grid spacing in the radial and circumferential and out-of-plane (streamwise) directions respectively. For the calculation of the temporal velocity gradients the following expression was used:

\[
\left( \frac{\partial u_i}{\partial t} \right)_{l,j,k} \approx \frac{u_{i,j,k+1} - u_{i,j,k-1}}{2 \Delta t} \tag{9}
\]

For the reconstruction of time-averaged pressure fields, the temporal velocity gradients \( \left( \frac{\partial u_i}{\partial t} \right) \) were all set to zero and velocity data from the time-average field was used in Eq. (2), (3) and (4).

For the integration of the pressure gradients across the domain the space marching technique reported by Baur and Koengeter [21] was used. For the calculation of the static pressure at any given node, four already calculated pressure values from neighbouring nodes are used as integration paths to evaluate an average static pressure at given cross-flow plane \( k \) as follows:

\[
p_{i,j} = \frac{1}{4} \left[ (p_{i+1,j-1} + \Delta p_{i+1,j-1}) + (p_{i+1,j} + \Delta p_{i+1,j}) + (p_{i,j+1} + \Delta p_{i,j+1}) + (p_{i,j-1} + \Delta p_{i,j-1}) \right] \tag{11}
\]

The calculation proceeds towards the inner domain starting from the external far field boundary where the static pressure is known. A first order Taylor’s polynomial is used to evaluate the \( p_i + \Delta p_i \) terms across the plane as follows:

\[
p_{i,j} + \Delta p_{i,j} = p_{i,j} + \left( \frac{\partial p}{\partial r} \right)_{l,i,j} r_{ij} + \left( \frac{\partial p}{\partial \theta} \right)_{l,i,j} \theta_{ij} \tag{12}
\]

The spatial pressure gradient components \( \frac{\partial p}{\partial r} \) and \( \frac{\partial p}{\partial \theta} \) are calculated based on velocity data using Eq. (2) and (3).

For solving the Poisson Pressure Equation (Eq. (4)), the left hand side pressure terms are discretized using second order central derivatives as follows:
\[ \nabla^2 p_{i,j,k} = \left[ \frac{1}{r} \frac{\partial}{\partial r} \left( r \frac{\partial p}{\partial r} \right) + \frac{1}{r^2} \frac{\partial^2 p}{\partial \theta^2} + \frac{\partial^2 p}{\partial z^2} \right]_{i,j,k} \approx \frac{1}{r_{i,j,k}} \left( p_{i+1,j,k} + p_{i-1,j,k} + 2p_{i,j,k} + p_{i,j-1,k} \right) + \frac{1}{r_{i,j,k}^2} \frac{2p_{i,j,k} + p_{i,j-1,k}}{(r)^2} + \frac{1}{r_{i,j,k}^2} \frac{2p_{i+1,j,k} + p_{i,j,k}}{(\theta)^2} + \frac{1}{r_{i,j,k}^2} \frac{2p_{i,j,k} + p_{i,j,k-1}}{(z)^2} \] (13)

The spatial and temporal velocity derivatives that appear at the right hand side terms of Eq. (4) are discretized using Equations (7), (8), (9) and (10). The static pressure at each grid node across the computational plane is explicitly determined as:

\[ p_{i,j,k} = \left[ \frac{1}{2} \left( \frac{1}{r_{i,j,k}} + \frac{1}{r_{i,j,k}^2} \frac{1}{\theta} \right) \right] \left( \frac{1}{r_{i,j,k}} \frac{p_{i+1,j,k} + p_{i-1,j,k}}{2\Delta r} + \frac{1}{r_{i,j,k}^2} \frac{p_{i,j+1,k} + p_{i,j-1,k}}{(r)^2} + \frac{1}{r_{i,j,k}^2} \frac{p_{i,j,k+1} + p_{i,j,k-1}}{(\theta)^2} \right) f_{i,j,k} \] (14)

where \( f_{i,j,k} \) is the known velocity based term from Eq. (4). Equation (14) is solved using an iterative Gauss-Seidel method [22]. A successive over-relaxation factor was introduced to accelerate convergence. The domain was initialised with a static pressure distribution obtained from Eq. (6) and constant density applied across the AIP.

C. Computational methods

1. CFD method

The unsteady CFD calculations were performed using a Delayed Detached-Eddy-Simulation (DDES) with the \( k-\omega \) SST turbulence model. A pressure-based solver was used with a segregated PISO scheme [23]. The pressure spatial gradients were solved using a second-order scheme and a third order MUSCL scheme was used for momentum, energy and turbulence. The steady RANS simulations were performed using the same methods with the \( k-\omega \) SST turbulence model. For the unsteady DDES simulations, the temporal discretisation was addressed using a bounded second-order method [23]. The inlet boundary condition comprised specified uniform total pressure and total temperature profiles. The static pressure was specified at the domain exit and was adjusted to provide the required average Mach number at the inlet.

The overall duct domain was discretised using a multi-block structured mesh. The baseline mesh had 5 million nodes. The mesh had a H-grid structure in the central part of the duct, and an O-grid structure around the wall which
resulted in a good quality mesh as indicated by a 2x2x2 determinant greater than 0.8. The near wall boundary layers were resolved with a structured mesh which provided $y^+$ less than 1 over the full domain. Previous work [9] using this method for the same S-duct configuration, at a slightly greater Reynolds number, evaluated the mesh independence using grids of 3.1, 5.9 and 11.2 million nodes. For the unsteady DDES calculations the time step was adjusted to ensure similar Courant numbers for the different meshes. Between the medium and fine meshes the changes in time-averaged pressure ratio ($P_{0}/P_{0,ref}$) was less than 0.1% along with a 5.7% reduction in the temporal standard deviation of PR. The distortion metrics were more sensitive to the mesh resolution although they are not conserved thermodynamic properties due to the filters which were applied. The time-averaged metrics for swirl distortion changed by 3.2% while the key total pressure distortion parameters changed by between 1.8% and 8.0%.

2. Computational time steps

A time step $t$ of 6x10$^{-6}$ s was chosen for the $M_{in}$=0.60 case which corresponds to a non-dimensional time step $t$ of approximately 0.0018 with respect to the mean overall convective time through the duct. The flow convective time for the $M_{in}$=0.6 cases was approximately 3.4x10$^{-3}$ s. Previous work [9] investigated the effect of the time step for the $M_{in}$=0.60 configuration which was also simulated with a time step of 12x10$^{-6}$ s and the time-averaged and unsteady swirl and total pressure distortion metrics were compared. The results showed that the time-averaged values of the distortion metrics were changed by up to 3% with the time step reduction from 12x10$^{-6}$ to 6x10$^{-6}$ s. In this work the DDES calculations used 20 sub-iterations per time step which typically resulted in residuals in the order of 10$^{-6}$ for continuity equation and 10$^{-7}$ for momentum, energy, $k$ and $\omega$ equations, with a reduction of at least three orders of magnitude of all the residuals for each time step. For all the cases, a discarded interval of 15 mean convective times where used as a transition between the steady solution and the established unsteady flow field. The subsequent unsteady simulations were conducted for 55 mean throughflow times.

