



ASME 2013 MAY 22-24, 2013 • LAS VEGAS, NEVADA
VERIFICATION AND VALIDATION SYMPOSIUM

Solution Verification based on Grid Refinement Studies and Power Series Expansions

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1. Motivation

- CFD simulations require the assessment of their numerical uncertainty to establish their credibility
- Most of the existing methods for uncertainty estimation require data in the so-called “asymptotic range”
- This often means levels of grid refinement beyond those normally used in practical applications



1. Motivation

- Uncertainty estimation for practical calculations as to deal with several difficulties:
 - Grids not sufficiently refined to attain the “asymptotic range”
 - Scatter in the data originated by lack of geometric similarity, switches in the turbulence models, post-processing...
- Develop a reliable procedure for uncertainty estimation of practical calculations



2. Proposed procedure

- Power series expansions applied to data obtained from grid refinement studies
- Contributions of round-off and iterative errors are assumed to be negligible when compared to the discretization error
- Definition of the typical cell size, h_i , discussed at the ASME V&V 2012 Conference



2. Proposed procedure

- Error estimation

$$\mathcal{E}_\phi \cong \delta_{RE} = \phi_i - \phi_o = \alpha h_i^p$$

δ_{RE} Estimated error of variable ϕ (integral or local)

ϕ_i Solution of variable ϕ in grid i

ϕ_o Estimated exact solution of ϕ

α Grid related constant

h_i Typical cell size

p Observed order of grid convergence



2. Proposed procedure

- Alternative expansions for error estimation

$$\delta_1 = \phi_i - \phi_o = \alpha h_i$$

$$\delta_2 = \phi_i - \phi_o = \alpha h_i^2$$

$$\delta_{12} = \phi_i - \phi_o = \alpha_1 h_i + \alpha_2 h_i^2$$



2. Proposed procedure

- All error estimators are solved in the least-squares sense

$$S_{RE}(\phi_o, \alpha, p) = \sqrt{\sum_{i=1}^{n_g} (\phi_i - (\phi_o + \alpha h_i^p))^2}, \quad \frac{\partial S_{RE}}{\partial \phi_o} = 0, \quad \frac{\partial S_{RE}}{\partial \alpha} = 0, \quad \frac{\partial S_{RE}}{\partial p} = 0$$

$$S_1(\phi_o, \alpha) = \sqrt{\sum_{i=1}^{n_g} (\phi_i - (\phi_o + \alpha h_i))^2}, \quad \frac{\partial S_1}{\partial \phi_o} = 0, \quad \frac{\partial S_1}{\partial \alpha} = 0$$

$$S_2(\phi_o, \alpha) = \sqrt{\sum_{i=1}^{n_g} (\phi_i - (\phi_o + \alpha h_i^2))^2}, \quad \frac{\partial S_2}{\partial \phi_o} = 0, \quad \frac{\partial S_2}{\partial \alpha} = 0$$

$$S_{12}(\phi_o, \alpha) = \sqrt{\sum_{i=1}^{n_g} (\phi_i - (\phi_o + \alpha_1 h_i + \alpha_2 h_i^2))^2}, \quad \frac{\partial S_{12}}{\partial \phi_o} = 0, \quad \frac{\partial S_{12}}{\partial \alpha_1} = 0, \quad \frac{\partial S_{12}}{\partial \alpha_2} = 0$$



2. Proposed procedure

- Observed order of grid convergence p and standard deviations σ of the fits are used to select error estimator

$$\sigma_{RE} = \sqrt{\frac{\sum_{i=1}^{n_g} (\phi_i - (\phi_o + \alpha h_i^p))^2}{(n_g - 3)}}, \quad \sigma_1 = \sqrt{\frac{\sum_{i=1}^{n_g} (\phi_i - (\phi_o + \alpha h_i))^2}{(n_g - 2)}}$$
$$\sigma_2 = \sqrt{\frac{\sum_{i=1}^{n_g} (\phi_i - (\phi_o + \alpha h_i^2))^2}{(n_g - 2)}}, \quad \sigma_{12} = \sqrt{\frac{\sum_{i=1}^{n_g} (\phi_i - (\phi_o + \alpha_1 h_i + \alpha_2 h_i^2))^2}{(n_g - 3)}}$$



