Stellar and Galactic Archaeology with Bayesian Methods

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Spectroscopy

temperature, surface gravity, chemical abundances: Li, Be, Be,CNO, a-group, Fe-peak, s-r process, U,Th

abundance trends

metallicity gradients

- + rotation velocity
- + activity
- + radial velocity
- + mass, age
- + distances





Galactic archeology and the first stars



Nucleosynthesis and the origin of chemical elements



Chemical evolution of galaxies



chemo-dynamic correlations

age-, mass-metallicity relations



Stellar Spectroscopy Junior Research Group, Max-Planck Institute for Astronomy, Heidelberg Major progress in observations in the past 10 years: VLT's and, soon, ELT's

Ongoing large-scale stellar spectroscopic surveys: SDSS (Apogee, SEGUE), RAVE, Gaia-ESO, GALAH

Future: 4MOST (20 million spectra, optical), MOONS (IR)









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Maria Be





Basic model atmosphere theory: non-LTE (NLTE), 3D hydrodynamics, magnetic fields, winds, sphericity, molecular opacities, binarity, chromospheres, etc

What are the physical conditions of the most likely model?



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What are the physical conditions of the most likely model?

- all observed stars are point sources
- observed spectra are not perfect -> noise + data reduction problems
- stellar models are not perfect
- stellar spectra are in reality not so different → parameter degeneracies and correlations

The physical challenge

1. What is a <u>good</u> stellar model?

2. What type of physics can we afford computationally?

The statistical challenge

1. What is the <u>best-fit</u>stellar model?

2. Do we have prior knowledge from previous or complementary experiments?

Even the best observed spectra are worth nothing without good model comparison methods

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Max-Likelihood Spectroscopy

'observed' spectrum \rightarrow the goal is to estimate T_{eff} , log(g), and metallicity of a star

What if we rely only on the classical approach: maximum-likelihood L?

$$L \sim exp(-\chi^2/2)$$

$$\chi^2 = \sum_j \left(\frac{D_j^{\text{obs}} - D_j(\text{Teff, logg, } Z)^{\text{theor}}}{\sigma_j}\right)^2$$

where D_j , j = 1... n are observables, i.e., spectrum in a given frequency bin

However, L attains its global maximum only if for each model characterized by [T_{eff}, log(g), [Fe/H]] there is a unique model spectrum



movies under www.mpia.de/~bergemann/outreach

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Max-Likelihood stellar parameters



'Orthodox' (standard) methods are suitable:

- selection effects lead to major biases
- *imperfect* data often disregarded
- parameter degeneracies caused by physical limitations of the models
- correlated errors, ...
- often there is just not enough information in the observed spectrum



Ad-hoc 'correction' of stellar parameters using stellar models

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R ~ 200 000, the spectrum of the Sun



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R ~ 2000, the SDSS spectrum of a solar-like 'twin'



At low R and S/N most of spectral information is washed out

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High-resolution observations

The Milky Way disk



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Small errors (high-quality data) \rightarrow well-defined PDF

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low-quality data \rightarrow blurred or multi-component PDF

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Bayesian model testing

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Complementary experiments

log(g)

large stellar surveys observe millions of stars

spectroscopy: Sloan Digital Sky Survey, Gaia-ESO, Apogee... → stellar spectra: Luminosity, Temperature

photometry: VISTA, 2MASS, PS1, Skymapper ... → magnitudes in different filters:

asteroseismology, stellar evolution: CoRoT, Kepler → mass, age of a star

astrometry: Hipparcos, Gaia (launched 2013) → distances

Bayesian model testing Stellar Spectroscopy

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Bayesian spectroscopy

1) In our context, the 'core' parameter space is defined by: metallicity (expressed by iron abundance), effective temperature, and surface gravity

T_{eff}, log g, [Fe/H]

2) Their plausibility is estimated based on the information contained in:

- observed stellar spectra
- model stellar spectra
- stellar evolution models (not all luminosities, masses, and ages are possible)
- parallaxes, photometry (constraints on log g (π) and L (color))
- WMAP \rightarrow constraints on the max age

a set of parameters $X = X_1, \ldots, X_n$ a set of observations $O = O_1, \ldots, O_m$

The goal is to construct a full posterior PDF in all parameters

$$P(\boldsymbol{X}|\boldsymbol{O}) = \frac{P(\boldsymbol{X})}{P(\boldsymbol{O})}P(\boldsymbol{O}|\boldsymbol{X}),$$

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L(max) attains its global maximum for **all** three isochrones!