3. CFD validation

The CFD method adopted in this study was previously validated based on experimental data for the same S-duct geometry but at a slightly increased size and Reynolds number [9]. The Mach number range was the same as considered here as $M_{in}$=0.27 to 0.60. Overall the simulated time-averaged total pressure ratio was 0.993 and 0.971 for $M_{in}$ =0.27 and 0.60, respectively, which was the same as the measurements. The maximum unsteady PR typically agreed within 5% of the measured data. Overall, the DDES method has been examined and validated to demonstrate that it is capable of simulating these types of flow fields. Consequently, in the assessment of the pressure
reconstruction methods, the DDES data is an appropriate source of high-resolution, representative data. For the reconstruction analyses in this current work the CFD data at the AIP was extracted and interpolated, by Kriging, onto a uniform polar grid with 50 radial and 180 circumferential points.

III. Results and discussion

A. Time-averaged pressure field reconstruction

The first assessment of the reconstruction method was performed by considering a steady flow field. The steady total pressure reconstruction was carried out using numerically calculated pressure and velocity data from steady RANS simulations. The steady total pressure recovery AIP distributions are shown in Fig. 3 for \(M_\infty=0.6\). The DSI based mean static pressure field was calculated from average velocity flow field data at the AIP by using the time-averaged form of the instantaneous momentum equations (Equations (2) and (3)). PPE based static pressure fields were reconstructed from the steady Poisson pressure expression (Equation (4)). The total pressure at the AIP nodes was calculated from the isentropic pressure relation (Equation (6)) and the reconstructed static pressure values as:

\[
\frac{p_0}{p} = \left(1 + \frac{\gamma}{2} \frac{1}{M^2}\right)^{\frac{\gamma}{\gamma-1}}
\]

(15)

In order to assess the mean total pressure field reconstruction accuracy, the difference between the reconstructed \(p_{0,\text{rec}}\) and the RANS calculated total pressure \(p_{0,\text{CFD}}\) was determined at each point across the AIP as:

\[
p_0 = \frac{p_{0,\text{rec}}}{p_{0,\text{CFD}}}
\]

(16)

The accuracy of a pressure reconstruction algorithm was reduced to a single index defined across the AIP as:

\[
l = \sqrt{\frac{\sum_{i=1}^{n_r} \sum_{j=1}^{m_\theta} (p_{\text{rec}} - p_{\text{CFD}})^2}{n_r m_\theta}}
\]

(17)
where $q$ is the mean dynamic head at the inlet of the S-duct defined as $q = \frac{1}{2} \rho U_{in}^2$ with $\rho$ the flow density and $U_{in}$ the meanline flow velocity at the inlet. Index $I$ is useful when comparing between different pressure reconstruction methods as it represents the non-dimensional root mean square of the average reconstruction error across the domain.

![Fig. 3 RANS computed steady AIP total pressure recovery distribution for the S-duct with H/D=2.44. at M_{in}=0.6.](image)

Distributions of the reconstructed total pressure ratio ($p_0/p_{0,ref}$) at the AIP which is located 0.296D\_in downstream of the S-duct exit plane, as well as the discrepancy from the time-averaged RANS computed total pressure field, are shown in Fig. 4 and Fig. 5. This is for the high offset S-duct (H/D=2.44) and with an inlet Mach number $M_{in}$ of 0.6 defined at a reference plane 0.934D\_in upstream of the S-duct inlet. These reconstructions use the DSI scheme for various combinations of treatment for the density and out-of-plane velocity gradient terms. For the total pressure field reconstructions the static pressure distribution imposed along the domain boundary was obtained from the RANS simulations.

For the reconstruction with variable density and the $\left(\frac{\partial u_l}{\partial z}\right)$ terms included, the reconstructed total pressure distribution ($p_0/p_{0,ref}$) exhibits the main features of the original data (Fig. 4a). The error is relatively uniformly distributed across the AIP with peak values in the order of ±0.7% (Fig. 4b). For the same base flow field and when the $\left(\frac{\partial u_l}{\partial z}\right)$ term was neglected in the reconstruction, the broad topology remained the same (Fig. 4c) and the peak error increased to approximately 1% (Fig. 4d). Relative to the baseline DSI reconstruction case (Fig. 4a), the impact of the density terms showed that for the case with a constant density (Fig. 4e, f) there is a very slight increase in the error which predominately affects the upper sector of the AIP where it increases from 0.7% to 1.2%. Finally, the
overall combined effect of assuming a constant density as well as a \(\frac{\partial u}{\partial z}\) = 0 (Fig. 4g, h) gives the worst result with the peak error at approximately 1.5%. The same characteristics are reflected in the \(I\)-index (Table 1), whereby the effect of the \(\frac{\partial u}{\partial z}\) term is relatively more important than the treatment of the density. However, the level of the errors is still relatively low with \(I\) in the region of 0.017 for the worst case. This finding is pertinent to pressure field reconstruction when planar experimental velocimetry data is available, such as reported by Zachos et al [12], which do not enable the calculation of the out-of-plane velocity gradients.

Fig. 4 Steady DSI reconstruction for the S-duct with \(H/D = 2.44\) at \(M_\infty = 0.6\). Top row: time-averaged reconstructed pressure distributions, bottom row: discrepancy from RANS predicted fields.
Fig. 5 Steady PPE reconstruction for the S-duct with H/D_{in}=2.44 at M_{in}=0.6. Top row: time-averaged reconstructed total pressure distributions, bottom row: discrepancy from RANS predicted fields.

The PPE integration approach also has an impact on the reconstruction and the relative sensitivities to the treatment of the density and \( \left( \frac{\partial u_i}{\partial x} \right) \) terms (Fig. 5). Overall, the PPE approaches also reconstruct the same general \( p_0/p_{0,ref} \) topologies (Fig. 5) in comparison with the DSI method (Fig. 4) and the original time-averaged flow (Fig. 3). For the configuration with variable density and including the \( \left( \frac{\partial u_i}{\partial x} \right) \) terms, the peak error for the PPE method slightly increases to 1.1\% (Fig. 5b) relative to the same assumptions using the DSI method which had a peak error of 0.7\% (Fig. 4b). Overall for all of the PPE cases, the reconstruction accuracy remains in the region of 1.1\%. Relative to the DSI method, the PPE approach is less sensitive to the effects of changes to the treatment of density and \( \left( \frac{\partial u_i}{\partial x} \right) \) (Fig. 5). When \( \left( \frac{\partial u_i}{\partial x} \right) = 0 \) the maximum error is generally unaffected although there is a very slight increase in \( I \)-index from 5.3x10^{-3} to 5.6x10^{-3} for M_{in}=0.6 (Fig. 5b, d, Table 1). When density is held constant (Fig. 5f, h) the effects are even less pronounced with a small decrease in \( I \) from 5x10^{-3} to 4.8x10^{-3} (Fig. 5f, h, Table 1).