2. Proposed procedure

- Alternative weighted fits also performed with

$$w_i = \frac{1}{h_i} \frac{1}{\sum_{i=1}^{n_g} \frac{1}{h_i}}$$

that lead, for example, to

$$S_{RE}^w(\phi_o, \alpha, p) = \sqrt{\sum_{i=1}^{n_g} w_i (\phi_i - (\phi_o + \alpha h_i^p))^2}$$
$$\sigma_{RE}^w = \sqrt{\frac{\sum_{i=1}^{n_g} n_g w_i (\phi_i - (\phi_o + \alpha h_i^p))^2}{(n_g - 3)}}$$



2. Proposed procedure

- Monotonic convergence for $p > 0$ or $p^w > 0$

- $0.5 \leq p \leq 2 \Rightarrow \varepsilon_\phi = \begin{cases} \delta_{RE} \leftarrow \sigma_{RE} \leq \sigma_{RE}^w \\ \delta_{RE}^w \leftarrow \sigma_{RE} > \sigma_{RE}^w \end{cases}$
- $p > 2 \Rightarrow \varepsilon_\phi = \delta$ of fit with $\sigma = \min(\sigma_1, \sigma_1^w, \sigma_2, \sigma_2^w)$
- $p < 0.5 \Rightarrow \varepsilon_\phi = \delta$ of fit with $\sigma = \min(\sigma_1, \sigma_1^w, \sigma_2, \sigma_2^w, \sigma_{12}, \sigma_{12}^w)$



2. Proposed procedure

- Anomalous behaviour

- $p < 0$ and $p^w < 0$ or impossible to determine
- Error may be estimated from fits with fixed exponents
- $\varepsilon_\phi = \delta$ of fit with $\sigma = \min(\sigma_1, \sigma_1^w, \sigma_2, \sigma_2^w, \sigma_{12}, \sigma_{12}^w)$



2. Proposed procedure

- Uncertainty estimation intends to satisfy

$$\phi_i - U_\phi \leq \phi_{exact} \leq \phi_i + U_\phi$$

95% of the times

- Quality of available data measured from

$$\Delta_\phi = \frac{(\phi_i)_{\max} - (\phi_i)_{\min}}{n_g - 1}$$

and σ of the selected fit for error estimation



2. Proposed procedure

- Uncertainty is obtained from the estimated error ε_ϕ
and a safety factor F_s

$$U_\phi = F_s \varepsilon_\phi$$

- Definition of the safety factor F_s follows the G.C.I of P.J. Roache

- Reliable error estimate, $F_s = 1.25 \Leftrightarrow \begin{cases} 0.5 \leq p \leq 2.1 \wedge \sigma_{RE} < \Delta_o \\ 0.5 \leq p^w \leq 2.1 \wedge \sigma_{RE}^w < \Delta_o \end{cases}$
- Otherwise, $F_s = 3$



2. Proposed procedure

- “Good” error estimation, $\sigma < \Delta_o$

$$U_\phi = F_s \varepsilon_\phi + \sigma + |\phi_i - \phi_{fit}|$$

- “Bad” error estimation, $\sigma \geq \Delta_o$

$$U_\phi = 3 \frac{\sigma}{\Delta_o} (\varepsilon_\phi + \sigma + |\phi_i - \phi_{fit}|)$$

