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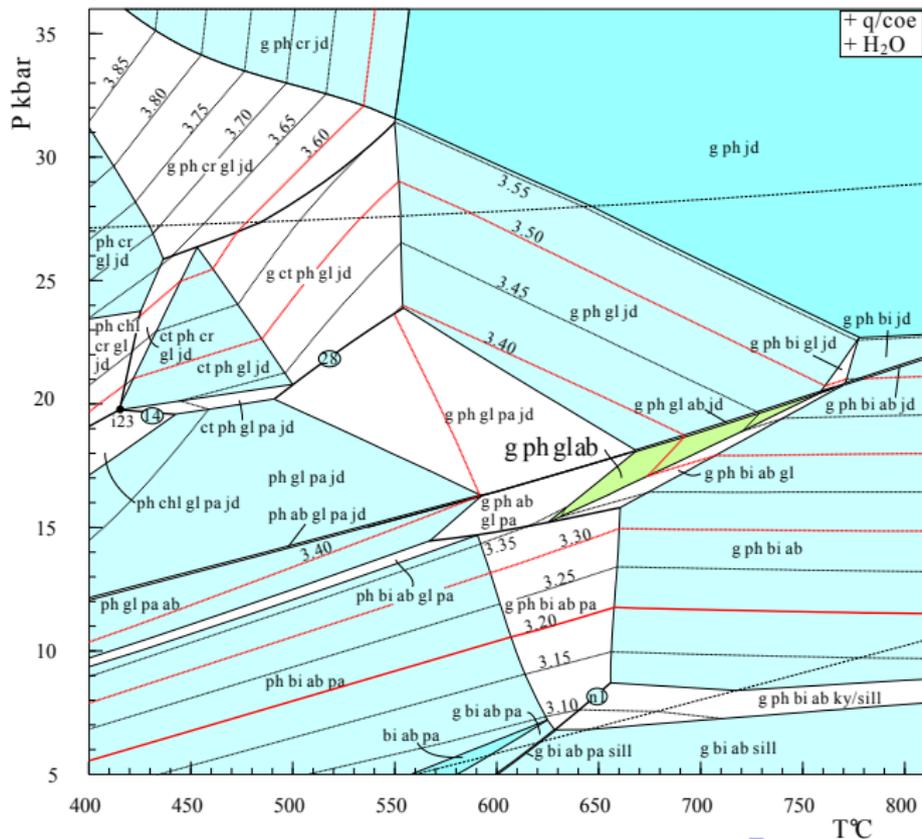
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 - ▶ “optimal thermobarometry”: average PT , or \overline{PT}
- ▶ calculated pseudosections
 - ▶ a powerful sort of thermobarometry, via
 - ▶ mineral stability fields
 - ▶ mineral proportion isopleths
 - ▶ mineral composition isopleths

pseudosection approach

- ▶ we have had a go at the pseudosection sort of thermobarometry, at least indirectly via learning how to calculate them

a garnet-glaucophane schist from the Tianshan



logic

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 - ▶ acknowledging that single reaction methods should really be considered to be a subset of these
- ▶ in particular I want to look at
 - ▶ sources of uncertainty (in the context of the under-reporting of uncertainties in single-reaction thermobarometry),
 - ▶ and making all the methods thermodynamically consistent with each other

background

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- ▶ conventional thermobarometry has tended to involve one or two equilibria, “directly-calibrated” from experimental and other data (e.g. g-bi Fe-Mg exchange thermometry combined with GASP)

- ▶ thermobarometry in an internally-consistent thermodynamic dataset setting (i.e. with those data having been through a thermodynamic “filter”) for example \overline{PT}

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 - ▶ merit of consistency with other mineral equilibria methods
 - ▶ possibility of realistic assignment of uncertainties, and so
 - ▶ possibility of recognising when there is little or no thermobarometric in a mineral assemblage

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- ▶ address the apparent disconnect between reported uncertainties on PT in this, and those implied by \overline{PT}
- ▶ I'll do this by looking at sources of uncertainty in general,
- ▶ then look at the g-cpx Fe-Mg exchange thermometer as an example.