Pressure field reconstructions at a lower inlet Mach number of 0.27 were also conducted with both the DSI and PPE methods. Relative to the M_{in}=0.6 cases, the error in the total pressure fields at the AIP reduced to approximately
± 0.2% when the out-of-plane gradients were accounted for. When \( \left( \frac{\partial u_i}{\partial z} \right) = 0 \), the error marginally increased to 0.6% across the AIP. Based on the accuracy index \( I \), the low Mach cases (Table 1) show that for both the DSI and PPE methods the errors are broadly independent of \( M_{in} \) across this range. In addition, the sensitivity to treatment of density and the out-of-plane velocity gradient is similarly relatively independent of the Mach number.

Overall, both the DSI and PPE based pressure reconstruction methods performed sufficiently well for steady total pressure field reconstruction from velocity data. Total pressure fields obtained by the PPE integration provide slightly more accurate reconstructions relative to the DSI method across the range of \( M_{in} \) and are less sensitive to changes to the treatment of density and \( \left( \frac{\partial u_i}{\partial z} \right) \) terms. Given that broadly the discrepancies of both methods remain at very low levels regardless of the treatment of the out-of-plane \( \left( \frac{\partial u_i}{\partial z} \right) \) velocity gradient or density terms, it can be concluded that in principle these procedures could potentially allow the derivation of steady field data from planar, 3-component PIV measurements with a sufficient level of accuracy. The effect of the reconstruction on the conventional flow distortion metrics is considered in Section C.

### Table 1 Accuracy index \( I \) for reconstructed steady total pressure fields.

<table>
<thead>
<tr>
<th>DENSITY</th>
<th>( \frac{\partial u_i}{\partial z} )</th>
<th>( I \times 10^3 )</th>
<th>( I \times 10^3 )</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>( M_{in} = 0.6 )</td>
<td>( M_{in} = 0.27 )</td>
<td></td>
</tr>
<tr>
<td>DSI</td>
<td>variable on</td>
<td>6.7</td>
<td>6.4</td>
</tr>
<tr>
<td></td>
<td>variable off</td>
<td>15</td>
<td>16</td>
</tr>
<tr>
<td></td>
<td>constant on</td>
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<td></td>
<td>constant off</td>
<td>15</td>
<td>17</td>
</tr>
<tr>
<td>PPE</td>
<td>variable on</td>
<td>5.3</td>
<td>5.5</td>
</tr>
<tr>
<td></td>
<td>variable off</td>
<td>5.6</td>
<td>6.5</td>
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<tr>
<td></td>
<td>constant on</td>
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<td>5.5</td>
</tr>
<tr>
<td></td>
<td>constant off</td>
<td>4.8</td>
<td>6.5</td>
</tr>
</tbody>
</table>
B. Unsteady pressure field reconstruction

Delayed Detached Eddy Simulations (DDES) provided time histories of the flow field through the S-duct for the configuration with $M_a=0.6$. The simulations showed a highly dynamic flow in which the topology of the total pressure and velocity flow field at the AIP varied significantly with large changes in the distortion metrics [7,9]. Furthermore, both the total pressure and swirl distortions at the AIP were calculated to be substantially different from the symmetric time-averaged flow field. Consequently, the wide range of flow field characteristics and features which arise from the DDES simulations provide a robust test for the unsteady reconstruction methods considered in this work. The DDES simulation time step $\Delta t$ was $6 \times 10^{-6}$ s with the solution saved every 3 time steps which results in a time interval of $18 \times 10^{-6}$ s between two subsequent datasets. From the full DDES simulation, a representative set of 400 time steps, equivalent to 2.2 mean throughput convective times, were used to provide a sample of dynamic distortion fields for the assessment of the proposed reconstruction methods.

For this set of unsteady flow field data at the AIP, the methods to reconstruct the pressure field from the 3 components of velocity were evaluated. The accuracy of the unsteady total pressure reconstruction methods considered the variable density DSI approach and the impact of out-of-plane velocity gradients $\left(\frac{\partial u_i}{\partial z}\right)$ as well as temporal velocity gradients $\left(\frac{\partial u_i}{\partial t}\right)$ (Table 2). The metric to quantify the overall reconstruction tools is the averaged root mean square error defined in Eq. (17) as:

$$\bar{I} = \frac{1}{n} \sum_{i=1}^{n} I_i$$

where $n$ is the total number of instantaneous flow fields considered.

Fig. 6 shows instantaneous total pressure ratio ($p_{i}/p_{i,ref}$) distributions at the AIP for four representative normalised time instances across a range between $t^*=0.025$ and 0.225 where $t^*=400 \times 18$ $\mu$s. Fig. 6 highlights the remarkable flow non-uniformities across the AIP for a given time step as well notable changes over time. As discussed by MacManus et al [9] steady state or even unsteady RANS calculations are unable to reveal the complex underpinning aerodynamic structures of the flow fields. In addition, low bandwidth and low resolution distortion rakes, typically 8x5, can hardly provide the required spatial and temporal resolution to sufficiently measure these flow fields experimentally. Numerical methods of higher fidelity, such as unsteady DDES, or experimental techniques capable
of delivering sufficiently high spatial and temporal resolution such as S-PIV or time-resolved S-PIV are better positioned to capture the complexity of the underpinning flow mechanisms [7,9].

Fig. 7 and Fig. 8 show reconstructed total pressure fields using the unsteady DSI method as well as the relative discrepancies from the full DDES solution (Eq. (16)). For the reconstruction of the pressure fields shown in Fig. 7 the out-of-plane \( \frac{\partial u_i}{\partial x} \) and temporal velocity \( \frac{\partial u_i}{\partial t} \) gradients are included in the analysis. The relative effect of excluding these terms is shown in Fig. 8. This is pertinent when total pressure reconstruction is attempted based on time-resolved, tomographic velocimetry data. The flow density was set as variable in both cases. The total pressure fields generated using the unsteady DSI approach show a discrepancy within a typical range of \( \pm 0.5\% \) which locally increases up to \( \pm 1.8\% \) from the correspondent DDES flow snapshots (Fig. 7). The DSI algorithm demonstrated notable robustness in reconstructing the pressure fields of the different instances of the unsteady velocity fields despite the high rates of velocity gradients in both the circumferential and radial directions of the AIP.