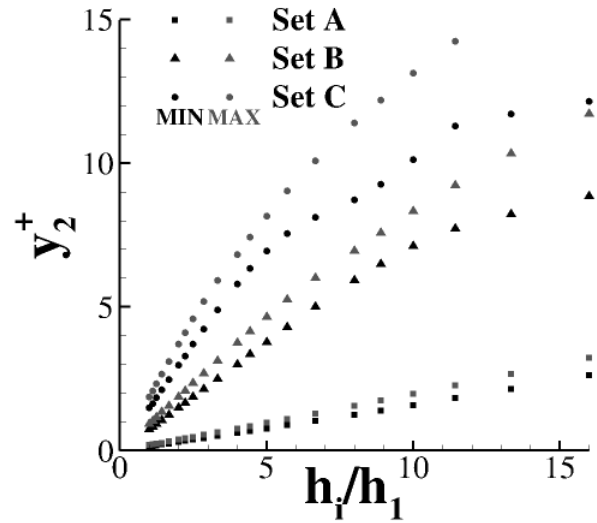
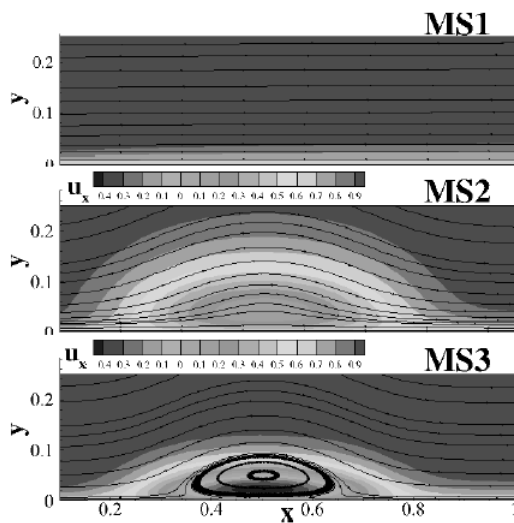
- For data without noise and monotonic convergence the method reduces to the G.C.I.



3. Examples of application

1. Two-dimensional Manufactured Solutions that mimic a near-wall turbulent flow
2. Turbulent flow over a flat plate
3. Flow over a backward facing step
4. Flow around a tanker at model scale Reynolds number

3. Examples of application, 2-D MS's



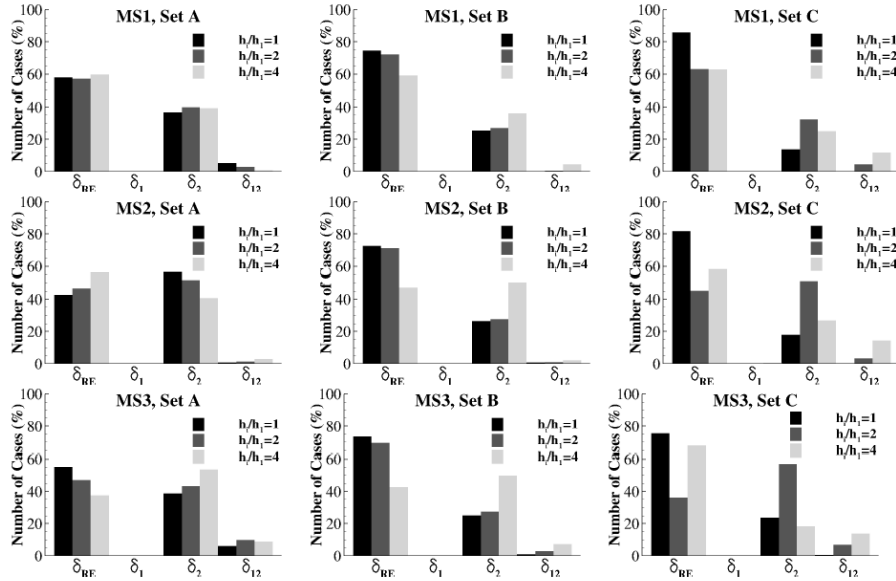
3. Examples of application, 2-D MS's

- Sets of 21 geometrically similar grids with different near-wall spacing
- Uncertainty of mean flow quantities estimated for three levels of grid refinement:

$$\frac{h_i}{h_1} = 1 \text{ (801} \times \text{801)}, \frac{h_i}{h_1} = 2 \text{ (401} \times \text{401)}, \frac{h_i}{h_1} = 4 \text{ (201} \times \text{201)}$$
- Check $F_e = U_\phi / e_\phi$ and the error estimate used to obtain the numerical uncertainty

