

# Automatic search for optimal models using Levenberg-Marquardt algorithm

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23 May 2016

# Numerical codes

## Internal structure and evolution code

- Cesam2k (*Morel & Lebreton, 2008*)  
if rotation: cestam (*Marques et al., 2013*), not used in H&H
- options for input physics and parameters, e.g. *Lebreton & Goupil (2014)*
  - solar mixture: GN93 (Grevesse & Noels 93), AGSS09 (Asplund et al. 09)
  - EoS: OPAL 2005 (update of Rogers & Nayfonov 02)
  - opacities: OPAL96 (Iglesias & Rogers 96), Wichita (Ferguson et al. 05)
  - nuclear rates: NACRE (Angulo et al. 99), LUNA (Formicola et al. 04, for e.g.  $^{14}N(p, \gamma)^{15}O$ )
  - convection: MLT (Böhm-Vitense, 65), CGM (Canuto et al. 96)
  - atomic diffusion: Michaud & Profitt 93, Burgers 69
  - overshooting  $d_{ov} = \min(\alpha_{ov} \times H_p, \alpha_{ov} \times R_{cc})$
  - atmosphere: T( $\tau$ ) laws ; Eddington (grey), Hopf, Kurucz, MARCS, etc.
  - conduction: Pothekin et al. 05
- compared to ASTEC, CLES, GARSTEC, STARROX by the ESTA CoRoT team (Monteiro et al. 06, Lebreton et al. 08a,b).

# Numerical codes

## Oscillation codes (adiabatic)

- LOSC (*Scuflaire et al., 2008*)
- GYRE (*Townsend & Teitler, 2014*)

# The Levenberg-Marquardt method

- Newton algorithm

- Based on a 2<sup>nd</sup>-order Taylor expansion of  $\chi^2$  around current location
- Exact for quadratic functions
- Efficient to find a close minimum, inefficient to approach a distant one

$$\delta \mathbf{a} = -\mathbf{H}^{-1} \cdot \nabla \chi^2$$

Hessian matrix

$$H_{kl} \equiv \frac{\partial^2 \chi^2}{\partial a_k \partial a_l} = 2 \sum_{i=1}^N \left[ \frac{1}{\sigma_i^2} \frac{\partial y_i(\mathbf{a})}{\partial a_k} \frac{\partial y_i(\mathbf{a})}{\partial a_l} \right] \frac{y_i - y_i(\mathbf{a})}{\sigma_i^2} \frac{\partial^2 y_i(\mathbf{a})}{\partial a_k \partial a_l}$$

Gradient

$$\nabla_k \chi^2(\mathbf{a}) \equiv \frac{\partial \chi^2}{\partial a_k} = -2 \sum_{i=1}^N \frac{y_i - y_i(\mathbf{a})}{\sigma_i^2} \frac{\partial y_i(\mathbf{a})}{\partial a_k}$$

$\mathbf{a}$  = set of free parameters

- Gradient-descent method

- Steepest descent opposite to the gradient:  $\delta a_k = -\gamma_k \nabla_k \chi^2$
- Optimal step:  $\gamma_k = 1/H_{kk}$
- Efficient to approach the minimum, but often slow to converge

# The Levenberg-Marquardt method

$$\delta \mathbf{a} = -\mathbf{M}^{-1} \cdot \nabla \chi^2$$

Newton algorithm

$$M_{kl} = H_{kl}$$

Gradient-search method

$$M_{kl} = H_{kl} \times \delta_{k,l}$$

- **Levenberg-Marquardt method**

- Combine both methods:

$$\delta \mathbf{a} = -\mathbf{M}^{-1} \cdot \nabla \chi^2 \quad \text{with} \quad \begin{cases} M_{kk} &= H_{kk}(1 + \lambda) \\ M_{kl} &= H_{kl} \text{ if } k \neq l \end{cases}$$

- $\lambda \gg 1$ : Gradient-search method (approach)
  - $\lambda \ll 1$ : Newton method (convergence)
  - Initialize  $\lambda$  to a “large” value
  - Decrease  $\lambda$  as the function to be minimized decreases
  - Typically **only  $\sim 10$  iterations** needed to converge

# Model optimization

## Levenberg-Marquardt method, Miglio & Montalbán (2005)

- Choose a set of observational constraints  $x_{i,\text{obs}}$ 
  - non seismic:  $T_{\text{eff}}$ ,  $L$ ,  $[\text{Fe}/\text{H}]_{\text{surf}}$ ,  $\log g$ , radius, etc.
  - seismic: frequencies  $\nu_{n,l}$ , frequency separations or ratios
- Adjust unknown star properties
  - age, mass, initial helium abundance  $Y_0$ , initial metallicity  $[\text{Fe}/\text{H}]_0$
  - mixing-length parameter  $\alpha_{\text{conv}}$ , overshooting parameter  $\alpha_{\text{ov}}$ , etc.
- Other properties are outputs of the optimized model (R, internal properties)
- Need at least as many constraints as unknown parameters