A numerical experiment was conducted using the unsteady DSI reconstruction method and neglecting the out-of-plane \( \frac{\partial u_i}{\partial x} \) and temporal velocity \( \frac{\partial u_i}{\partial t} \) gradients. The rationale for this numerical experiment was to identify the extent to which the omission of the \( \frac{\partial u_i}{\partial x} \) and \( \frac{\partial u_i}{\partial t} \) terms compromises the calculation of the pressure gradients (Eq. (2) and (3)). This is pertinent to unsteady total pressure field reconstructions when only low temporal resolution, single-plane velocity data is available. If only temporally underresolved velocity data is available, inclusion of the \( \frac{\partial u_i}{\partial t} \) terms into the static pressure calculation (Eq. (2) and (3)) would produce erroneous calculation of the static pressure gradients [24]. Hence the current investigation aims to quantify for this type of flow field the fidelity of the pressure field reconstruction for the \( \frac{\partial u_i}{\partial x} \) terms neglected from the calculation. A rigorous study on the identification of the temporal resolution required for the velocity data to generate representative static pressure fields around an airfoil located downstream of a circular rod was conducted by Violato et al [24]. This study reports an estimate of the maximum time interval between two subsequent velocity fields, \( \Delta t \), above which the static pressure gradients of Equations (2) and (3) cannot be representatively evaluated due to erroneous calculation of the flow acceleration. The outcome of Violato’s studies [24] showed the effect of the unsteady velocity acquisition rate on the evaluation of the flow accelerations included in the pressure reconstruction equations for the external flows.
considered. This is pertinent to the determination of an appropriate setup for an unsteady velocity measurement method such as time-resolved planar S-PIV or tomographic PIV.

Reconstructed total pressure fields using the unsteady DSI method are shown in Fig. 8 where the out-of-plane \( \left( \frac{\partial u_i}{\partial z} \right) \) and temporal velocity \( \left( \frac{\partial u_i}{\partial t} \right) \) gradients have been neglected. These reconstructed total pressure fields result in local maximum differences from the DDES original data of up to ±8% .This is notably higher than the 1.8% maximum discrepancy when the out-of-plane \( \left( \frac{\partial u_i}{\partial z} \right) \) and temporal \( \left( \frac{\partial u_i}{\partial t} \right) \) terms are included (Fig. 7). The mean accuracy index for each case (Table 2) highlights this loss in fidelity as \( \bar{T} \) increases from 9x10\(^{-3}\) to 49x10\(^{-3}\) when the out-of-plane and the temporal terms are both neglected. However, in spite of this relative increase in the error, this simplified unsteady DSI reconstruction generates total pressure fields whose main characteristics are representatively captured in comparisons with the original field (Fig. 6). This suggests that useful total pressure field reconstructions may be feasible from planar, temporally underresolved velocity information, despite the accuracy penalty introduced by the lack of streamwise and temporal velocity gradients. The next section assesses this aspect through an evaluation of the reconstructed flow distortion metrics.

![Fig. 6 Instantaneous total pressure fields predicted from DDES simulations for the high offset S-duct (H/D\(_{in}=2.44\)) at M\(_{in}=0.6\)](image)

**Table 2 Average accuracy index \( \bar{T} \) for unsteady DSI reconstruction for the S-duct at M\(_{in}=0.6\).**

<table>
<thead>
<tr>
<th></th>
<th>( \left( \frac{\partial u_i}{\partial z} \right) )</th>
<th>( \left( \frac{\partial u_i}{\partial t} \right) )</th>
<th>( \bar{T} \times 10^3 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>DSI</td>
<td>on</td>
<td>on</td>
<td>9</td>
</tr>
<tr>
<td>DSI</td>
<td>off</td>
<td>off</td>
<td>49</td>
</tr>
</tbody>
</table>

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Fig. 7 Unsteady DSI total pressure field reconstruction and discrepancy from DDES. Out-of-plane and time derivative terms included. S-duct with H/D_{in}=2.44 at M_{in}=0.6.

Fig. 8 Unsteady DSI total pressure field reconstruction and discrepancy from DDES. Out-of-plane and time derivative terms neglected. S-duct with H/D_{in}=2.44 at M_{in}=0.6. Contours limited to be consistent with other figures.
C. Total pressure distortion metrics reconstruction

1. Steady reconstruction

Current industry practice for the assessment of flow distortion typically relies mainly on steady state and occasionally unsteady experimental measurements for the quantification of distorted flow fields for compressor or fan systems [4,5]. These allow the characterization of distortion through a set of distortion descriptors. A standard measurement arrangement for advanced engine intakes uses a total pressure rake at the AIP comprising an array of 8 spokes with 5 probes each (Fig. 9). To assess the total pressure distortion a range of descriptors is typically considered and the calculations are based on a “ring and rake” approach (Fig. 9). The total pressure recovery coefficient \( PR = \frac{P_{0,AIP}}{P_{0,in}} \) is also a property of interest. The circumferential distortion index (CDI) assesses the uniformity of the circumferential total pressure distribution and is defined as follows [25]:

\[
CDI = \max_{i=1}^{n_{\text{radius}}} \left( 0.5 \left( \frac{P_{0,i}}{P_{0,\text{avg}}} + \frac{P_{0,\text{min}_i}}{P_{0,\text{avg}}} \right) \right)
\]  

where \( P_{0,\text{avg}} \) is the average total pressure, \( P_{0,i} \) the average total pressure of the pressure distribution of the \( i \)-th ring and \( P_{0,\text{min}_i} \) the minimal pressure of the \( i \)-th ring. Finally the radial distortion can be assessed by the radial distortion index (RDI). The formula follows a similar logic as CDI and is defined as follows [25]:

\[
RDI = \max \left( \frac{P_{0,\text{avg}}}{P_{0,\text{inner ring}}}, \frac{P_{0,\text{avg}}}{P_{0,\text{outer ring}}} \right)
\]

where \( P_{0,\text{inner ring}} \) is the average total pressure of the pressure distribution of the inner ring and \( P_{0,\text{outer ring}} \) is the average total pressure at the outer ring. Finally, DC(60) is an overall distortion metric which is defined as the difference between the average total pressure, \( P_{0,\text{avg}} \), and the lowest average total pressure in a sector of 60° angle, \( P_{0,60\circ,\text{avg}} \) and non-dimensionalized by the mean dynamic head \( q \) of the AIP [25].

\[
DC(60) = \frac{P_{0,\text{avg}} - P_{0,60\circ,\text{avg}}}{q}
\]

where \( P_{0,\text{avg}} \) is the mean total pressure and \( P_{0,60\circ,\text{avg}} \) is the mean total pressure measure in a sector of 60 degrees.

