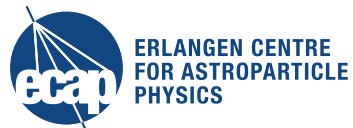


Flavor identification in ORCA

Thomas, Heid
MANTS Garching, 14th-15th October 2013





Outline

- Motivation
- Methods
 - Random decision forests
 - Used features (Examples)
- Results
- Conclusion



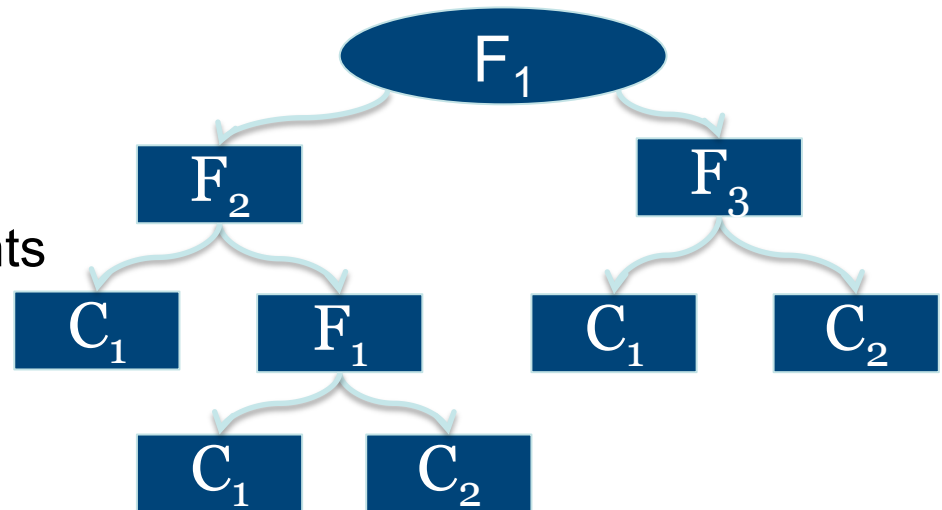
Motivation

- Best signal for ORCA in muon neutrinos



Random decision forests

- Random Decision Tree
 - Cut on feature F_x
 - Classify as class C_x
- Trained with Monte Carlo Events



