Reconstruction of Neutral Final-State Particles in Neutrino-Argon Interactions

Margot MacMahon



J. Kopp, P. Machado, <u>M. MacMahon</u> and I. Martinez-Soler

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The next generation of neutrino experiments will be able to:

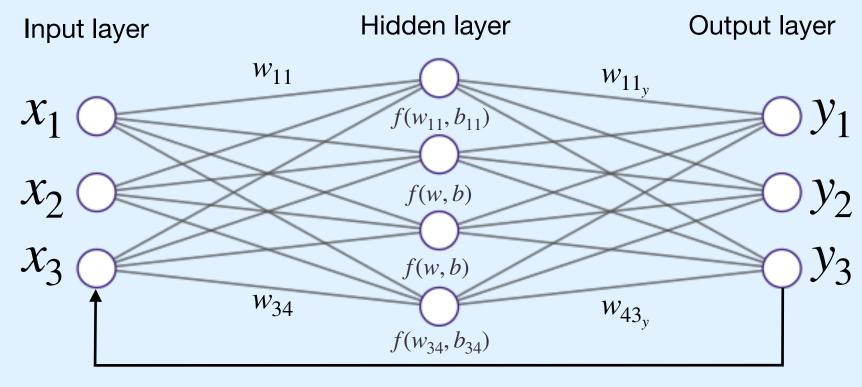
- \bullet Resolve neutrino oscillation parameters δ_{CP} , θ_{23} and the neutrino mass ordering
- Detect ν_e flux from near supernovae explosions
- Search for proton decay



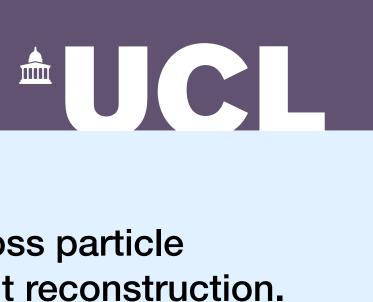
B. Abi et al, DUNE collaboration, 2002.03005 (2020)

Neural networks are widely used across particle physics experiments to improve event reconstruction.

Easily applied to regression problems, such as predicting the energy of an incoming particle.



Error backpropagation



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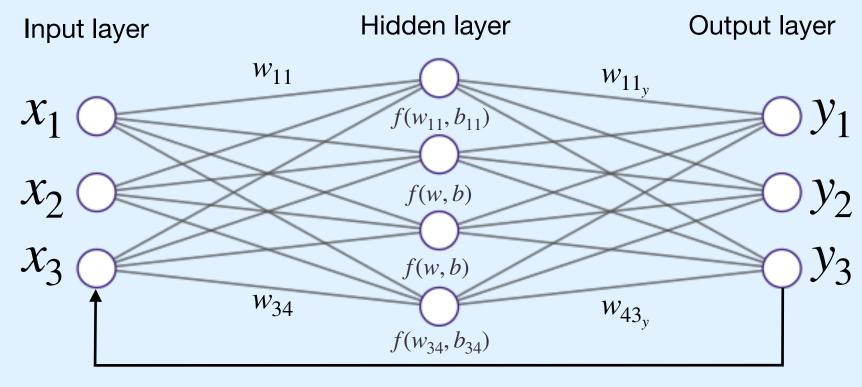


Invisible particles in final state of neutrino nucleon interaction still pose a challenge to accurate energy reconstruction.