what is involved in thermobarometry? (1)

- ▶ *essential* idea in thermobarometry:
 - ▶ the **extrapolation** of experimental data on mineral properties and mineral equilibria in P , T and composition

what is involved in thermobarometry? (1)

- ▶ *essential* idea in thermobarometry:
 - ▶ the **extrapolation** of experimental data on mineral properties and mineral equilibria in P , T and composition
- ▶ theme:
 - ▶ use equilibrium thermodynamics, as well as statistics, in order to do this. And common sense!

what is involved in thermobarometry? (2)

- ▶ formulation
- ▶ calibration
- ▶ application

what is involved in thermobarometry? (2)

- ▶ formulation
 - ▶ thermodynamic modelling, but no one good model, so
 - ▶ technical judgement necessary, as well as heuristics
- ▶ calibration
 - ▶ good data selection
 - ▶ but more parameters than data to constrain them,
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- ▶ reporting results
 - ▶ realistic assessment/assignment of uncertainties

uncertainty in thermobarometry (1)

an aim is to calculate appropriate uncertainties on calculated PT values. How do the uncertainties arise? First:

- ▶ bias

bias is (one of) our most serious problems, not least because we cannot always tell when we are dealing with a bias problem. . .

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uncertainty in thermobarometry (4)

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garnet-clinopyroxene geothermometry

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- ▶ calibration primarily from high PT experimental work
- ▶ uncertainties commonly under-reported ($\pm 30^\circ\text{C}$!)

The garnet–clinopyroxene Fe^{2+} –Mg geothermometer: an updated calibration

E. KROGH RAVNA

Department of Geology, University of Tromsø, N-9037 Tromsø, Norway (erlingr@ibg.uit.no)

ABSTRACT Multiple regression analysis on an extended dataset has been performed to refine the relationship between temperature, pressure, composition and the Fe–Mg distribution between garnet and clinopyroxene. In addition to a significant dependence between the distribution coefficient K_D and $X_{\text{Ca}}^{\text{Grt}}$ and $X_{\text{Mg}\#}^{\text{Grt}}$, as shown by the experimental data, the effect of $X_{\text{Mn}}^{\text{Grt}}$ has also been incorporated using data from natural Mn-rich garnet–clinopyroxene pairs. Multiple regression of data ($n=360$) covering a large span in pressure, temperature and composition from 27 experimental datasets, combined with 49 natural high-Mn granulites from Ruby Range, Montana, USA, and Karnataka, India, yields the P – T –compositional relationship ($r^2=0.98$):

$$T(^{\circ}\text{C}) = [(1939.9 + 3270 X_{\text{Ca}}^{\text{Grt}} - 1396 (X_{\text{Ca}}^{\text{Grt}})^2 + 3319 X_{\text{Mn}}^{\text{Grt}} - 3535 (X_{\text{Mn}}^{\text{Grt}})^2 + 1105 X_{\text{Mg}\#}^{\text{Grt}} - 3561 (X_{\text{Mg}\#}^{\text{Grt}})^2 + 2324 (X_{\text{Mg}\#}^{\text{Grt}})^3 + 169.4 P(\text{GPa}) / (\ln K_D + 1.223)] - 273$$

where $K_D = (\text{Fe}^{2+}/\text{Mg})^{\text{Grt}} / (\text{Fe}^{2+}/\text{Mg})^{\text{Cpx}}$, $X_{\text{Ca}}^{\text{Grt}} = \text{Ca} / (\text{Ca} + \text{Mn} + \text{Fe}^{2+} + \text{Mg})$ in garnet, $X_{\text{Mn}}^{\text{Grt}} = \text{Mn} / (\text{Ca} + \text{Mn} + \text{Fe}^{2+} + \text{Mg})$ in garnet, and $X_{\text{Mg}\#}^{\text{Grt}} = \text{Mg} / (\text{Mg} + \text{Fe}^{2+})$ in garnet. The Fe^{2+} –Mg equilibrium

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- ▶ assumption of no ferric iron in the experiments used to calibrate the thermometer (maybe grossly unfair!?)
- ▶ *but* it is the best g-cpx thermometer out there currently...

g-cpx example

let's look at an example: Proyer *et al.* (2004, *Contributions to Mineralogy and Petrology*, **147**, 305–318) for a Dabie Shan coesite-bearing eclogite (SM93).