Many different Trees \longrightarrow Random decision forest

- Define a majority for the decision

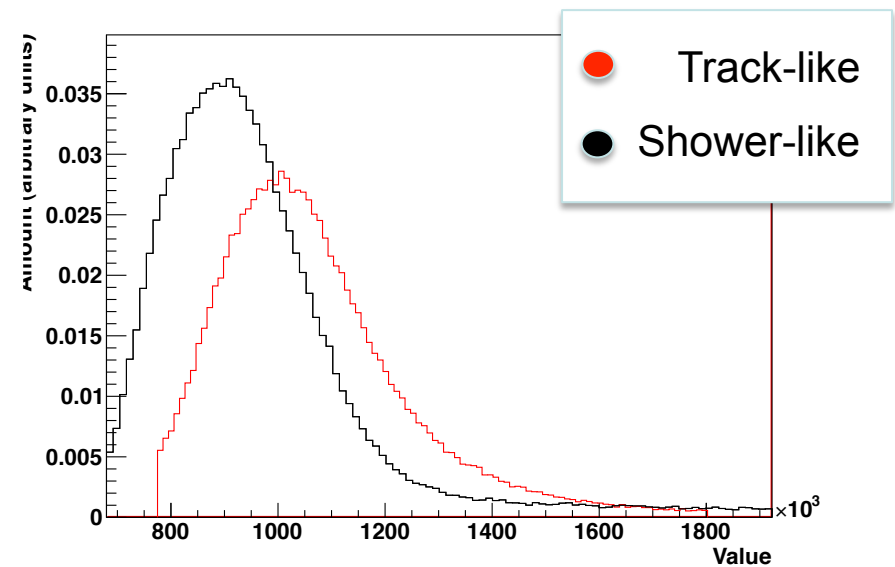
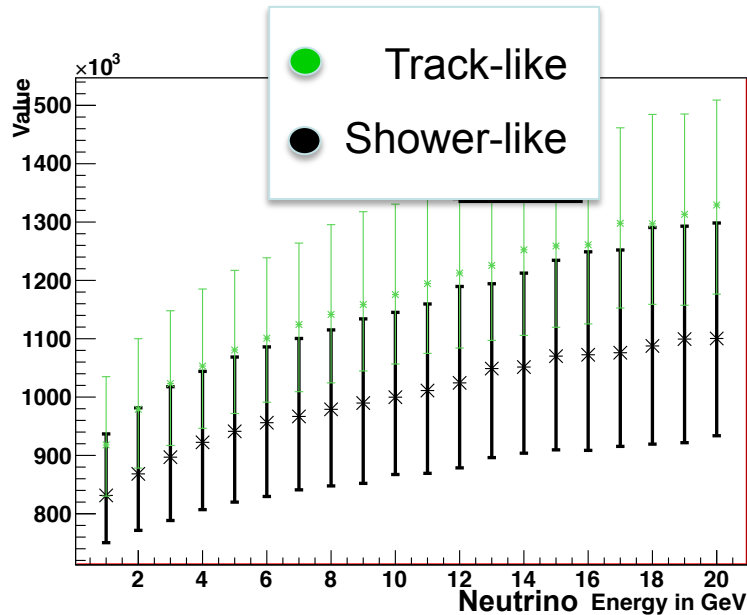
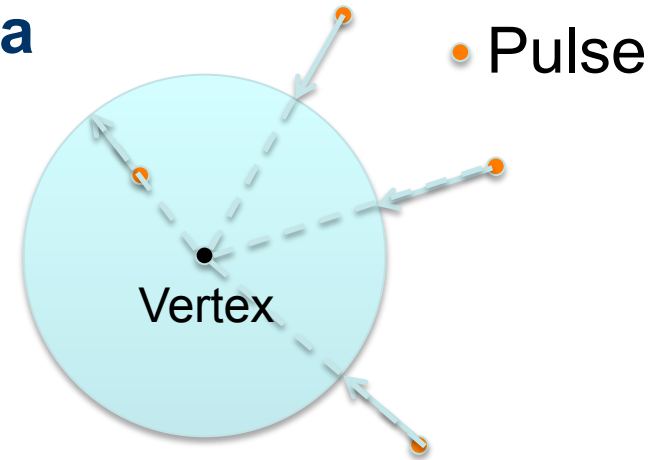


Used Monte Carlo Events

- Physical Events
 - Catania MC Production
 - Genie
 - Event selection:
 - Minimum 5 MC Hits
 - Vertex in detector
 $|z| < 50$; $r < 50$
 - No noise
 - Around 10000 Events
- Premium Events
 - Catania MC Production
 - Genie
 - Nearly all light is contained
 - km3v4r5
 - Geasimv4r12
 - Modk40seawiet
 - 300000 Events

