

Recent Developments and Applications of Bayesian Data Analysis in Fusion Science: Integrated Data Analysis

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Fusion Science



- physical basis of a fusion power plant
- like the sun, such a plant is to generate energy from fusion of atomic nuclei (E=mc²)
- > a plasma is a hot (100 Mio °C) ionized gas (hydrogen and deuterium)
- The physical properties are studied experimentally and theoretically
- The link between theory and experiment is given by Data Analysis



ASDEX Upgrade with divertor I



ASDEX Upgrade discharge #4496



Diagnosing the plasma





cooling water calorimetry

HXR: 4 detectors at walls of the experimental hall

IDA for Nuclear Fusion



Different measurement techniques (diagnostics: LIB, DCN, ECE, TS, REF, ...) for the same quantities (n_e, T_e, ...) and parametric entanglement in data analysis

- Redundant data:
 - reduction of estimation uncertainties (combined evaluation, "super fit")
 - detect and resolve data inconsistencies (reliable/consistent diagnostics)

Complementary data:

- ➢ resolve parametric entanglement
- resolve complex error propagation (non-Gaussian)
- synergistic effects (parametric correlations, multi-tasking tools (TS/IF, CXRS/BES))
- > automatic in-situ and in-vivo calibration (transient effects, degradation, ...)
- Goal: Coherent combination of measurements from different diagnostics
 ➢ replace combination of results from individual diagnostics
 ➢ with combination of measured data → one-step analysis of pooled data
 - in a probabilistic framework (unified error analysis!)





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Drawbacks of conventional data analysis: iterative

- (self-)consistent results?
- difficult to be automated
- information propagation?
- data and result validation?

- (cumbersome; do they exist?)
- (huge amount of data from steady state devices: W7X, ITER, ...)(Single estimates as input for analysis of other diagnostics?)(How to deal with inconsistencies?)
- non-Gaussian error propagation? (frequently neglected: underestimation of the true error?)
- often backward inversion techniques (noise fitting? numerical stability? loss of information?)
- result: estimates and error bars (sufficient? non-linear dependencies?)

Probabilistic combination of different diagnostics (IDA)

- \checkmark uses only forward modeling (complete set of parameters \rightarrow modeling of measured data)
- additional physical information easily to be integrated
- $\boldsymbol{\checkmark}$ systematic effects \rightarrow nuisance parameters
- $\boldsymbol{\checkmark}$ unified error interpretation \rightarrow Bayesian Probability Theory
- ✓ result: probability distribution of parameters of interest

IDA offers a unified way of combining data (information) from various experiments (sources) to obtain improved results

Probabilistic (Bayesian) Recipe



Reasoning about parameter θ :

(uncertain) prior information

- + (uncertain) measured data
- + physical model

 $\begin{array}{l} p(\theta) & prior \text{ distribution} \\ d = D + \epsilon \\ D = f(\theta) \end{array} \Big\} p(d|\theta) & likelihood \text{ distribution} \end{array}$

+ Bayes theorem

$$p(\theta|d) = \frac{p(d|\theta) \times p(\theta)}{p(d)}$$

posterior distribution

+ additional (nuisance) parameter β

$$p(\theta|d) = \int d\beta \, p(\theta, \beta|d)$$

$$= \int d\beta \frac{p(d|\theta, \beta) \times p(\theta) \times p(\beta)}{p(d)}$$
marginalization (integration)

generalization of Gaussian error propagation laws

+ parameter averaging (model comparison)

 $p(d|M) = \int d\theta p(\theta, d|M) = \int d\theta p(d|\theta, M) p(\theta)$ prior predictive value

Integrated Data Analysis





 n_e , T_e : Thomson scattering, interferometry, soft X-ray

Using interdependencies: Combination of results from a set of diagnostics (W7-AS)



Probabilistic framework

R. Fischer, A. Dinklage, and E. Pasch, Bayesian modelling of fusion diagnostics, PPCF, 45, 1095-1111 (2003)



Using synergism: Combination of results from a *set* of diagnostics



full probabilistic correlation structure





- Lithium beam impact excitation spectroscopy (LIB)
- Interferometry measurements (DCN)
- Electron cyclotron emission (ECE)
- Thomson scattering (TS)
- Reflectometry (REF)
- Equilibrium reconstructions for diagnostics mapping

R. Fischer et al., Integrated data analysis of profile diagnostics at ASDEX Upgrade, Fusion Sci. Technol., 58, 675-684 (2010)

(2) Z_{eff} :

- Bremsstrahlung background from various \geq CXRS spectroscopies
- Impurity concentrations from CXRS \geq

S. Rathgeber et al., Estimation of profiles of the effective ion charge at ASDEX Upgrade with Integrated Data Analysis, PPCF, 52, 095008 (2010)



Bayesian Recipe for IDA: LIB+DCN+ECE+TS+REF

Reasoning about parameter n_e , T_e :