Si	Al	Fe ³⁺	Fe ²⁺	Mg	Ca	Na
2.01554	0.48883	0	0.11303	0.41074	0.44315	0.52342

with minor Ti, Cr, Mn and K, not shown.

This example is also used in an accompanying prac. Why? Because they have done Mössbauer on their minerals so they *know* how much ferric iron they have.

In a lot of conventional thermobarometry ferric iron is a major issue (because it is a critical unknown).

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 - ▶ mineral analysis uncertainty (1% relative on wt% oxides)
 - ▶ what to do about Fe^{3+} ??
charge balance calculation... (gives $\text{Fe}^{3+} = 0$);
or ignore it!! (is the same in this case)

mineral recalculation

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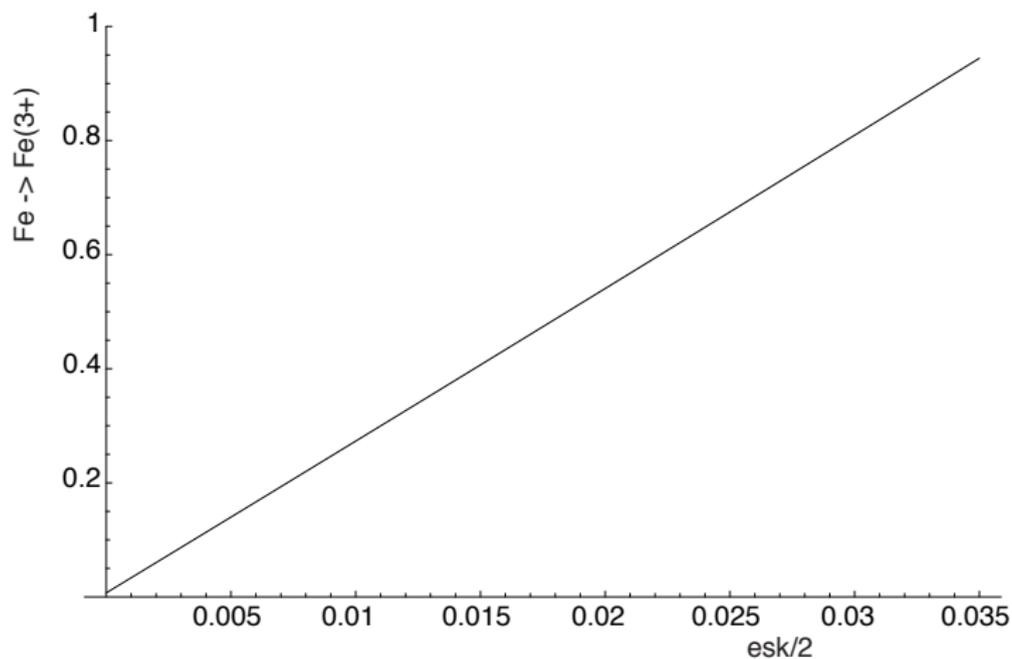
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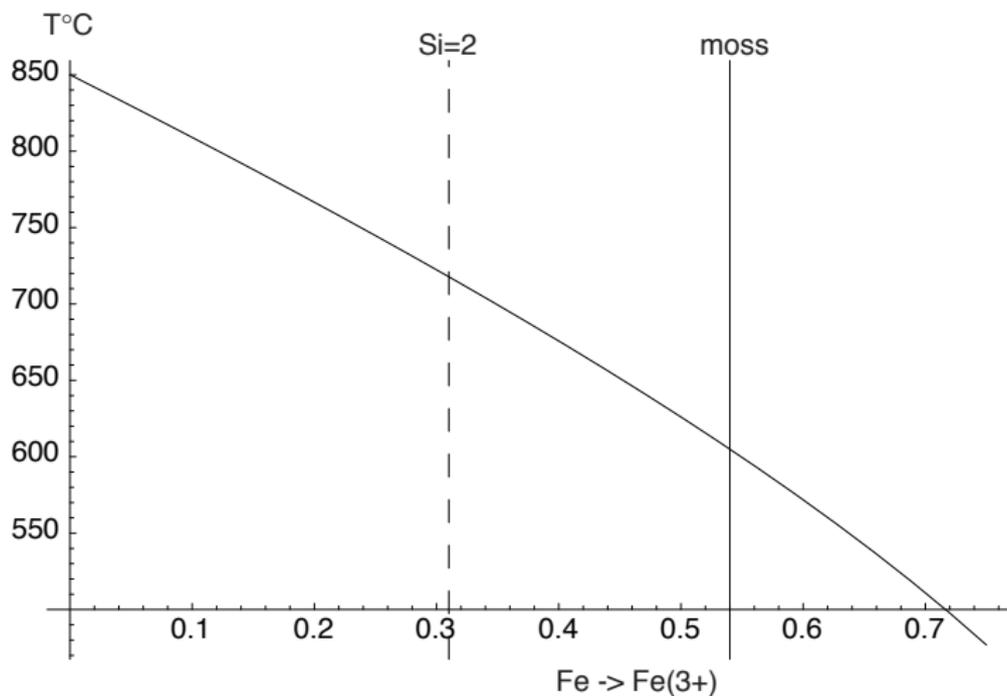
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- ▶ should use error propagation of mineral analysis uncertainty, using a charge balance calculation to get Fe^{3+} ,
- ▶ but what charge balance calculation? What if there is significant eskolaite in the cpx? (And what is significant?)
- ▶ alternatively, do a forward calculation, *specifying* Fe^{3+} (and eskolaite) and calculating the “best” analysis (by least squares) corresponding to the specification.