Features (Example): Tensor Of Inertia

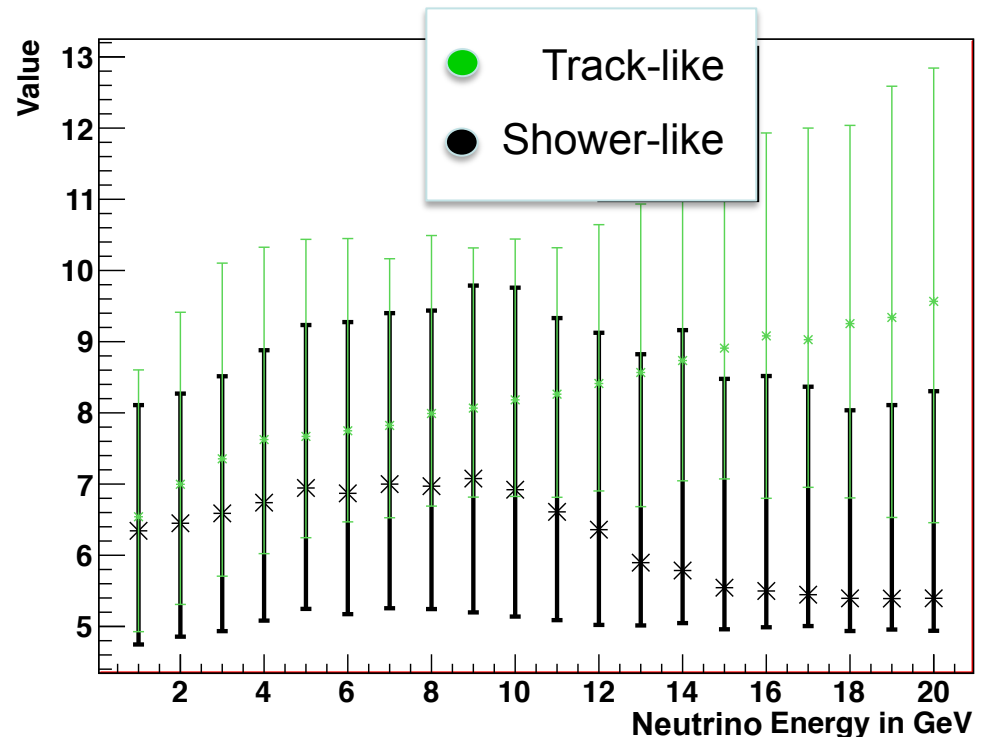
- Eigenvalues of tensor of inertia
- Projection of all pulses to sphere around reconstructed vertex



Features (Selection): QStrategy Quality Parameter

- Qstrategy shower reconstruction algorithm (ANTARES-PHYS-2012-11)
 - Based on time residuals under shower hypothesis

$$M = \frac{1}{N} \sum_i^N 2 * \sqrt{1 + \frac{(\Delta T)^2}{2 * \sigma^2}} + 2$$





Best of features

Fraction of number of hits in shower selection and track selection

QStrategy quality parameter

RMS of time residuals

FWHM of time differences

Number of Hits

Mean distance of hit from reconstructed vertex

Smallest eigenvalue of tensor of inertia

FilteringFit rlogl

...



Results



Definitions for Evaluation

$$Purity = \frac{N(\text{correctly classified track-like events})}{N(\text{events classified as track-like})}$$

$$Classification\ rate = \frac{N(\text{correctly classified track-like events})}{N(\text{track-like events})}$$