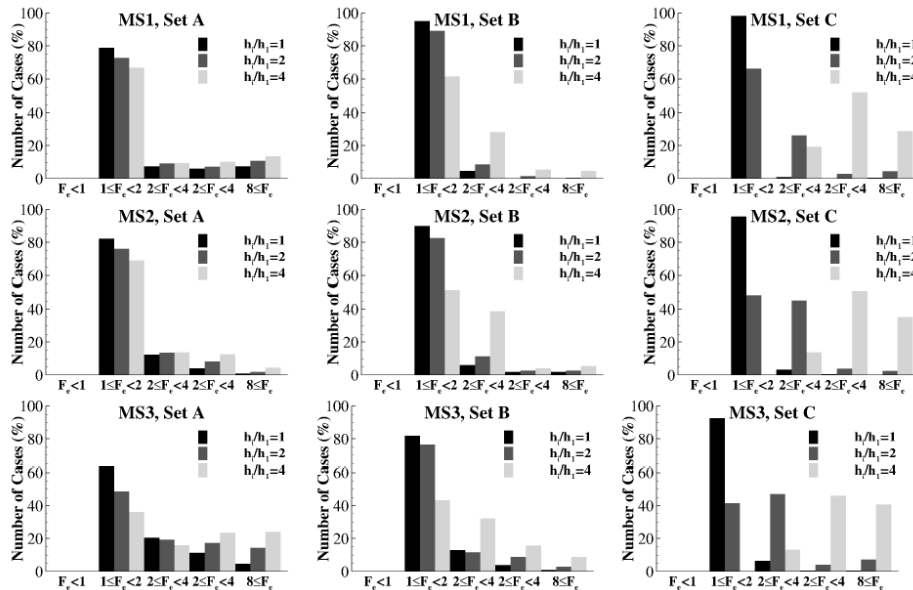
3. Examples of application, 2-D MS's

U_x



3. Examples of application, 2-D MS's

$U_x \quad F_e = \frac{U_\phi}{e_\phi}$



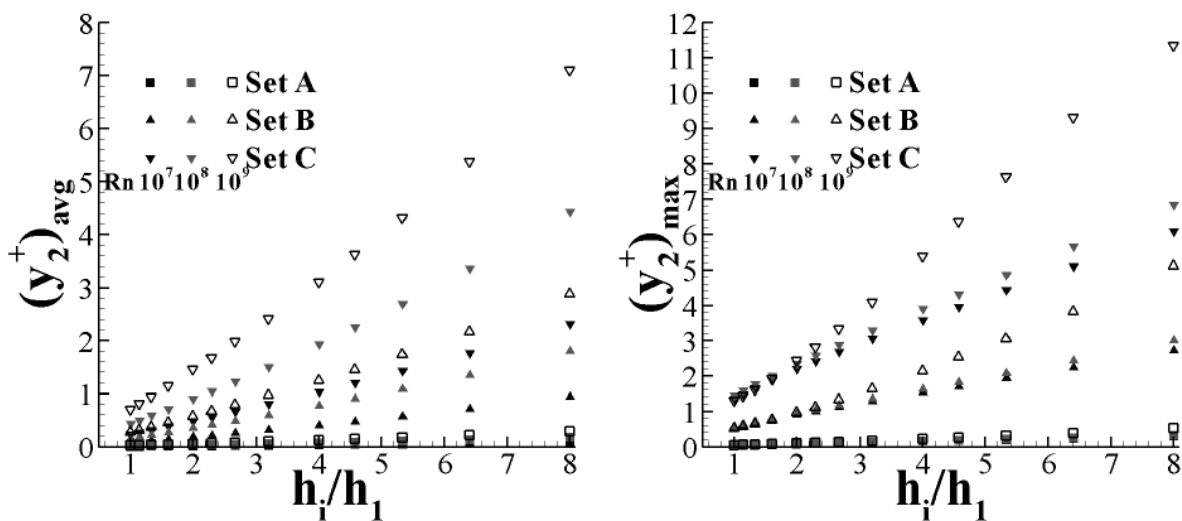
3. Examples of application, Flat plate

- Sets of 13 geometrically similar grids with different near-wall spacing for Reynolds numbers of 10^7 , 10^8 and 10^9
- Spalart & Allmaras and SST $k-\omega$ turbulence models