effect of eskolaite



dependence of T on $\text{FeO} \rightarrow \text{Fe}_2\text{O}_3$ conversion (1)



summarising

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- ▶ Proyer *et al.* (2004), *determined* Fe^{3+} , so in that case uncertainties are reduced (do the prac to see the details)

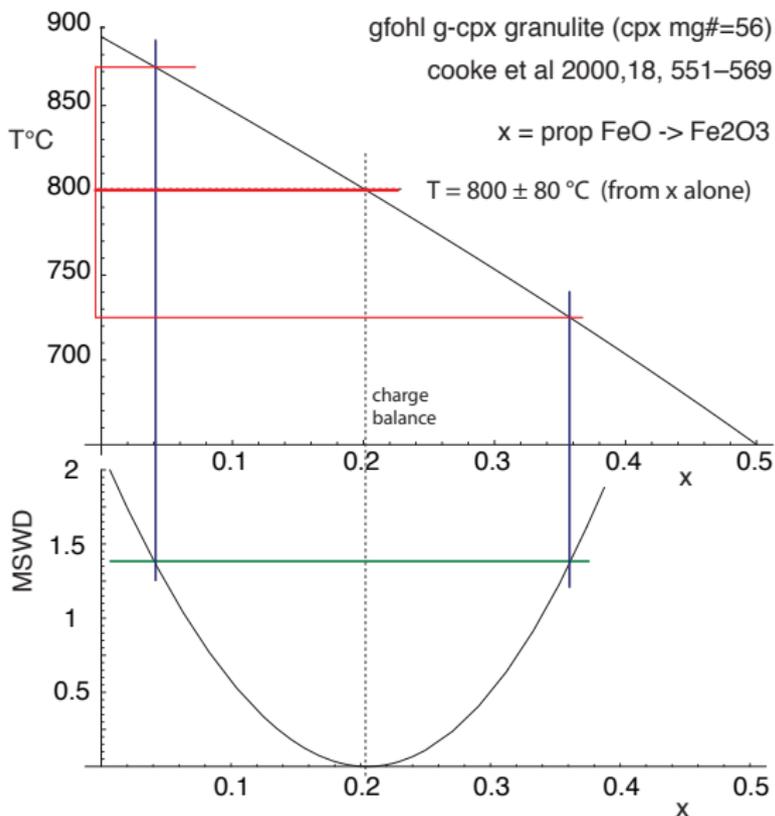
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- ▶ real implications for “maximum astonishment” publications: unfortunately doing the wrong thing with g-cpx thermometry is used to support what I think is a misidentification of peak metamorphic mineral assemblage...