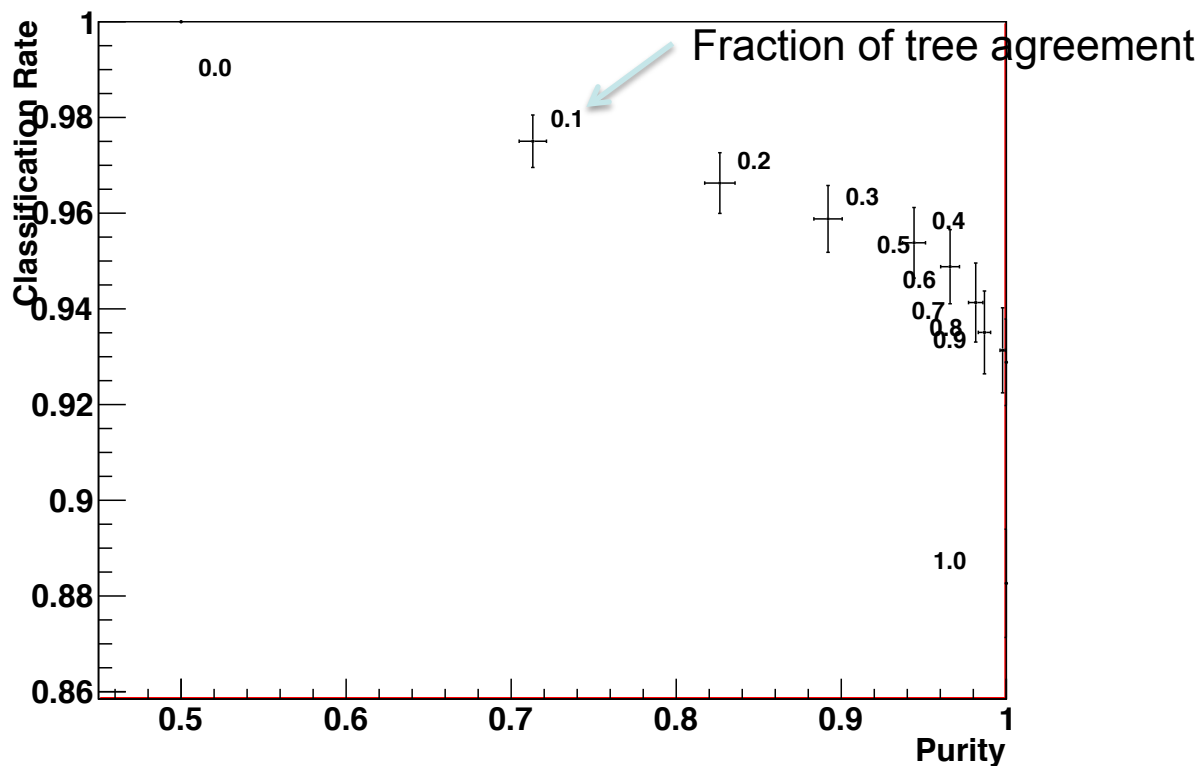




Physical Events without noise

- Event selection:
 - Minimum 5 MC Hits
 - Vertex in detector
 $|z| < 50$; $r < 50$
- No noise

Classification Rate versus Purity for 6 GeV

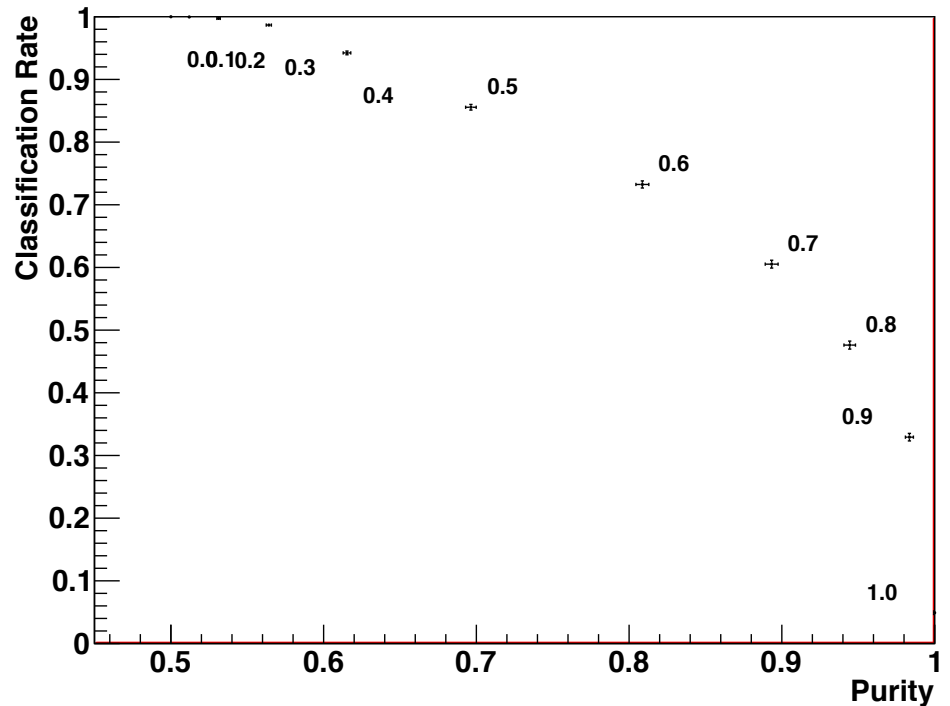




Premium Events with noise

- Nearly all light is contained

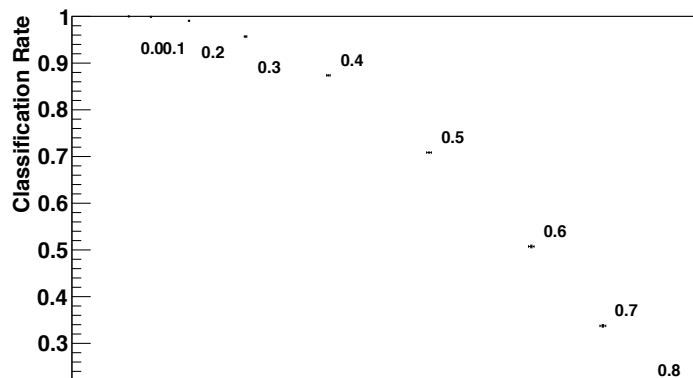
Classification Rate versus Purity for 6 GeV