- Uncertainty of mean flow, turbulence and integral quantities estimated for three levels of grid refinement:

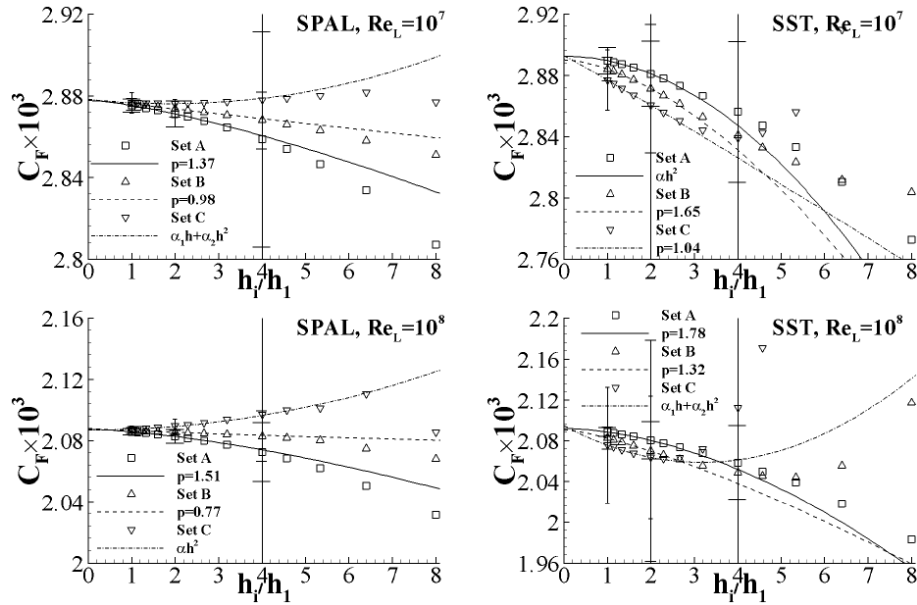
$$\frac{h_i}{h_1} = 1 \text{ (1537} \times \text{193)}, \frac{h_i}{h_1} = 2 \text{ (769} \times \text{97)}, \frac{h_i}{h_1} = 4 \text{ (385} \times \text{49)} \quad Rn = 10^7$$

- Check overlap of error bars and the error estimate used to obtain the numerical uncertainty