(uncertain) prior information $p(n_e, T_e)$ prior distribution+ experiment 1: $d_{LiB} = D_{LiB}(n_e, T_e) + \epsilon$; $p(d_{LiB}|n_e, T_e)$ prior distribution+ experiment 2: $d_{DCN} = D_{DCN}(n_e) + \epsilon$; $p(d_{DCN}|n_e)$ likelihood+ experiment 3: $d_{ECE} = D_{ECE}(T_e) + \epsilon$; $p(d_{ECE}|T_e)$ likelihood+ experiment 4: $d_{TS} = D_{TS}(n_e, T_e) + \epsilon$; $p(d_{REF}|n_e, T_e)$ $p(d_{REF}|n_e)$ + experiment 5: $d_{REF} = D_{REF}(n_e) + \epsilon$; $p(d_{REF}|n_e)$ $p(d_{REF}|n_e)$

posterior

distribution

+ Bayes theorem $p(n_{e}, T_{e}|d_{TS}, d_{ECE}, d_{LiB}, d_{DCN}, d_{REF}) \propto p(d_{TS}|n_{e}, T_{e}) \times p(d_{ECE}|T_{e}) \times p(d_{LiB}|n_{e}, T_{e}) \times p(d_{LiB}|n_{e}, T_{e}) \times p(d_{DCN}|n_{e}) \times p(d_{REF}|n_{e}) \times p(d_{REF}|n_{e}) \times p(n_{e}, T_{e})$

Lithium-beam diagnostic





Lithium beam impact excitation spectroscopy (LiI: Li(2p) \rightarrow Li(2s), λ = 670.8 nm) measured Lil radiation from modeled line intensity (arb. units) N neutral Lithium (E=30-80 keV) $D_{LiB}[n_e(x_{LiB})]$ 0 5 10 0 15 beam axis [cm] System of coupled linear differential equations: $\frac{dN_i(x)}{dx} = \sum_{i=1}^{N_{Li}} \{n_e(x)a_{ij}(T_e(x)) + b_{ij}\}N_j(x) ; N_i(x=0) = \delta_{1i}$

solved for a given profile $n_e(x)$ to obtain occupation density Li(2p): $N_2(x|n_e)$

$$D_{LiB}[n_e(x_i)] = \alpha s_i N_2(x)$$
; $i = 1,...,35$

Lithium-beam diagnostic: Comparison old-new data analysis



comparison of profiles from old data analysis and new probabilistic method





- Old: density profile stops just before pedestal (temporal binning: 20 ms)
- → New: Pedestal well determined
- New: high reliability of profile for $\rho_{pol} > 0.93$
- New: Temporal resolution: 50 μs

R. Fischer, E. Wolfrum, J. Schweinzer, 34th EPS Conference on Controlled Fusion and Plasma Physics. 2007

LIB: Profile Uncertainties



•Uncertainties are necessary to evaluate the quality of the estimated profile error bands, confidence intervals, ...

- \rightarrow how to be estimated? \rightarrow different techniques
- \rightarrow for what purpose? \rightarrow use the error bars for further analysis
- → interpretation?

- \rightarrow what do error bars tell us at all?
- \rightarrow correlations (p_e = n_e T_e)?



R. Fischer, E. Wolfrum, J. Schweinzer, PPCF 50 (2008) 085009



#212416



R. Fischer et al., Integrated data analysis of profile diagnostics at ASDEX Upgrade, Fusion Sci. Technol., 58, 675-684 (2010)

LIB + DCN Interferometry







- full density profiles
- smooth profile with steep edge
- uncertainties:
 - → ρ_{pol} > 0.95: Li-Beam
 - → ρ_{pol} < 0.95: DCN (5 LOS)

LIB + DCN: Temporal resolution



#22561, 2.045-2.048 s, H-mode, type I ELM



LIN: Lithium beam only

IDA: Lithium beam + DCN Interferometry

density profiles with temporal resolution of 1 ms (routinely)

R. Fischer, E. Wolfrum, J. Schweinzer, PPCF 50 (2008) 085009

IDA: ELM resolved profiles





- temporal resolution 50 μs (new in 2012: 5 μs)
- no ELM averaging \rightarrow single ELMs
- no correlation in data analysis of neighboring time frames!