In standard *L(max)* approach, ages suffer from a 'terminal age bias', i.e., short-lived evolution stages get **un-physically** high probability.

Jørgensen & Lindegren 2005

run over the full multi-D grid of spectroscopic models

use adaptive, iteratively refined mesh guided by photometry + prior

Final values of Teff, log(g), [Fe/H]:

$$P(X_j|\boldsymbol{O}) = \int \int P(X_1, \ldots, X_n|\boldsymbol{O}) \, \mathrm{d}x_1 \ldots \, \mathrm{d}x_{j-1} \, \mathrm{d}x_{j+1} \ldots \, \mathrm{d}x_n.$$

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Results

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Schoenrich & Bergemann 2014, MNRAS, 443 Maria Bergemann

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Bayesian Teff

Bayesian Teff

Bayesian gravity

Bayesian gravity

Schoenrich & Bergemann 2014

Stars from Sloan Digital Sky Survey

Bayesian

Spectroscopic only

Schoenrich & Bergemann 2014, MNRAS, 443

Bayesian: summary

✓ Pros

- All parameters (stellar parameters, distances, ages) within one single, consistent analysis
- Automatic detection of pathologic (or interesting...) cases
- Ability to quantify systematic shifts/errors

✓ Cons

- The analysis scheme is too rigid: we cannot handle objects with physical properties that are not within the pre-computed model grids
- expanding the basic parameter set is expensive $3D \rightarrow 4D$ (?) .. We need 30
- Inclusion of priors all stellar populations are different (the rate of star formation, IMF?)

Even the best observed spectra and good statistics are worth nothing without good models

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models of stellar atmospheres

Basic model atmosphere theory – the models are usually trained on a given class of stars

We do not have a single consistent set of models which describe all types of stars found in nature:

- rotate up to 100 km/s
- pulsate
- lose mass in winds
- magnetic fields (kG)
- exist as binaries or multiple systems (overlapping spectra)
- mass motions (inflows, outflows)
- Circumstellar dust shells

Classical stellar atmosphere models

- Iocal thermodynamic equilibrium
- Hydrostatic equilibrium
- 1-dimensional
- ✓ plane-parallel → semi-infinite
- plus about 30 ad-hoc free parameters

Classical stellar atmosphere models

Hydrostatic equilibrium

Citations/Publication Year

- 1-dimensional
- \checkmark plane-parallel \rightarrow semi-infinite
- plus about 30 ad-hoc free parameters

Publication Year for 1998ApJ...499..914S

the first paper on 3D hydro-dynamical model atmospheres of stars Stein & Nordlund, ApJ, 499, 1998

$$\nabla P_{rad} = -1/c \int_{0}^{2012} (\kappa_{v} + \sigma_{v}) F_{v} dv$$

$$\overset{\text{vblication Year}}{=} \begin{bmatrix} \sim 3400 \text{ citations} \\ \text{the most widely-used} \\ 1D \text{ p-p grid of stellar spectra} \\ \text{in astronomy} \\ \text{Kurucz, ApJS, 40, 1979} \end{bmatrix}$$

Kurucz+

Publication Year

- Iocal thermodynamic equilibrium
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Does that work?

The observed image of the Sun Swedish Solar Telescope (1m)

© van der Voort, University of Oslo

State-of-the-Art

- 3D hydrodynamics
- non-LTE (consistent treatment of the radiation field and physical state of the gas —> gas must respond to the radition loss from the surface)
- ab initio
- complete sampling of opacity sampling (up to 100 million spectral lines)
- no for unphysical calibrations
 ('mixing length',
 'microturbulence')

3D Hydrodynamical models

the same scales - both images 20x20 Mm (!)

(c) Asplund

3D NLTE spectroscopy

average spatially (x,y)

Bergemann et al. in prep.

Bergemann et al. in prep.

State-of-the-Art <3D> NLTE: no free parameters

Summary

Stellar model atmospheres

need consistent improvements on models: 1D-3D, (N)LTE, rotation, stellar evolution

Bayesian-type (full – Prob.) schemes

- All parameters within one single, consistent analysis
- Automatic detection of pathologic (or interesting...) cases
- direct ability to quantify systematic shifts/errors, in future: reddening, distances, binary fractions, He

Good algorithms efficiently combining models and Bayesian are needed:

4MOST survey (2021) – 20 million stellar spectra with distances & kinematics (Gaia follow-up)

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