In this work an 8x5 rake and ring arrangement was used for the distortion descriptors calculation using CFD and reconstructed pressure fields (Fig. 9). Pressure recovery and distortion characteristics based on the time-averaged RANS predicted flow fields are summarised in Table 3 and Table 4 for the S-duct at \( M_{\infty}=0.6 \) and \( M_{\infty}=0.27 \), respectively.
The distortion metrics calculated from the reconstructed steady total pressure fields for both the DSI and PPE approaches (Fig. 4 and Fig. 5) are summarised in Table 3 and Table 4 for the $M_a=0.6$ and 0.27 cases. The discrepancy between the reconstructed values and the values calculated directly from the original RANS pressure fields is also shown. This is calculated as:

$$\Delta x = \frac{x_{rec}}{x_{CFD}}$$

where $x_{rec}$ is the metric of interest (Eq. (19) to (21)) calculated from the reconstructed pressure data and $x_{CFD}$ the same property based on the original CFD pressure field.

Both the steady DSI and steady PPE reconstructed total pressure fields show a relatively accurate calculation of PR with a discrepancy from the RANS based value not higher than 0.5% (Table 3). The out-of-plane $\left(\frac{\partial u_i}{\partial z}\right)$ velocity gradients have no impact on the PR reconstruction fidelity, while the discrepancy in PR only slightly increases from -0.4% to -0.5% for total pressure fields reconstructed with constant density across the whole domain. At $M_a=0.6$, all the distortion descriptors, DC60, RDI and CDI, are underestimated by both methods and for various combinations of out-of-plane gradients and density assumptions. These errors range from -1.2% to -18.5%. DC(60) is underestimated by both DSI and PPE reconstruction approaches with the highest discrepancies from the constant density DSI approach where, neglecting $\left(\frac{\partial u_i}{\partial z}\right)$, the error is -18.5% (Table 3). This reduces to -10% when the variable density and $\left(\frac{\partial u_i}{\partial z}\right)$ terms are included.

PPE reconstructed total pressure fields underestimate the DC(60) by around -7% relative to the RANS data (Table 3). This under-prediction is not affected by the out-of-plane $\left(\frac{\partial u_i}{\partial z}\right)$ terms and remains constant when the total pressure...
field is reconstructed with constant density. CDI is consistently underpredicted by both PPE and DSI by around 3-5% relative to the RANS data and is insensitive to the out-of-plane gradients or treatment of density. Finally, RDI is also underestimated by both methods with a relative error of between -1.2% and -7.9% although the absolute differences are small. PPE is broadly insensitive to density and \( \frac{\partial u_i}{\partial x} \) terms. For the DSI method the error increases from -2.5% to -7.9% when the \( \frac{\partial u_i}{\partial x} \) terms are neglected.

Total pressure distortion descriptors were reconstructed for \( M_{in}=0.27 \) using both DSI and PPE methods and variable density across the AIP (Table 4). At this lower \( M_{in} \) the original total pressure ratio \( \left( \frac{P_{0,AIP}}{P_{0,in}} \right) \) increases from 0.963 at \( M_{in}=0.6 \) to 0.998 although the DC60 distortion metric only slightly changes from 0.231 to 0.223. At this lower Mach number, and also partially due to the definition of the terms, both CDI and RDI are substantially reduced from 0.068 and 0.041 to 0.013 and 0.008, respectively. Both the DSI and PPE methods reflect the changes in PR and there is no difference between the reconstructed and original data for PR, CDI and RDI (Table 4). At this lower Mach number the difference in DC60 has reduced for both the DSI and PPE methods. The PPE is still insensitive to the \( \frac{\partial u_i}{\partial x} \) terms and typically overestimates DC60 by up to 2%. The DSI approach is more sensitive to these terms and the error ranges from about -4 to -11%.

The analysis on the descriptor reconstruction for steady data shows that generally the two reconstruction approaches offer broadly the same level of accuracy in descriptor estimation across the range of \( M_{in} \) except for DC(60). DC(60) is underestimated by a maximum of almost -20% by both DSI and PPE methods. These errors reduce to about -11% and 2% respectively for \( M_{in}=0.27 \). These assessments give an indication of the uncertainty in steady total pressure distortion descriptors based on planar, mean flow velocity data such as that from planar S-PIV measurements. A measured velocity dataset which includes the out-of-plane velocity terms, such as from tomographic PIV experiments, would provide higher confidence in the total pressure descriptors. However, for this steady flow, the benefits are very small.
Table 3 Reconstructed steady distortion metrics and discrepancy from RANS simulations. High offset S-duct at $M_{in}=0.6$.

<table>
<thead>
<tr>
<th>DENSITY</th>
<th>$\frac{\partial u_1}{\partial z}$</th>
<th>PR</th>
<th>ΔPR %</th>
<th>DC(60)</th>
<th>ΔDC(60) %</th>
<th>CDI</th>
<th>ΔCDI %</th>
<th>RDI</th>
<th>ΔRDI %</th>
</tr>
</thead>
<tbody>
<tr>
<td>RANS</td>
<td>resolved included</td>
<td>0.963</td>
<td>-</td>
<td>0.231</td>
<td>-</td>
<td>0.068</td>
<td>-</td>
<td>0.041</td>
<td>-</td>
</tr>
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<td>DSI</td>
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<td>-0.4</td>
<td>0.210</td>
<td>-10.0</td>
<td>0.066</td>
<td>-3.0</td>
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<tr>
<td></td>
<td>variable off</td>
<td>0.959</td>
<td>-0.4</td>
<td>0.205</td>
<td>-12.7</td>
<td>0.066</td>
<td>-3.0</td>
<td>0.041</td>
<td>-1.2</td>
</tr>
<tr>
<td></td>
<td>constant on</td>
<td>0.958</td>
<td>-0.5</td>
<td>0.205</td>
<td>-12.7</td>
<td>0.065</td>
<td>-4.6</td>
<td>0.040</td>
<td>-2.5</td>
</tr>
<tr>
<td></td>
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<td>0.195</td>
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<td>0.038</td>
<td>-7.9</td>
</tr>
<tr>
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<td>0.216</td>
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<td>-4.6</td>
<td>0.038</td>
<td>-7.9</td>
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<tr>
<td></td>
<td>variable off</td>
<td>0.958</td>
<td>-0.5</td>
<td>0.216</td>
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<td>0.065</td>
<td>-4.6</td>
<td>0.038</td>
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</tr>
<tr>
<td></td>
<td>constant on</td>
<td>0.959</td>
<td>-0.4</td>
<td>0.216</td>
<td>-6.9</td>
<td>0.065</td>
<td>-4.6</td>
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<tr>
<td></td>
<td>constant off</td>
<td>0.958</td>
<td>-0.5</td>
<td>0.216</td>
<td>-6.9</td>
<td>0.065</td>
<td>-4.6</td>
<td>0.038</td>
<td>-7.9</td>
</tr>
</tbody>
</table>

Table 4 Reconstructed steady distortion metrics and discrepancy from RANS simulations. High offset S-duct at $M_{in}=0.27$.