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- ▶ the uncertainty on a temperature is therefore $2\sqrt{25^2 + 30^2}$, giving $T = 800 \pm 80^\circ\text{C}$
- ▶ and this is certainly a minimum...

now to move on

- ▶ generalising, uncertainties in conventional thermobarometry *are* commonly under-reported
- ▶ let's now look at \overline{PT}

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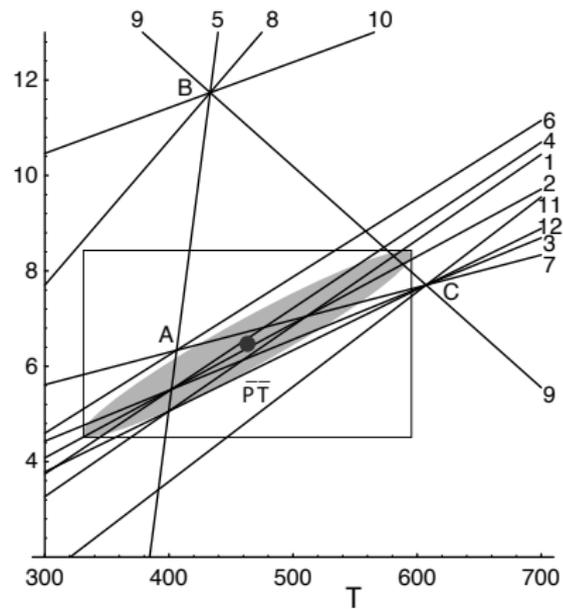
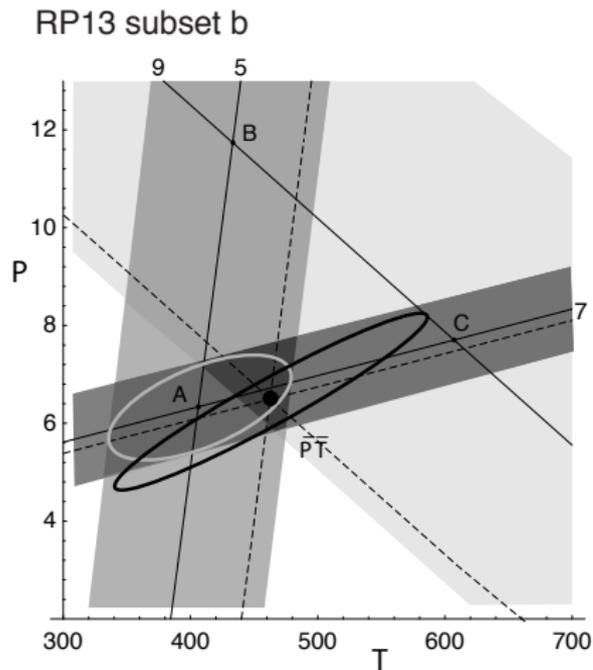
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conventional and \overline{PT} thermobarometry