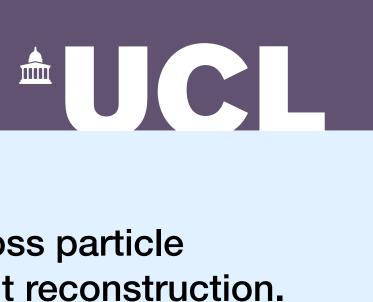
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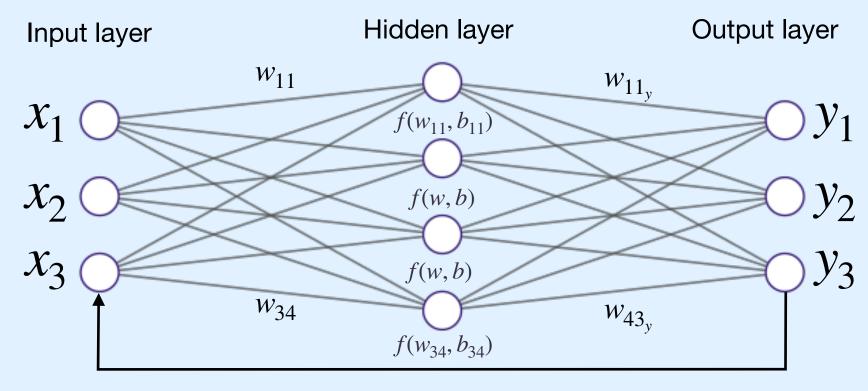


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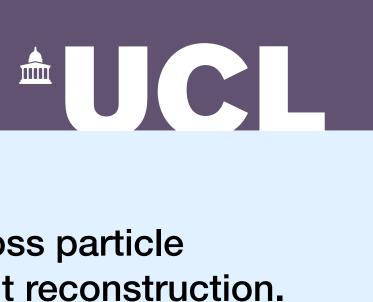
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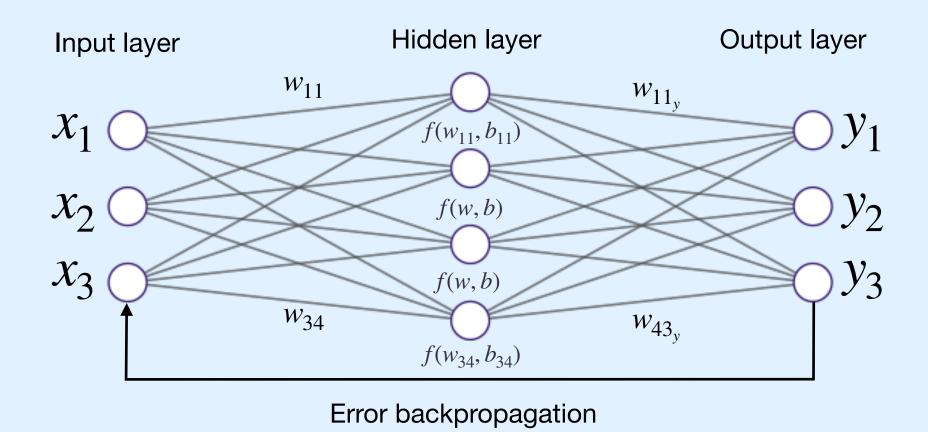


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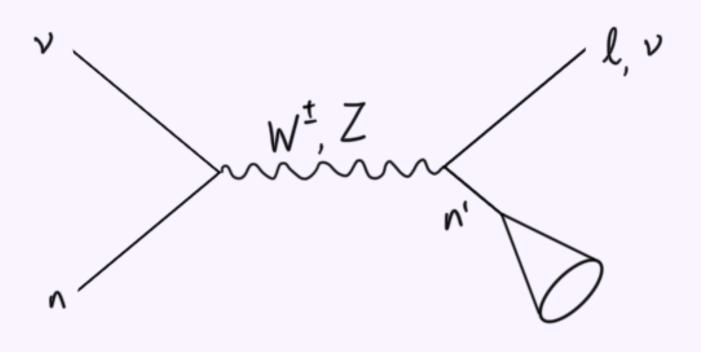
Train a neural network to reconstruct neutrino energy. Use dataset with detector constraints applied. Test performance for beam and atmospheric neutrinos.