<table>
<thead>
<tr>
<th>DENSITY</th>
<th>$\frac{\partial u_1}{\partial z}$</th>
<th>PR</th>
<th>ΔPR %</th>
<th>DC(60)</th>
<th>ΔDC(60) %</th>
<th>CDI</th>
<th>ΔCDI %</th>
<th>RDI</th>
<th>ΔRDI %</th>
</tr>
</thead>
<tbody>
<tr>
<td>RANS</td>
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<td>0.998</td>
<td>-</td>
<td>0.223</td>
<td>-</td>
<td>0.013</td>
<td>-</td>
<td>0.008</td>
<td>-</td>
</tr>
<tr>
<td>DSI</td>
<td>variable on</td>
<td>0.998</td>
<td>0</td>
<td>0.215</td>
<td>-3.7</td>
<td>0.013</td>
<td>0.0</td>
<td>0.008</td>
<td>0.0</td>
</tr>
<tr>
<td></td>
<td>variable off</td>
<td>0.998</td>
<td>0</td>
<td>0.201</td>
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<td>0.013</td>
<td>0.0</td>
<td>0.008</td>
<td>0.0</td>
</tr>
<tr>
<td>PPE</td>
<td>variable on</td>
<td>0.998</td>
<td>0</td>
<td>0.228</td>
<td>2.2</td>
<td>0.013</td>
<td>0.0</td>
<td>0.008</td>
<td>0.0</td>
</tr>
<tr>
<td></td>
<td>variable off</td>
<td>0.998</td>
<td>0</td>
<td>0.226</td>
<td>1.3</td>
<td>0.013</td>
<td>0.0</td>
<td>0.008</td>
<td>0.0</td>
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2. Unsteady reconstruction

Unsteady velocity and wall static pressure data from the DDES simulations was used to reconstruct the total pressure field and to calculate the total pressure ratio ($P_{0, AIP} / P_{0,in}$) as well as the pressure distortion descriptors for each time step. The DDES data was also used to calculate the same parameters based on the original CFD distributions of total pressure and thereby enable an assessment of the unsteady DSI reconstruction method. The distortion descriptors were evaluated based on a typical 8x5 rake resolution at the AIP (Fig. 9). Figure 10 and Fig. 11 show the comparison between the unsteady distortion metrics calculated directly from DDES data against those calculated from the DSI reconstructed unsteady total pressure fields for the S-duct at $M_{in}=0.6$ as a function of non-dimensional time.
defined as \( t/t_{\text{conv}} \). The accuracy of the reconstructed descriptors is quantified by means of the root mean square discrepancy from the DDES based values as:

\[
\Delta x_{\text{rms}} = \sqrt{\frac{\sum_{i=1}^{n}(\Delta x)^2}{n}}
\]

(23)

where \( n \) is the total number of snapshots and \( \Delta x \) defined in Eq. (22).

Figure 10 shows the temporal variation of the distortion descriptors from the original DDES data as well as for the DSI reconstructed total pressure fields with out-of-plane \( \left( \frac{\partial u}{\partial z} \right) \) and temporal gradient \( \left( \frac{\partial u}{\partial t} \right) \) terms included. PR, CDI and RDI reconstruction shows an rms discrepancy from the original DDES data of about 0.3\%, 4\% and 5\%, respectively (Fig. 10 a, c, d, Table 5). The DC(60) time history is slightly more underestimated by approximately 9\% (Fig. 10b). For this dataset, the PR exhibits very little variation with time step although there are more notable variations in the distortion descriptors particularly for DC60 and RDI. Overall, the reconstruction method clearly follows the temporal characteristics of the descriptors from CFD and captures the local maxima and minima with particularly good agreement in CDI and RDI (Fig. 10c, d).

The reconstructed unsteady pressure recovery and distortion descriptors using the DSI method are susceptible to uncertainties imposed by the out-of-plane \( \left( \frac{\partial u}{\partial z} \right) \) and temporal velocity \( \left( \frac{\partial u}{\partial t} \right) \) changes. These uncertainties are more pronounced in DC(60), CDI and RDI while PR is generally less affected by these terms. The errors for CDI and RDI increase to about 15\% and 10\%, respectively, while \( \Delta PR_{\text{rms}} \) remains below 1\% (Table 5). In addition, \( \Delta DC(60)_{\text{rms}} \) increases to approximately 30\% from about 9\% previously (Table 5, Fig. 11). When these terms are neglected, although the error increases, the reconstructed time history of the distortion descriptors still reproduces the main unsteady aspects and temporal changes in the metrics (Fig. 11).

A time-resolved, tomographic PIV system can potentially provide both spatial and temporal information that is required to apply the unsteady DSI method and therefore the uncertainties and characteristics highlighted in Fig. 10 and Table 5 could be expected for this type of flow field. However, a planar, temporally underresolved measurement system is less complicated and can provide potentially useful distortion assessments, albeit at a reduced accuracy. Overall, descriptor reconstruction with planar velocity data without the temporal derivatives included seems to be susceptible to around 30\% error in terms of DC(60) while CDI and RDI show a maximum discrepancy from the original DDES based values of approximately 15\%.
Table 5 Reconstructed unsteady distortion metrics and discrepancy from DDES data for the S-duct at $M_{in}=0.6$.

<table>
<thead>
<tr>
<th></th>
<th>$\frac{\partial u_i}{\partial z}$</th>
<th>$\frac{\partial u_i}{\partial t}$</th>
<th>$\Delta PR_{rms}$ %</th>
<th>$\Delta DC(60)_{rms}$ %</th>
<th>$\Delta CDI_{rms}$ %</th>
<th>$\Delta RDI_{rms}$ %</th>
</tr>
</thead>
<tbody>
<tr>
<td>DDES included</td>
<td>included</td>
<td>0.944</td>
<td>0.246</td>
<td>0.112</td>
<td>0.050</td>
<td>-</td>
</tr>
<tr>
<td>DSI on</td>
<td>on</td>
<td>0.940</td>
<td>0.34</td>
<td>9.2</td>
<td>3.6</td>
<td>0.048</td>
</tr>
<tr>
<td>DSI off</td>
<td>off</td>
<td>0.941</td>
<td>0.50</td>
<td>29.0</td>
<td>14.9</td>
<td>0.047</td>
</tr>
</tbody>
</table>

Fig. 10 Unsteady distortion metrics and discrepancy from DDES. Unsteady DSI with out-of-plane and time gradients enabled. (a): PR, (b): DC(60), (c): CDI and (d): RDI. S-duct with H/D$_a=2.44$ at $M_{in}=0.6$. 