$\chi^2$ -minimization, accounting for correlations

$$\chi^2 = \sum_{i=1}^{N_{\text{obs}}} (x_{i,\text{mod}} - x_{i,\text{obs}})^T \cdot C^{-1} \cdot (x_{i,\text{mod}} - x_{i,\text{obs}})$$

# Model optimization

Sets of seismic parameters

**Individual frequencies:**  $\nu_{n,I}$

- whole observed range
- truncated range: keep modes with orders close to  $\nu_{\max}$ ; drop high order modes (large freq. uncertainties)

**Frequency separations:**  $\langle \Delta\nu \rangle$ ,  $\Delta\nu_I$ ,  $\langle d_{02/01/10} \rangle$ ,  $d_{02/01/10}(n)$

**Both cases need to be corrected from surface effects**

Treatment of surface effects: several options

*Kjeldsen et al. 2008, Brandão et al. 2011*

*Ball & Gizon, 2014*

*Sonoi et al. 2015*

# Model optimization

## Sets of seismic parameters

**Frequency separation ratios:**  $r_{02/01/10}(n)$

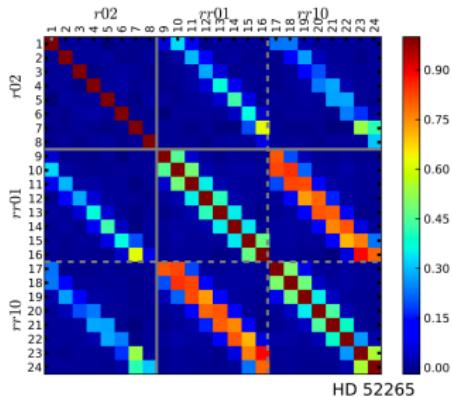
- définitions: *Roxburgh & Vorontsov (2003)*
- model ratios interpolated to observed freq. (Roxburgh & Vorontsov, 13)
- data are correlated: need to evaluate the covariance matrix

# Case of correlated observables

- Correlated observables (e.g. large, small separations, ratios...):

$$\chi^2 = [\mathbf{y} - \mathbf{y}(\mathbf{a})]^T \cdot \mathbf{C}^{-1} \cdot [\mathbf{y} - \mathbf{y}(\mathbf{a})]$$

- where  $\mathbf{C}$  is the covariance matrix between the observables
- can be calculated analytically for linear combinations of frequencies, or else numerically through Monte Carlo simulations



# Conditioning of covariance matrix

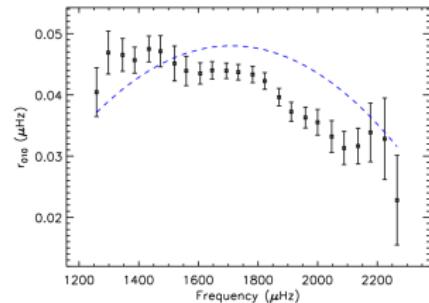


Conditioning of covariance matrix can be very poor!

- Example: fit of a 2<sup>nd</sup>-order polynomial to  $r_{010}$  ratios

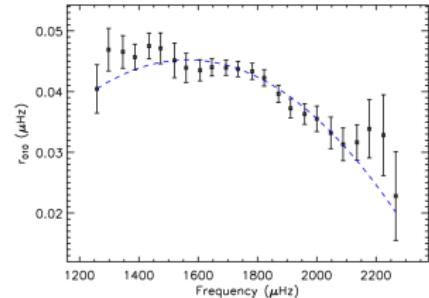
$$\det(\mathbf{C}) \sim 10^{-100}$$

$$\text{cond}(\mathbf{C}) \sim 10^6$$



- Solution: project onto the subspace corresponding to the  $N$  highest eigenvalues to improve the conditioning of  $\mathbf{C}$

$$\chi^2 = [\mathbf{P}(\nu) - \mathbf{r}_{010}]^T \cdot (\mathbf{M}^T \mathbf{W} \mathbf{M}) \cdot [\mathbf{P}(\nu) - \mathbf{r}_{010}]$$