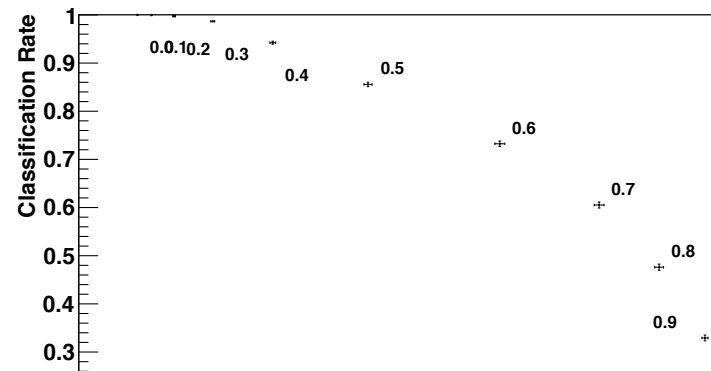


Premium Events with noise

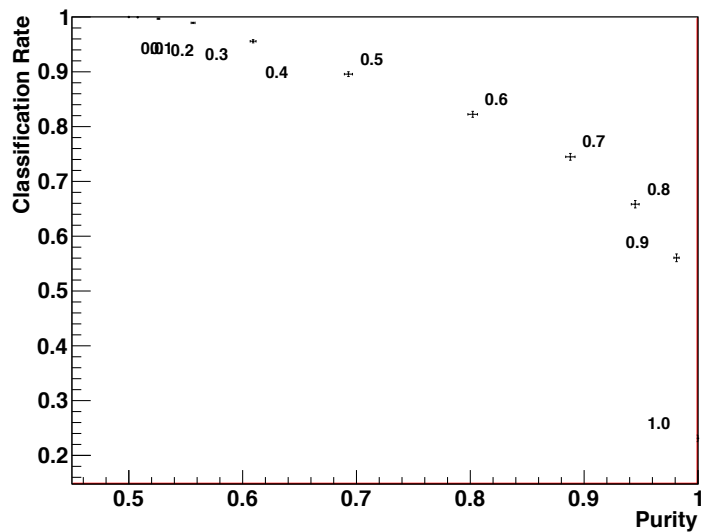
Classification Rate versus Purity for 3 GeV



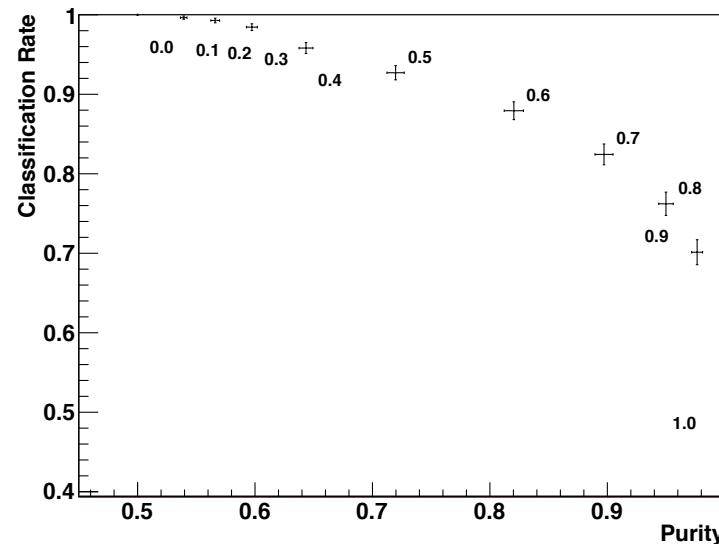
Classification Rate versus Purity for 6 GeV