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D. Effect of boundary conditions

For the reconstruction of the steady total pressure fields (Fig. 4, Fig. 5) a static pressure distribution was imposed along the outer boundary grid nodes of the domain (Fig. 1). This boundary condition was obtained from the numerical simulations and comprises approximately 200 points which is equivalent to the azimuthal computational grid resolution. For the reconstruction of the unsteady total pressure fields (Fig. 7 and Fig. 8) the boundary static pressure profile was updated for each time step from the respective unsteady DDES data. Hence, the unsteady total pressure field reconstruction was performed using a temporally synchronous static pressure boundary condition with the velocity field. However, it is pertinent to examine the behaviour of the pressure reconstruction algorithms in relation to the number and nature (steady or unsteady) of the static pressure imposed along the boundary of the domain. The outcome of this investigation quantifies the fidelity of the total pressure reconstruction methods when the boundary condition is experimentally measured from a number of circumferentially located wall static pressure tappings.
The effect of spatial resolution of the static pressure at the boundary on the steady total pressure distortion metrics is shown in Table 6 for the S-duct configuration at $M_{in}=0.6$. For these numerical experiments the circumferential static pressure distribution was approximated using 6, 12 and 18 samples from the same set of numerical simulations. A circumferential resolution of 6 wall static pressure values is insufficient while a spatial resolution of 18 points is required to sufficiently capture the local maximum (Table 6). For this example, PR, CDI and RDI are insensitive to the circumferential resolution. It has, however, a more notable impact on DC(60) which is overpredicted by around 10% with 6 points but underpredicted by -11% with 18 circumferential points.

<table>
<thead>
<tr>
<th>No. of boundary points</th>
<th>$PR$</th>
<th>ΔPR%</th>
<th>DC(60)</th>
<th>ΔDC(60)%</th>
<th>$CDI$</th>
<th>ΔCDI%</th>
<th>RDI</th>
<th>ΔRDI%</th>
</tr>
</thead>
<tbody>
<tr>
<td>6</td>
<td>0.958</td>
<td>-0.5</td>
<td>0.255</td>
<td>9.4</td>
<td>0.066</td>
<td>-3.0</td>
<td>0.04</td>
<td>-2.5</td>
</tr>
<tr>
<td>12</td>
<td>0.959</td>
<td>-0.4</td>
<td>0.212</td>
<td>-9.1</td>
<td>0.066</td>
<td>-3.0</td>
<td>0.04</td>
<td>-2.5</td>
</tr>
<tr>
<td>18</td>
<td>0.959</td>
<td>-0.4</td>
<td>0.208</td>
<td>-11.0</td>
<td>0.066</td>
<td>-3.0</td>
<td>0.04</td>
<td>-2.5</td>
</tr>
<tr>
<td>200 (DSI)</td>
<td>0.959</td>
<td>-0.4</td>
<td>0.210</td>
<td>-10.0</td>
<td>0.066</td>
<td>-3.0</td>
<td>0.04</td>
<td>-2.5</td>
</tr>
<tr>
<td>200 (RANS)</td>
<td>0.963</td>
<td></td>
<td>0.231</td>
<td>-</td>
<td>0.068</td>
<td>-</td>
<td>0.041</td>
<td></td>
</tr>
</tbody>
</table>

The behaviour of the unsteady DSI total pressure reconstruction method was assessed for different spatial resolution of static pressure tappings along the domain boundary. A variable density, unsteady DSI approach was used for the total pressure reconstructions with the out-of-plane $\left(\frac{\partial u}{\partial x}\right)$ and temporal velocity $\left(\frac{\partial u}{\partial t}\right)$ gradients enabled. The impact of the number of boundary static pressure points on the unsteady reconstruction accuracy was assessed for the S-duct at $M_{in}=0.6$. Steady as well as unsteady boundary static pressures were applied. The rationale for the latter is a scenario whereby synchronous, unsteady wall static pressure measurements are acquired along with the velocity data across the domain. The reconstructed AIP static pressure accuracy $\bar{I}$ (Eq. (18)) is insensitive to the circumferential resolution when steady static pressure is used and with a mean accuracy index $\bar{I}$ of 0.021 for this unsteady dataset. When unsteady wall static pressure data is used for the boundary condition, the accuracy improves and is sensitive to the spatial resolution. For this sample unsteady data, $\bar{I}$ reduces to 0.015 with 6 wall static data points and this improves.
monotonically to 0.01 when the spatial resolution increases to 18. This is asymptotically approaching the $\bar{T}$ level of 0.009 which was achieved when the full resolution of 200 data points was used (Table 2).

The accuracy of the reconstructed unsteady total pressure distortion descriptors was also evaluated in relation to the wall static pressure boundary conditions. The accuracy of the descriptor time series is expressed as the rms discrepancy from the time series calculated directly from unsteady DDES data using the entire unsteady static pressure profile around the domain as boundary condition. Overall, the rms error in PR, CDI and RDI for the unsteady data are relatively insensitive to the number of wall static points across the range of 6 to 18 (Table 7). The error in DC(60) is affected and the rms error reduces from about 19% to 10% when the spatial resolution is increased (Table 7). As anticipated, the number of boundary points has no impact on descriptor accuracy when applied in a time-averaged manner except for DC(60) where the discrepancy seems to reduce to about 10% for 18 points from about 19% for 6.

For the steady boundary condition case, Table 8 shows that all four parameters (PR, DC(60), CDI and RDI) demonstrate no dependency on the number of time-averaged boundary points with DC(60) manifesting the highest discrepancy from the DDES based values of around 30%. CDI and RDI show a similar discrepancy between 5-6% as for the unsteady boundary condition case. Finally, PR is not affected by the nature of the boundary condition applied as $\Delta PR_{\text{rms}}$ remains constant at <0.5%. Overall, although the unsteady static pressure profile seems to have a positive impact on the accuracy of the reconstructed flow field, the total pressure distortion metrics and the pressure recovery show small improvements. This is not the case for DC(60) reconstruction where an unsteady boundary pressure profile with a relatively large number of points seems to be essential for a credible DC(60) estimate.

### Table 7 Effect of unsteady boundary points on unsteady distortion metrics. Unsteady DSI with temporal and out-of-plane gradients enabled for the S-duct with $H/D_{\text{at}}=2.44$ at $M_{\text{n}}=0.6$.