Dataset and Network



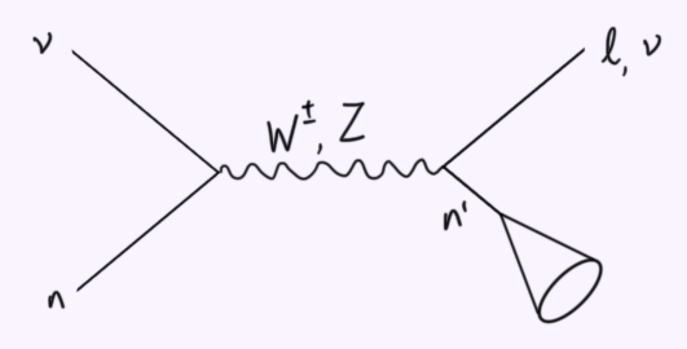
Generate $5 \times 10^5 \nu_{\alpha}(\overline{\nu_{\alpha}})$ events with energies from 0.2 - 6 GeV using NuWro.

Use DUNE flux for beam neutrinos, flat flux for atmospherics.





Dataset and Network



Particle	Minimum K.E.	Angular Uncertainty	Energy Uncertainty
Proton	$30 \mathrm{MeV}$	10°	10%
$\operatorname{Pion} \setminus \operatorname{Kaon}$	$30 { m MeV}$	10°	10%
Neutron	$30 \mathrm{MeV}$	10°	40%
Λ	$30 \mathrm{MeV}$	10°	10%
μ^{\pm}	5 MeV	2°	5%
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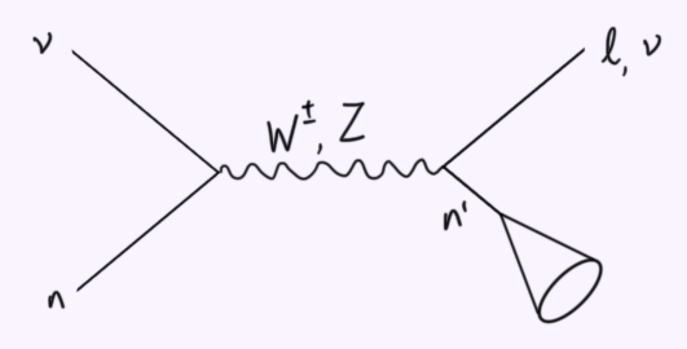
Randomly rotate events if training for atmospherics.

Remove invisible particles.





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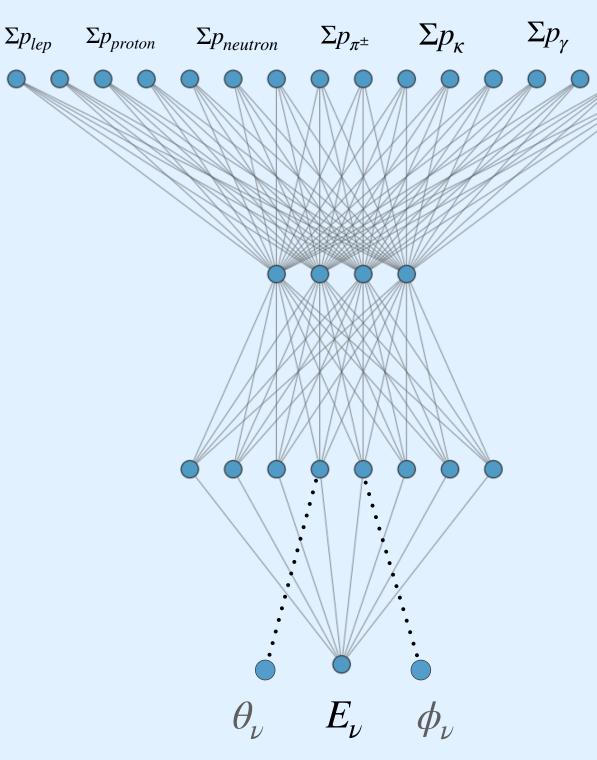
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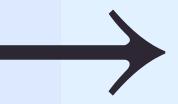
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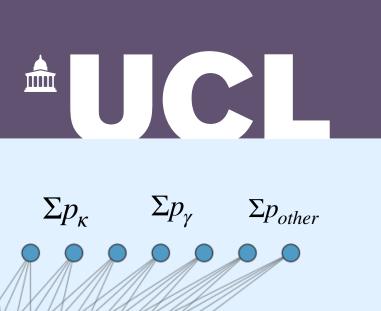
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Feed to fully connected layer regression network