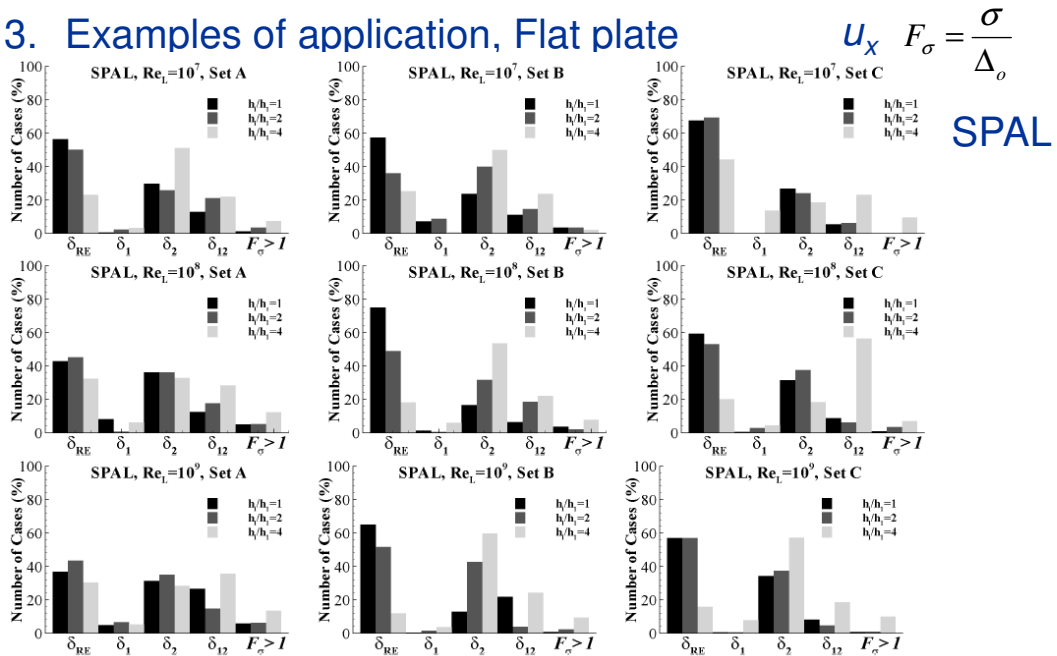
3. Examples of application, Flat plate



3. Examples of application, Flat plate



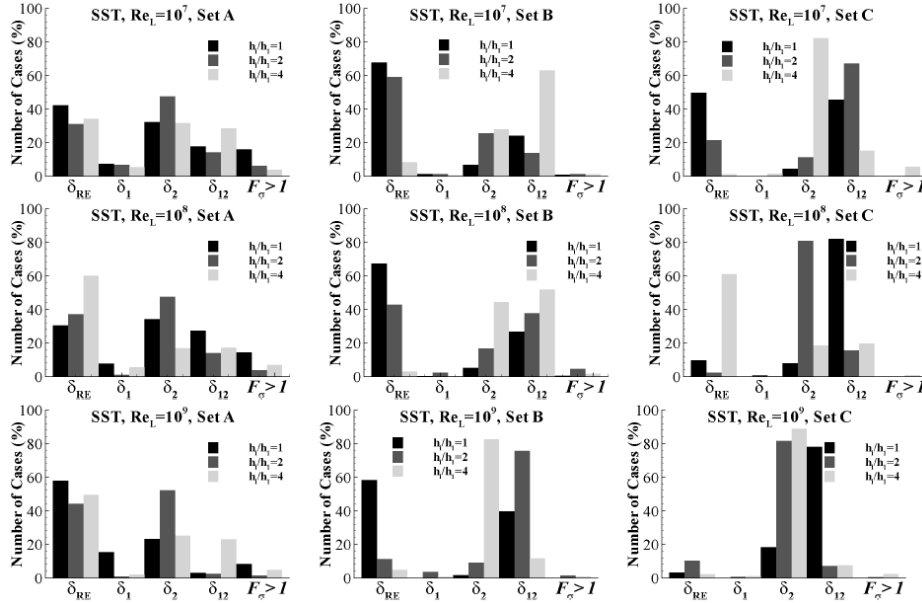
3. Examples of application, Flat plate



3. Examples of application, Flat plate

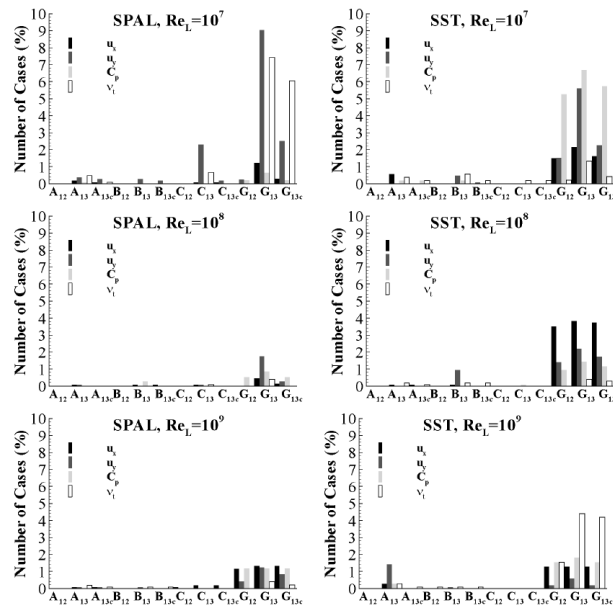
$$U_x F_\sigma = \frac{\sigma}{\Delta_o}$$

SST



3. Examples of application, Flat plate

Inconsistent error bars
 (non overlapping)



