thermobarometry

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- some overlap with my brasilia keynote...
- ...but hopefully covered at a slower pace!

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- multiple reaction (based on an internally-consistent dataset)
- calculated pseudosections

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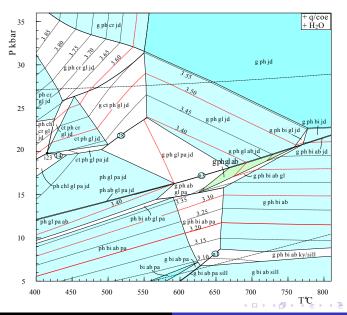
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- calculated pseudosections
 - a powerful sort of thermobarometry, via
 - mineral stability fields
 - mineral proportion isopleths
 - mineral composition isopleths

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 Tienshan



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- we have covered pseudosections at some length,
- ▶ so what I want to do is look at single- and multiple-reaction methods, particularly average PT, or \overline{PT}

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- we have covered pseudosections at some length,
- So what I want to do is look at single- and multiple-reaction methods, particularly average PT, or PT
 - 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, including pseudosections, thermodynamically consistent with each other

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 - 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 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.

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- ► theme:
 - use equilibrium thermodynamics, as well as statistics, in order to do this. And common sense!

formulation

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 - realistic assessment/assignment of uncertainties



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summarising, for PT calculations

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- sources of uncertainty arise from all of
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- ▶ uncertainties commonly under-reported (±30°C!)

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- ▶ not utilising all information available (ΔV and ΔS of the Fe-Mg exchange reaction)
- composition dependence not in an activity coefficient form
- assumption of no ferric iron in the experiments used to calibrate the thermometer (maybe grossly unfair!?)

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).

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 $Fe^{3+}=0$); or ignore it!! (is the same in this case)

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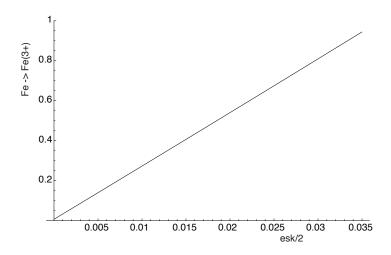
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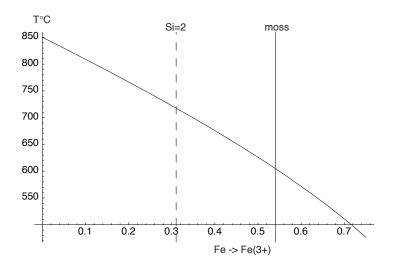
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- but what charge balance calculation? What if there is significant Ca-eskola molecule in the cpx? (And what is significant?)
- ▶ alternatively, do a forward calculation, *specifying* Fe³⁺ (and Ca-eskola molecule) and calculating the "best" analysis (by least squares) corresponding to the specification.



effect of eskolaite





summarising

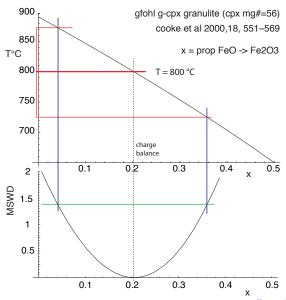
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- ▶ Proyer *et al.* (2004), *determined* Fe³⁺, so in that case uncertainties are reduced (do the prac to see the details)
- regularising calculations by using $Fe^{3+} = 0$ is indefensible (because of the strong upwards bias on T)



so let's estimate a T uncertainty for this cooke et al example

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- ▶ the uncertainty on a temperature is therefore $2\sqrt{25^2 + 30^2}$, giving $T = 800 \pm 80^{\circ}\mathrm{C}$
- 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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 - possibility of realistic assignment of uncertainties (via the goodness of fit parameter involved)

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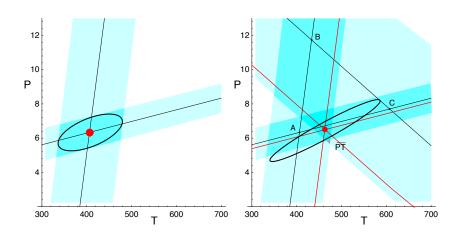
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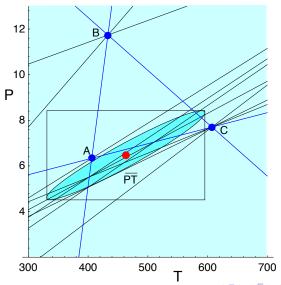
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- each reaction then defines a PT band, the bands being correlated with each other via the activities
- ▶ then \overline{PT} is the PT point that is as "close" as possible (in a least squares sense) to every reaction line



\overline{PT} thermobarometry (1)



\overline{PT} thermobarometry (2)



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► old way

new way

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 - run...
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- new way
 - take mode 1 datafile a-x coding
 - ightharpoonup convert recalculated analyses into $\{x, y, z \dots\}$
 - run!

using mode 1 coding

- ▶ show example of the new way:
- Proyer's Dabie Shan rock

another \overline{PT} example

▶ let's look at an example with g-bi Fe-Mg exchange as well as GASP, so featuring two major thermobarometers used out there. We shall see that there are also other reactions to consider, not just these two.

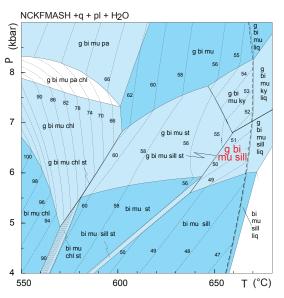
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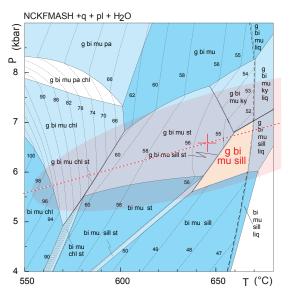
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- we can see the shape of the phase relations using a PT pseudosection

the setting for a \overline{PT} example



\overline{PT} result



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 - and so locating these relatively low variance mineral assemblages, amongst the many mineral assemblages that you might have from an area, is important thermobarometrically
- but pseudosections are likely to provide the most powerful tools for thermobarometry (as I will argue at the Brasilia meeting coming up)