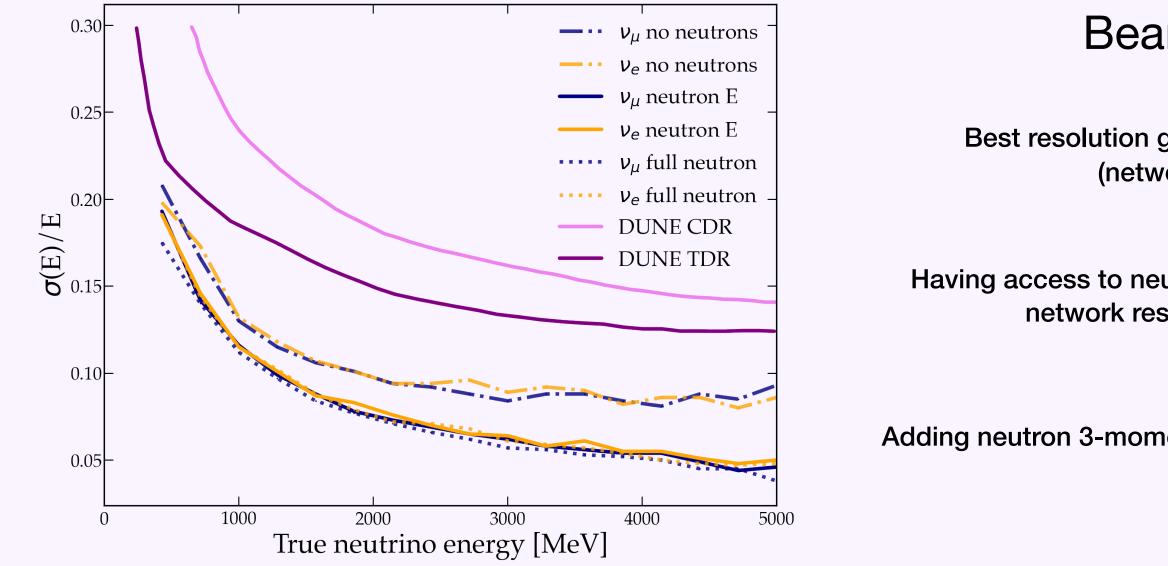








Results

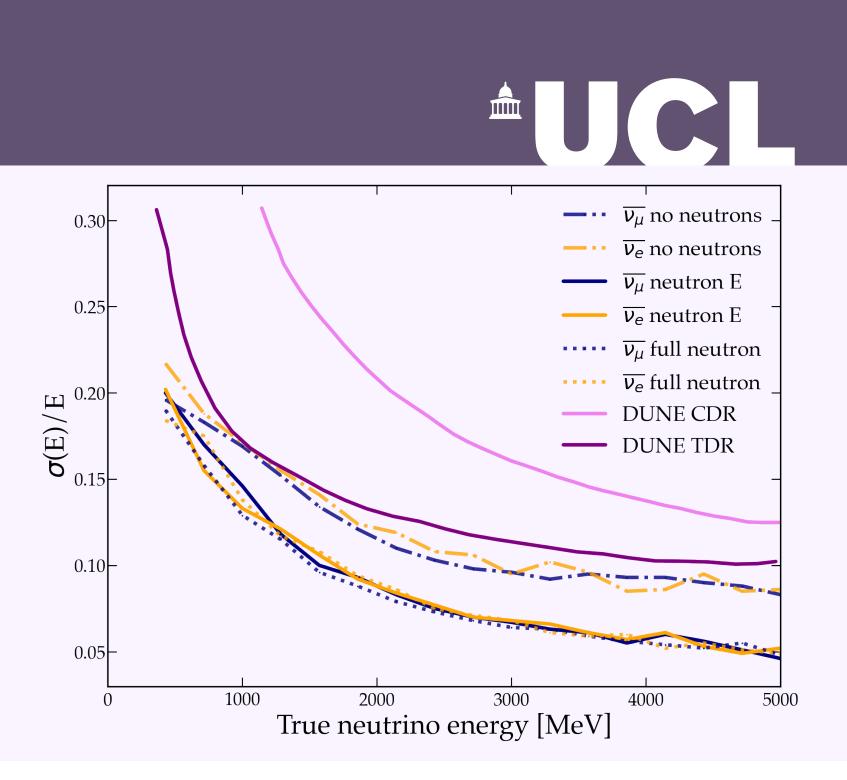


Beam neutrinos

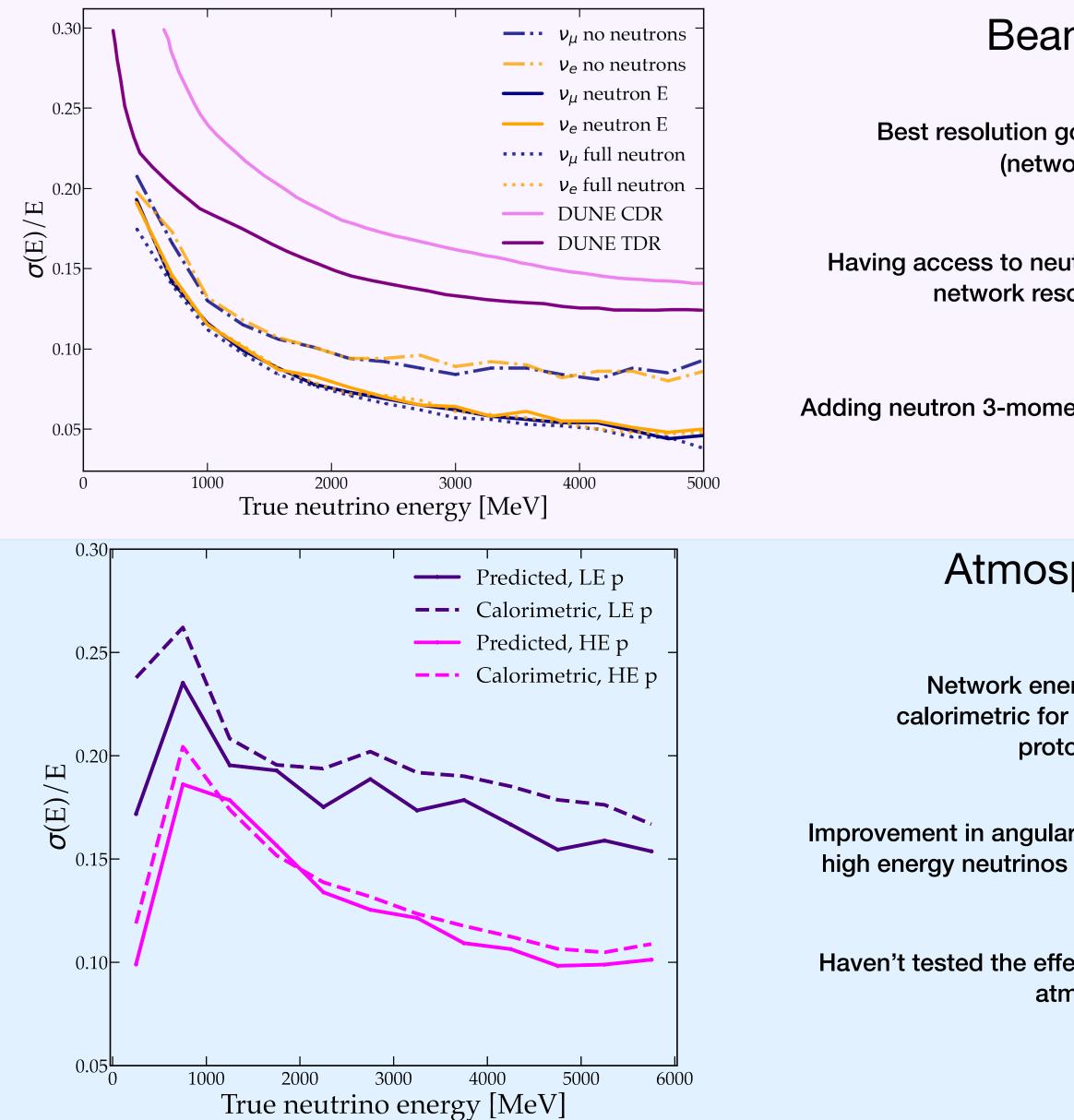
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Having access to neutron energy information improves network resolution by a factor of 1.5