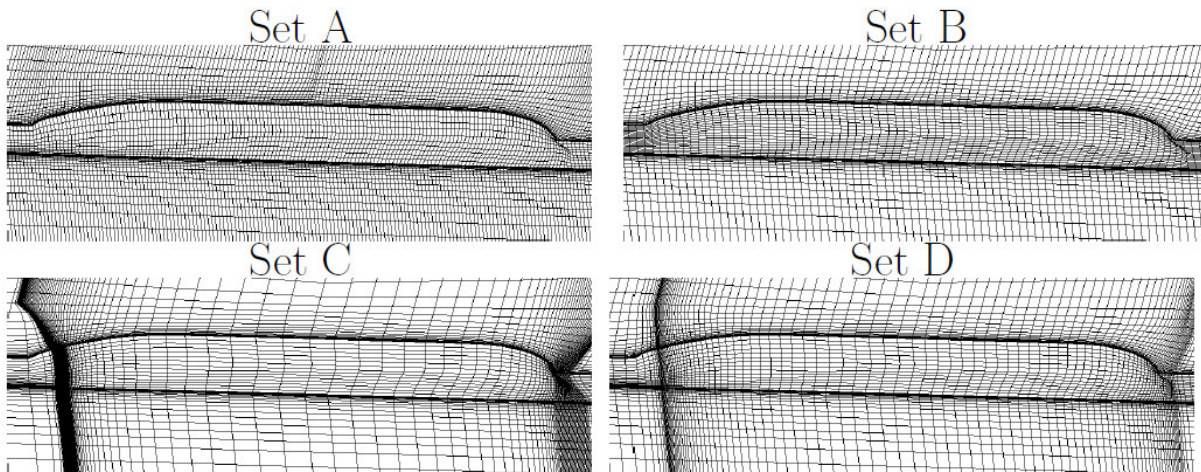


3. Examples of application, Tanker $Rn = 4.6 \times 10^6$

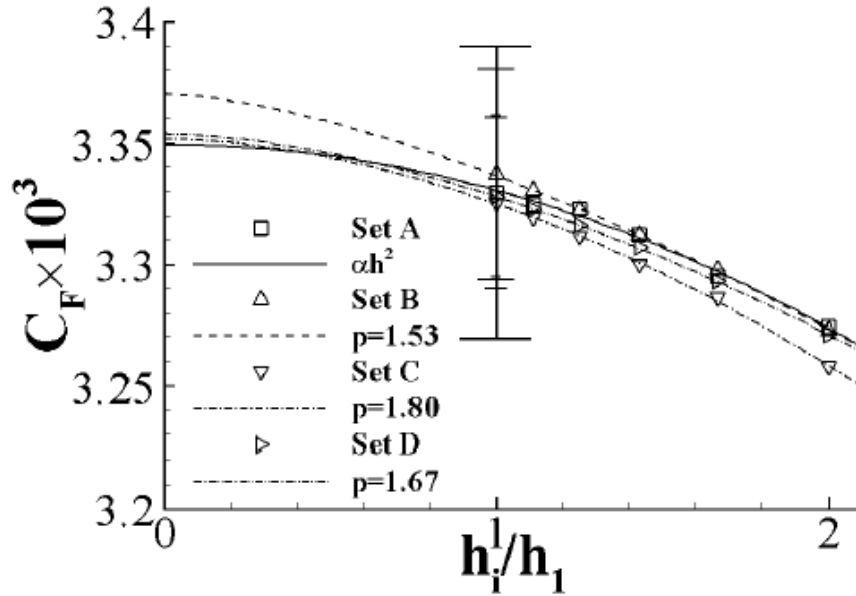
- 4 sets of 6 nearly-geometrically similar grids with 0.9×10^6 to 8×10^6 cells covering a grid refinement ratio of 2
- SST $k-\omega$ turbulence model
- Uncertainty of mean and turbulence flow quantities at the propeller plane and resistance coefficients
- Check overlap of error bars and the error estimate used to obtain the numerical uncertainty



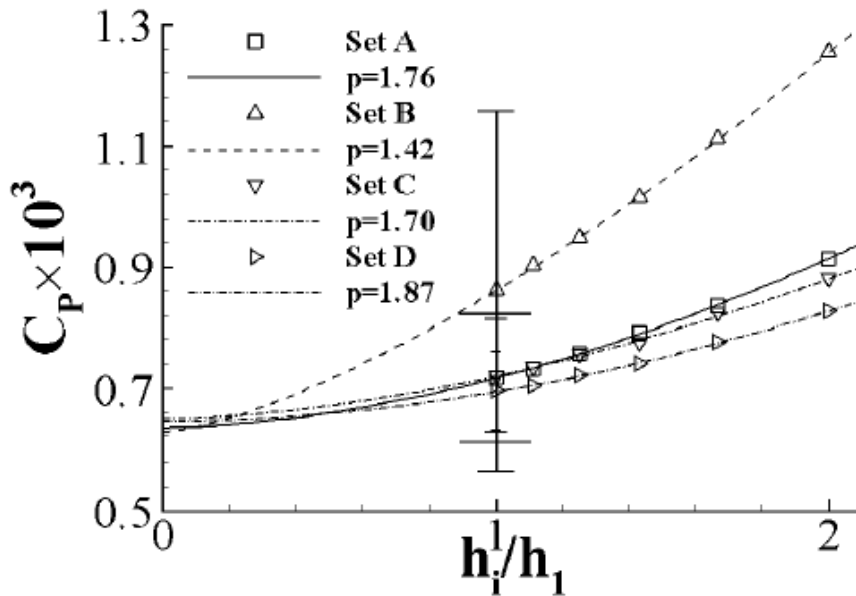
3. Examples of application, Tanker $Rn = 4.6 \times 10^6$



3. Examples of application, Tanker $Rn = 4.6 \times 10^6$



3. Examples of application, Tanker $Rn = 4.6 \times 10^6$





4. Final Remarks

- We presented a procedure for the estimation of the numerical uncertainty of any integral or local flow quantity based on grid refinement studies
- The uncertainty is based on an error estimation multiplied by a safety factor
- The error is estimated with power series expansions as a function of the typical cell size, which are fitted to the data in the least squares sense



4. Final Remarks

- Several alternative formulations are involved, including weighted and non-weighted fits of expressions with different exponents in the leading term of the series
- The selection of the best error estimate is based on the standard deviation of the fits
- For well-behaved data sets, i.e. monotonic convergence with the expected observed order of accuracy and no scatter in the data, the method reduces to the well known Grid Convergence Index