# Model initialization

## Levenberg-Marquardt method

→ occurrence of secondary minima

Other problem: helium-mass degeneracy

Optimization in 3 steps

- Optimization (*O1*), no seismic constraints
  - 3 observational constraints:  $T_{\text{eff}}$ ,  $L$ , [Fe/H]<sub>surf</sub>
  - take  $Y_0$  from  $(\Delta Y / \Delta Z)_{\odot}$ ,  $\alpha_{\text{conv}} = \alpha_{\text{conv}, \odot}$
  - fit age  $A$ , mass  $M$ , [Fe/H]<sub>0</sub>
- Optimization (*O2*), seismic constraints
  - 5 constraints:  $T_{\text{eff}}$ ,  $L$ , [Fe/H]<sub>surf</sub> +  $\langle \Delta\nu \rangle$ ,  $\nu_{\text{max}}$
  - use scaling relations  $(M, R) \implies (\langle \Delta\nu \rangle, \nu_{\text{max}})$
  - fit  $A$ ,  $M$ , [Fe/H]<sub>0</sub>,  $Y_0$ ,  $\alpha_{\text{conv}}$
- Further optimizations: approach from *O1* or/and *O2* parameters

# Model initialization: HH2a case

Id	constraints	optimized	fixed
O1	$T_{\text{eff}}$ , $L$ , [Fe/H] <sub>surf</sub>	$A$ , $M$ , [Fe/H] <sub>0</sub>	$(\Delta Y / \Delta Z)_{\odot}$ , $\alpha_{\text{conv}, \odot}$
O2	$O1 + \langle \Delta \nu \rangle$ , $\nu_{\text{max}}$	$O1 + Y_0, \alpha_{\text{conv}}$	-
O3	$O1 + r_{02}(n) + r_{010}(n) + \nu_{n_{\min}, l}$	$O2$	-

Id	A (Myr)	$M/M_{\odot}$	$Y_0$	$\alpha_{\text{conv}}$	$R/R_{\odot}$	$\chi^2/N_{\text{obs}}$
Exact	3216	1.182	0.250	0.50	1.335	-
O1	$5436 \pm 1505$	$1.145 \pm 0.038$	0.266	0.60	1.386	$3 \cdot 10^{-6}$
O2	$4398 \pm 1343$	$1.201 \pm 0.020$	$0.242 \pm 0.022$	$0.60 \pm 0.10$	1.338	1.5
O3 <sub>2</sub>	$3685 \pm 195$	$1.153 \pm 0.008$	$0.266 \pm 0.006$	$0.49 \pm 0.02$	1.329	1.6
O3 <sub>12</sub>	$3299 \pm 131$	$1.182 \pm 0.008$	$0.260 \pm 0.005$	$0.50 \pm 0.01$	1.338	1.2
O3 <sub>1</sub>	$3502 \pm 152$	$1.216 \pm 0.009$	$0.247 \pm 0.005$	$0.53 \pm 0.02$	1.355	1.3

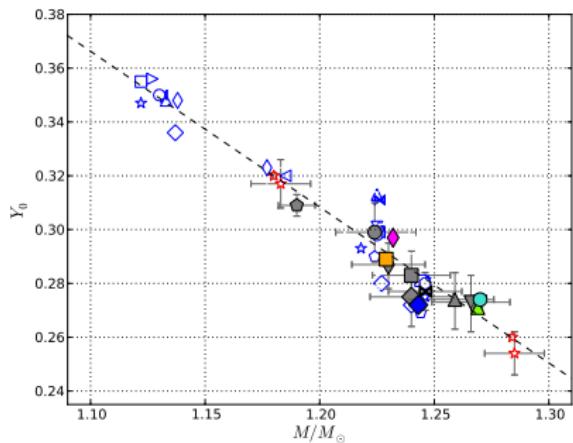
Models  $O3_{1,12,2}$

- different starting parameters
- similar fit to seismic constraints
- different ( $Y$ ,  $M$ ) combinations → degeneracy

# Model optimization: helium-mass degeneracy

A range of solutions for different ( $Y_0, M$ )

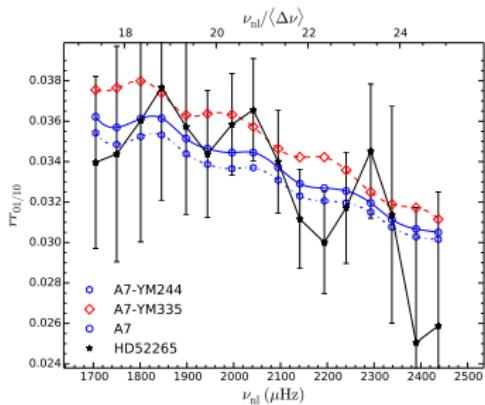
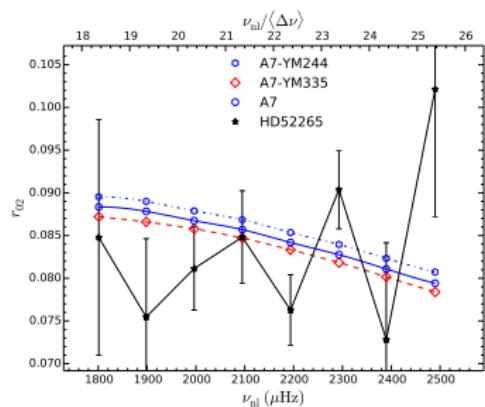
The better the global parameters, the narrower the range.