R. Fischer, E. Wolfrum, Ch. Fuchs, 35th EPS Conference on Controlled Fusion and Plasma Physics. 2008





R. Fischer et al., Integrated data analysis of profile diagnostics at ASDEX Upgrade, Fusion Sci. Technol., 58, 675-684 (2010)



IDA: LIB + DCN + ECE











IDA: LIB + DCN + ECE + TS



- → profiles from different diagnostics (TS, LIB, ECE) to not perfectly match at plasma edge
- → 2 additional parameters: scale of the LIB " ρ_{nol} "
 - $\rho_{\text{pol,ECE}} = s_{\text{ECE}}^* \rho_{\text{pol,LiB}} \rightarrow \Delta \rho_{\text{pol,ECE}}$ rel. shift of ECE channels at plasma edge
 - $\rho_{_{pol,TS}} = s_{_{TS}} * \rho_{_{pol,LiB}} \rightarrow \Delta \rho_{_{pol,TS}}$ rel. shift of TS channels at plasma edge





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Reflectometry

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- Goal: n_e profiles, plasma position control (ITER)
- Classical analysis: Abel inversion (O-mode)
 → location of cutoff layer



- Problems:
 - unphysical profiles
 - multiple analysis steps (phase of reflected wave \rightarrow group delay \rightarrow density)
 - error treatment/propagation; profile uncertainties
 - density initialization outside first cutoff layer

• IDA

- Forward modeling of measured data for given density profile
- Benefit:
 - Additional data available \rightarrow density initialization
 - Alignment

Reflectometry forward modelling





Refractive index:

$$\mu(r) = \sqrt{1 - \frac{n(r)}{n_c(f)}}; \quad n_c(f) = \frac{4\pi^2 \epsilon_0 m_e f^2}{e^2}$$

Forward model for group delay for a given density profile:

$$\tau(f, n(r)) = \frac{2}{c} \int_{0}^{\sqrt{r_{c} - r_{1}}} \frac{2x}{\sqrt{1 - \frac{n(r_{c} - x^{2})}{n_{c}}}} dx$$

IDA: LIB + DCN + Reflectometry







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IDA and the Magnetic Equilibrium



- ➢ Combine profile diagnostics LIB, DCN, ECE, TS, REF → n_e and T_e profile fits to all data at once
- > Mapping on a common coordinate grid using an existing equilibrium (EQH/EQI/FPP)
- Inconsistency: Equilibrium is not evaluated with kinetic profiles from IDA
 - *x* Position of magnetic axis, separatrix, inner flux surfaces?
 - *x* DCN: H2-H3 vertical plasma position often seems to be wrong up to ~1cm.
 - x ECE: (r,z) depends on equilibrium
 - x TS: vertical system relies very much on equilibrium
 - x Alignment of TS, ECE, LIB (with separatrix T_e) \rightarrow uncertainties in the equilibrium ???
- Goal: combine data from profile diagnostics with magnetic data
 - for a joint estimation of profiles and the magnetic equilibrium
- > Needs equilibrium code:
 - × CLISTE very successful, but code too sophisticated to be adapted to the IDA code
 - x New code based on the ideas (success) of CLISTE
 - (P. McCarthy, L. Giannone, P. Martin, K. Lackner, S. Gori)
 - *x* Extra: Parallel Grad-Shafranov solver \rightarrow real-time equilibrium (~100 µs)

(R. Preuss, M. Rampp, K. Hallatschek, L. Giannone)

Grad-Shafranov solver

Grad-Shafranov equation: Ideal magnetohydrodynamic equilibrium for

poloidal flux function $\boldsymbol{\varPsi}$ for axisymmetric geometry

$$\left(R\frac{\partial}{\partial R}\frac{1}{R}\frac{\partial}{\partial R} + \frac{\partial^2}{\partial z^2}\right)\Psi = -(2\pi)^2\mu_0\left(R^2P' + \mu_0FF'\right)$$

- 0.8 0.6 ⇒ 0.4 - EQH 0.2 #25764, 2.0s 0 0.5 -0.5 0 z [m] EQH IDE (L) Z 0.8 0.6 ∋ 0.4 - EQH 0.2 #25764, 2.0s 0 1.5 2 R [m] 1.5 2.0 2.5 1.0 R(m)
- New code developed capable to be used in the IDA concept:
- Magnetic measurements
- ✓ Profiles from other diagnostics
- ✓ Real-time solver (~100 µs)

R. Preuss et al., IPP-Report R/47 (2012)M. Rampp et al., Fusion Science and Technol., accepted



Comparison EQH/IDE: Temperature and density

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Summary: IDA



Probabilistic modeling of individual diagnostics

- forward modeling only (synthetic diagnostic)
- ✓ probability distributions: describes all kind of uncertainties
- ✓ multiply probability distributions, marginalization of nuisance parameters
- ✓ parameter estimates and uncertainties

Probabilistic combination of different diagnostics

- ✓ systematic and unified error analysis is a must for comparison of diagnostics
- error propagation beyond single diagnostics
- more reliable results by larger (meta-) data set (interdependencies, synergism)
- $\boldsymbol{\checkmark}$ redundant information \rightarrow resolve data inconsistencies
- $\boldsymbol{\checkmark}$ advanced data analysis technique \rightarrow software/hardware upgrades

Applications at W7-AS (Garching, Germany), JET (Culham, UK), TJ-II (Madrid, Spain), and ASDEX Upgrade (Garching, Germany), W7-X (Greifswald, Germany), Tore Supra (Cadarache, France)