Adding neutron 3-momentum yields minimum improvement



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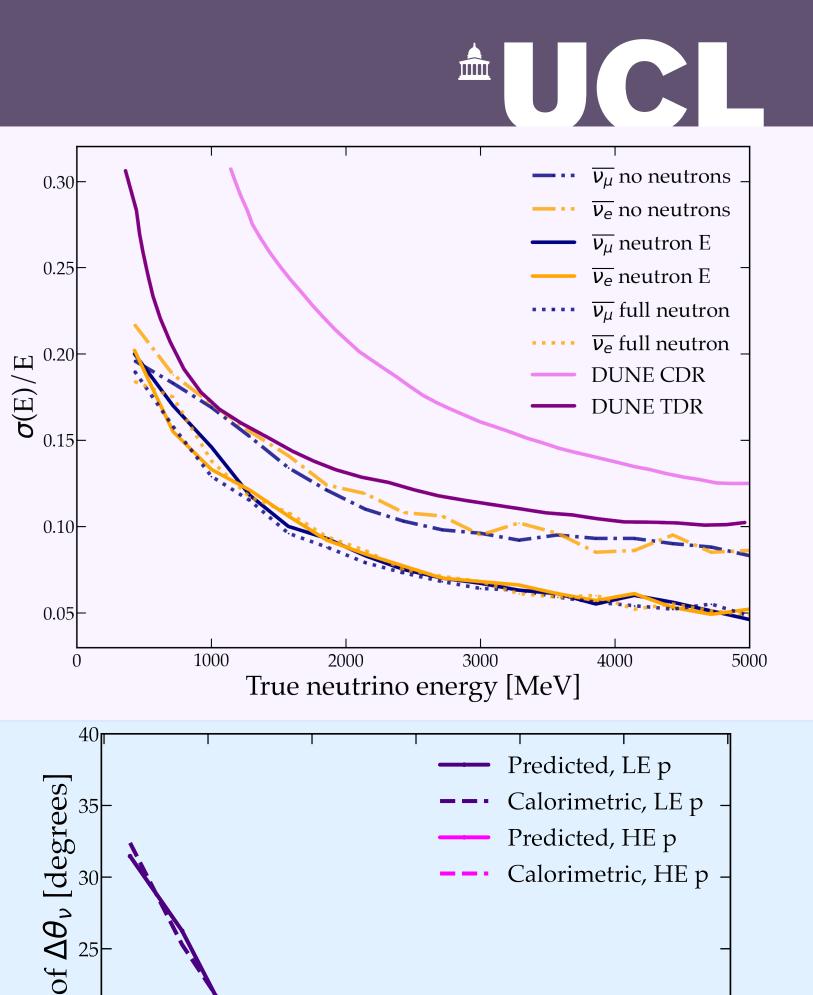


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Atmospheric neutrinos

Network energy resolution outperforms calorimetric for events without a high energy proton in the final state

Improvement in angular reconstruction is minimal except for high energy neutrinos with low final state hadronic energy

Haven't tested the effect of adding neutron information for atmospherics... yet!