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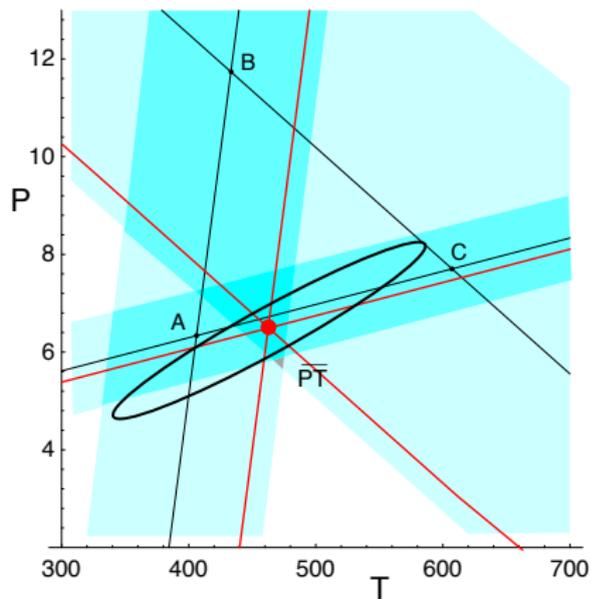
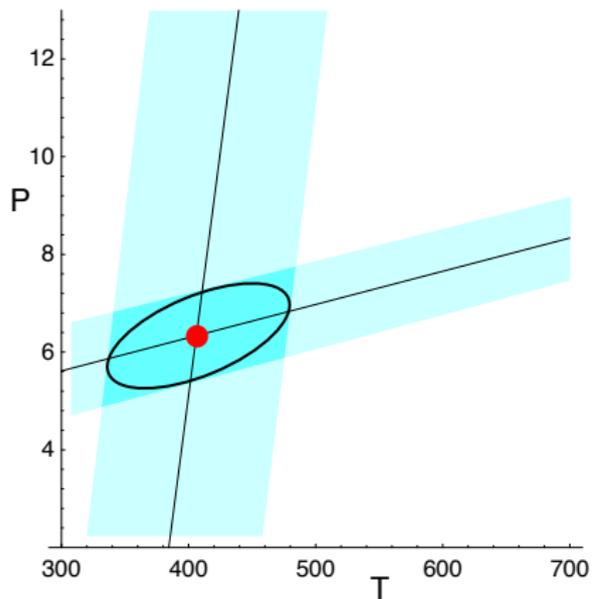
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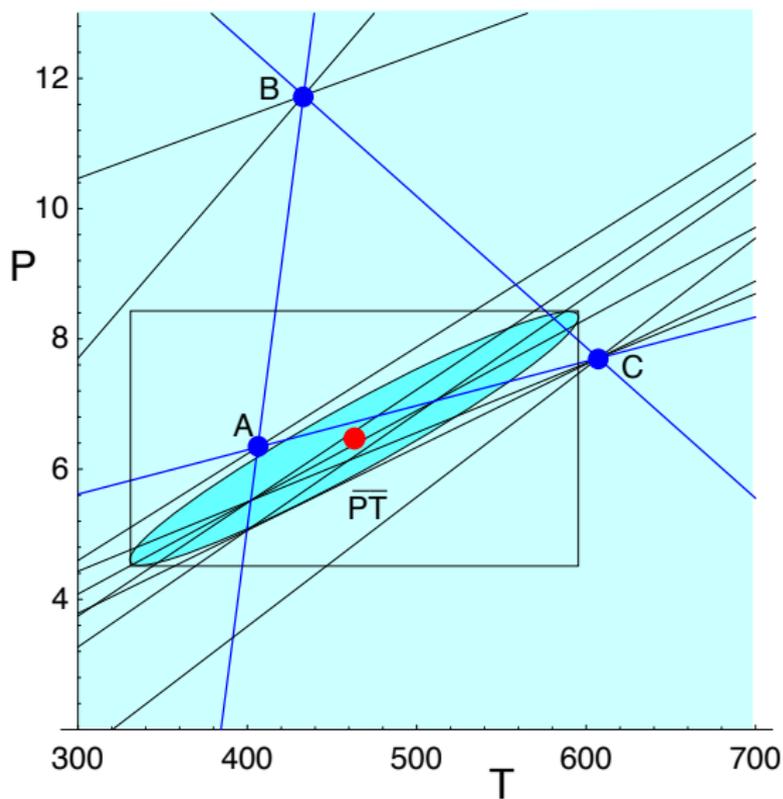
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\overline{PT} thermobarometry (1)



\overline{PT} thermobarometry (2)



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 - ▶ run!