for the CoRoT target HD52265:

$$Y_0 \simeq -0.58 \times M/M_\odot + 1.00$$

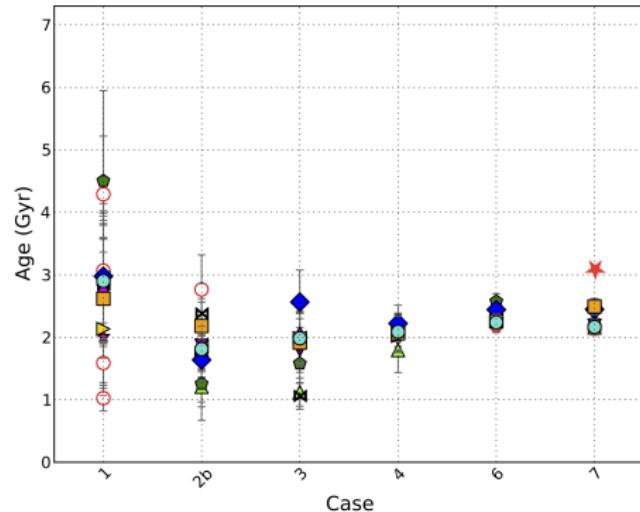
Lebreton & Goupil 14



# Sensitivity to model input physics & constraints

## Case of HD 52265

seismic constraints	adjusted parameters	fixed
no	$A, M, (Z/X)_0$	$Y, \alpha_{\text{conv}}$
$\langle \Delta v \rangle$	$A, M, (Z/X)_0, \alpha_{\text{conv}}$	$Y$
$\langle \Delta v \rangle, v_{\text{max}}$	$A, M, (Z/X)_0, Y, \alpha_{\text{conv}}$	-
$\langle \Delta v \rangle, \langle d_{02} \rangle$	$A, M, (Z/X)_0, Y, \alpha_{\text{conv}}$	-
$r_{02}(n)$ $rr_{01/10}(n)$	$A, M, (Z/X)_0, Y, \alpha_{\text{conv}}$	-
$v_{n,l}$	$A, M, (Z/X)_0, Y, \alpha_{\text{conv}}$	-



Fit with individual frequencies :

surface effect corrections:  $A = 2.17 \pm 0.02$  Gyr ;  $M = 1.27 \pm 0.02 M_{\odot}$

no surface effects:  $A = 3.15 \pm 0.03$  Gyr ;  $M = 1.30 \pm 0.02 M_{\odot}$

Lebreton & Goupil 14

# Estimation of errors

- Estimating errors for the fitted parameters:
  - Errors given by the **diagonal coefficients of the inverse of the Hessian matrix**
  - Hessian matrix obtained numerically by setting  $\lambda = 0$  after convergence
  - Results may depend on the choice of steps that used to estimate the derivatives  $\partial y_i(\mathbf{a})/\partial a_k$  for the Hessian
- Estimating errors for other parameters (e.g. radius):
  - No direct estimate provided through the fit
  - For a parameter  $\pi$  which is not a free parameter of the fit,  $\pi(a_j)$

$$\sigma_\pi = \sqrt{\sum_{i,j=1}^N C_{ij} \left( \frac{\partial \pi}{\partial a_i} \right) \left( \frac{\partial \pi}{\partial a_j} \right)}$$

Covariance matrix =  
inverse of Hessian

# Estimation of errors

- Testing the errors produced by LM algorithm
  - Input model close to HH2a
  - Use of classical ( $T_{\text{eff}}$ ,  $Z/X_{\text{surf}}$ ,  $L/L_{\odot}$ ) and seismic constraints
  - Add random normally-distributed noise to all observables according to the statistical errors of their measurements
  - Series of ~ 50 optimizations performed using LM algorithm (low precision)

Using individual frequencies ( $\ell = 0, 1, 2$ )