<table>
<thead>
<tr>
<th>No. of boundary points</th>
<th>PR</th>
<th>$\Delta PR_{\text{rms}}$ %</th>
<th>DC(60)</th>
<th>$\Delta DC(60)_{\text{rms}}$ %</th>
<th>CDI</th>
<th>$\Delta CDI_{\text{rms}}$ %</th>
<th>RDI</th>
<th>$\Delta RDI_{\text{rms}}$ %</th>
</tr>
</thead>
<tbody>
<tr>
<td>6</td>
<td>0.941</td>
<td>0.44</td>
<td>0.221</td>
<td>18.6</td>
<td>0.108</td>
<td>7.1</td>
<td>0.048</td>
<td>6.3</td>
</tr>
<tr>
<td>12</td>
<td>0.940</td>
<td>0.41</td>
<td>0.223</td>
<td>14.8</td>
<td>0.108</td>
<td>5.0</td>
<td>0.047</td>
<td>6.1</td>
</tr>
<tr>
<td>18</td>
<td>0.941</td>
<td>0.41</td>
<td>0.226</td>
<td>10.1</td>
<td>0.109</td>
<td>5.3</td>
<td>0.048</td>
<td>6.1</td>
</tr>
<tr>
<td>200 (DSI)</td>
<td>0.940</td>
<td>0.34</td>
<td>0.224</td>
<td>9.2</td>
<td>0.109</td>
<td>3.6</td>
<td>0.048</td>
<td>4.7</td>
</tr>
<tr>
<td>200 (DDES)</td>
<td>0.944</td>
<td>-</td>
<td>0.246</td>
<td>-</td>
<td>0.112</td>
<td>-</td>
<td>0.050</td>
<td>-</td>
</tr>
</tbody>
</table>
Table 8 Effect of steady boundary points on unsteady distortion metrics. Unsteady DSI with temporal and out-of-plane gradients enabled for the S-duct with H/D_{in}=2.44 at M_{in}=0.6.

<table>
<thead>
<tr>
<th>No. of boundary points</th>
<th>( PR )</th>
<th>( \Delta PR_{rms} )</th>
<th>( DC(60) )</th>
<th>( \Delta DC(60)_{rms} )</th>
<th>( CDI )</th>
<th>( \Delta CDI_{rms} )</th>
<th>( RDI )</th>
<th>( \Delta RDI_{rms} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>6</td>
<td>0.941</td>
<td>0.42</td>
<td>0.224</td>
<td>30.6</td>
<td>0.109</td>
<td>5.2</td>
<td>0.047</td>
<td>6.2</td>
</tr>
<tr>
<td>12</td>
<td>0.941</td>
<td>0.42</td>
<td>0.222</td>
<td>31.7</td>
<td>0.109</td>
<td>5.2</td>
<td>0.047</td>
<td>6.2</td>
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<tr>
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<td>0.42</td>
<td>0.223</td>
<td>32.5</td>
<td>0.109</td>
<td>5.2</td>
<td>0.047</td>
<td>6.2</td>
</tr>
<tr>
<td>200 (DSI)</td>
<td>0.940</td>
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<td>3.6</td>
<td>0.048</td>
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<tr>
<td>200 (DDES)</td>
<td>0.944</td>
<td>-</td>
<td>0.246</td>
<td>-</td>
<td>0.112</td>
<td>-</td>
<td>0.050</td>
<td>-</td>
</tr>
</tbody>
</table>

IV. Conclusions

Inlet flow distortion assessments traditionally rely on flow data from low-response pressure probes of low spatial and temporal resolution to capture the unsteady nature of the flow at the exit of complex engine intakes. This limitation can be addressed by employing S-PIV methods that enable the acquisition of rich, synchronous velocity data of substantially higher spatial and potentially temporal resolution that can be used for the characterisation of the swirl distortion patterns at the outlet of the S-ducts. However, the need for AIP pressure data and pressure based distortion metrics still remains as PIV systems provide velocity measurements. This work assesses a number of methods to reconstruct AIP pressure fields using velocimetry data in an effort to further exploit the advantages offered by the application of S-PIV. The reconstructed pressure fields would allow inlet total pressure distortion assessments of equally high resolution as their complimentary swirl distortion data.

The reconstruction of the AIP total pressure field was performed using two approaches: a) a direct spatial integration of the momentum equation (DSI method) and b) the integration of the Poisson pressure equation (PPE method). The flow at the exit of the complex intake was considered inviscid while the Reynolds stress contribution to the equations of motion was neglected. Velocity data from numerical simulations was used to test the reconstruction of the total pressure field. The resultant total pressure fields were compared with the original complete CFD data at the exit of a complex intake with an offset to inlet diameter ratio H/D_{in} of 2.44 for inlet Mach numbers M_{in} between 0.27 and 0.6.
Time-averaged, DSI based reconstructions were within ±1% of the steady CFD total pressure field. The out-of-plane velocity gradients \( \left( \frac{\partial u_i}{\partial z} \right) \) have a modest impact on the accuracy of the reconstruction. Steady PPE based reconstructions are generally less dependent upon the out-of-plane velocity gradients while they were marginally closer to the original CFD data than the correspondent DSI obtained fields. The reconstruction of the steady total pressure ratio and distortion descriptors, CDI and RDI, were in good agreement with the CFD values and showed very little dependency upon the treatment of flow density, \( \left( \frac{\partial u_i}{\partial z} \right) \) velocity gradients \( \left( \frac{\partial u_i}{\partial z} \right) \) or the pressure reconstruction method used. However, the distortion metric DC(60) is mostly susceptible to the reconstruction method with PPE based values showing a 7% difference from the original CFD data. DSI based DC(60) showed a maximum discrepancy of 10% from the RANS based estimation which increased to approximately 20% when the out-of-plane velocity gradients were neglected. The DSI method was also used to reconstruct unsteady pressure fields and total pressure distortion metrics. DSI showed a generally robust behaviour albeit susceptible to loss of accuracy when \( \left( \frac{\partial u_i}{\partial z} \right) = 0 \) . The effect of the boundary condition on the AIP total pressure distribution is insensitive to the resolution of the wall static pressure for steady flow fields. However, the DSI pressure field reconstruction is dependent upon the number of circumferential boundary points for unsteady data with the main differences seen in the DC(60) parameter.

Overall, the reconstruction of pressure field and distortion metrics based on steady and unsteady computed velocity data was assessed. The relative importance of temporal and out-of-plane velocity gradients on the accuracy of each method was examined. The main purpose of these studies is to enable the exploitation of synchronous velocimetry available at the exit of complex engine intakes in order to characterise their unsteady pressure distortion metrics at a similar spatial and temporal resolution. The studies showed that reconstruction of AIP pressure fields and pressure distortion metrics based on velocity information is possible for these types of flows.

References


“Evaluation of Integral Forces and Pressure Fields from Planar Velocimetry Data for Incompressible and Compressible Flows,”