an \overline{PT} example

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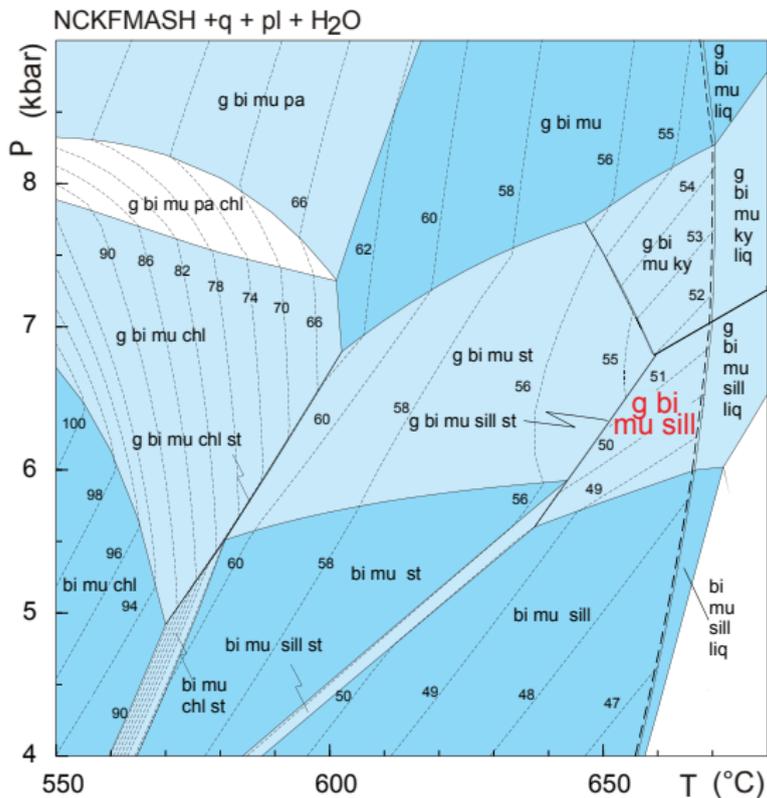
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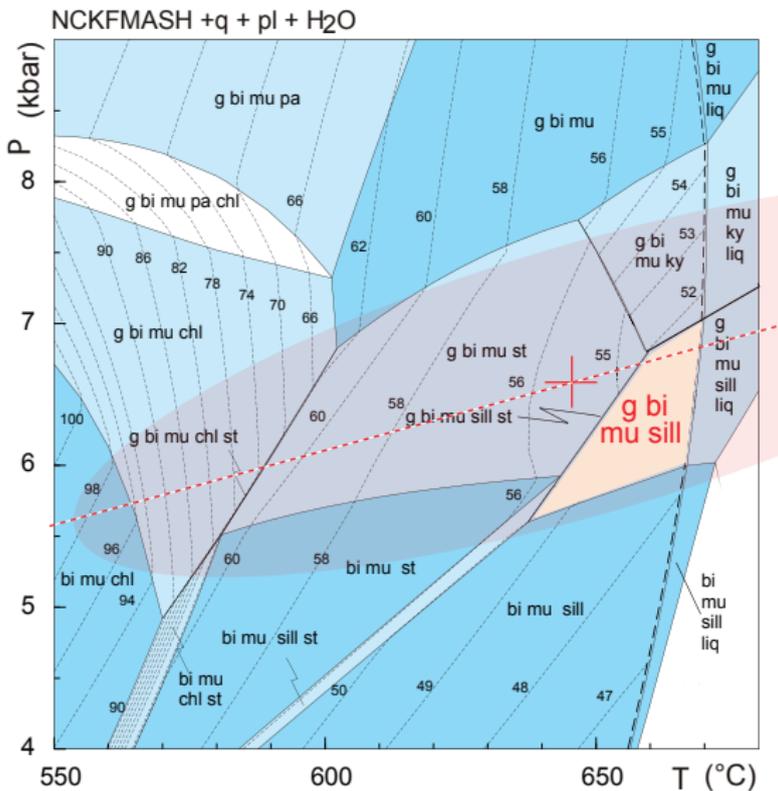
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the setting for a \overline{PT} example



\overline{PT} result



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- ▶ but pseudosections are likely to provide the most powerful tools for thermobarometry (as I will argue at the Brasilia meeting coming up)