	True value	Average of fitted values	RMS of fitted values	LM errors
Mass ( $M_{\odot}$ )	1.22	1.223	0.038	0.029
Age (Myr)	3000	3007	149	75
$Y_0$	0.28	0.278	0.019	0.012
$(Z/X)_0$	0.0215	0.0214	0.0020	0.0012
$\alpha_{\text{conv}}$	0.60	0.612	0.023	0.022
$R (R_{\odot})$	1.487	1.482	0.042	0.028

# Estimation of errors

- Testing the errors produced by LM algorithm
  - Input model close to HH2a
  - Use of **classical** ( $T_{\text{eff}}$ ,  $Z/X_{\text{surf}}$ ,  $L/L_{\odot}$ ) and **seismic constraints**
  - **Add random normally-distributed noise** to all observables according to the statistical errors of their measurements
  - **Series of ~ 50 optimizations** performed using LM algorithm (low precision)

Using combinations of frequencies ( $\delta v_{01}$ ,  $\delta v_{10}$ ,  $\delta v_{02}$ ,  $v_0$ )

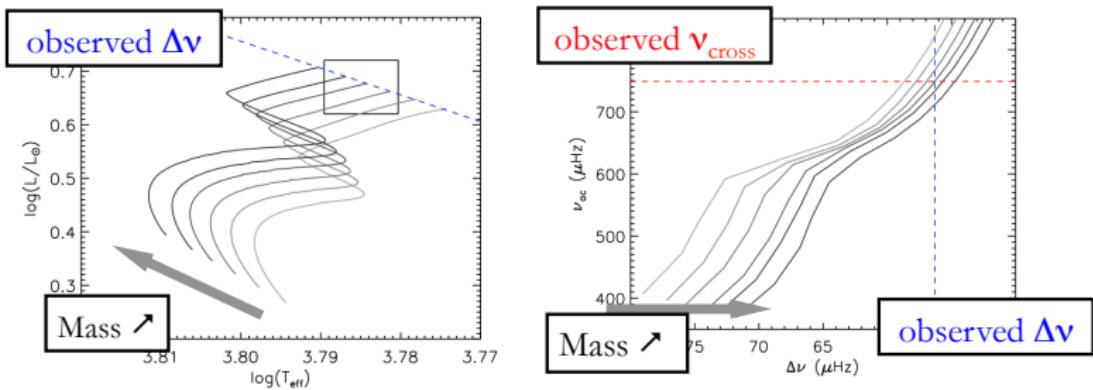
	True value	Average of fitted values	RMS of fitted values	LM errors
Mass ( $M_{\odot}$ )	1.22	1.227	0.026	0.019
Age (Myr)	3000	3068	194	44
$Y_0$	0.28	0.275	0.014	0.010
$(Z/X)_0$	0.0215	0.0220	0.002	0.001
$\alpha_{\text{conv}}$	0.60	0.592	0.04	0.02
$R (R_{\odot})$	1.487	1.491	0.011	0.009

# Examples of the use of LM algorithm

- **Main sequence stars:**
  - Detailed seismic modeling of specific targets (e.g. HD49933 Goupil et al. 2011, HD52265 Lebreton et al. 2014)
  - Test case of HD52265 to estimate the influence of seismic constraints in modeling (Lebreton et al. 2014)
  - Measuring the size of convective cores (Deheuvels et al. 2010, 2016)
  - ...
- **Subgiants** with mixed modes:
  - Classical Levenberg-Marquardt procedure fails because of the fast evolution of g-mode frequencies (core contraction)
  - Bijection between (Mass, age) and  $(\Delta v, v_{\text{cross}})$  for a given set of other parameters

# Examples of the use of LM algorithm

- Subgiants with mixed modes:



Free parameters of the fit:  $M, \tau, Y_0, \alpha_{\text{ov}}, (Z/X) \dots$

Successfully applied to CoRoT subgiant HD49385 (Deheuvels et al. 2011)

fixed to  $(\tilde{M}, \tilde{\tau})$  by fitting  $(\Delta\nu, \nu_g)$

# Levenberg-Marquardt optimization

## Advantages

- No restrictions on the number of unknown parameters to be estimated provided there are enough observational constraints
  - no necessary a priori assumption on  $Y$ ,  $(\Delta Y / \Delta Z)$ ,  $\alpha_{\text{conv/ov}}$
  - easy to modify/add physical processes and refit parameters
- Errors on the fitted parameters easily obtained via the Hessian matrix
- the procedure is adaptable to use other stellar evolution codes (MESA), oscillation codes (ADIPLS, GYRE, LOSC)

## Caveats

- Secondary minima
- Error estimates for parameters that are not fitted (ex. radius, convection zone boundaries, etc.) have to be evaluated a posteriori.
- The order of magnitude of the error bars is correct but rather imprecise (work to be done)