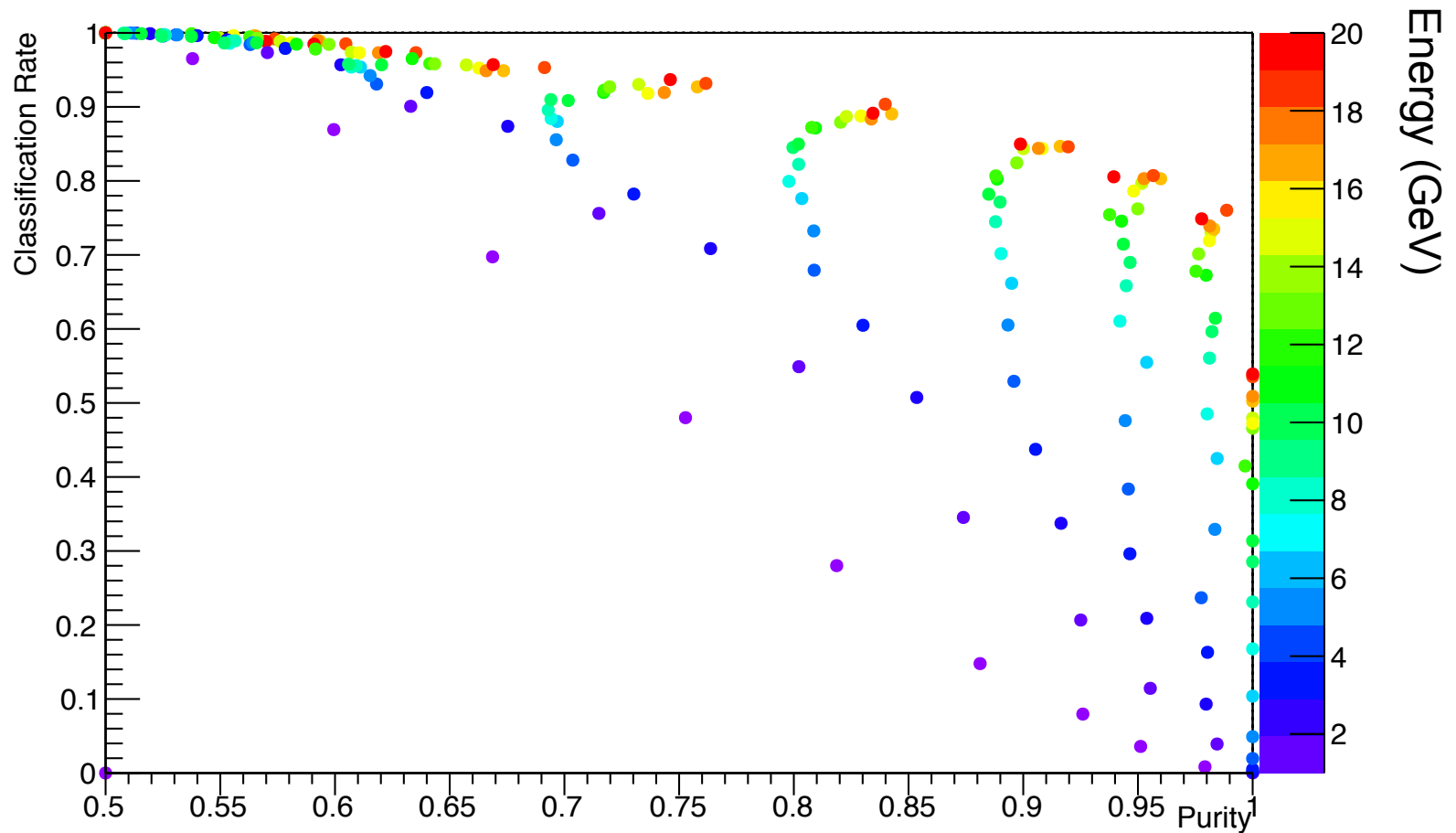
Classification Rate versus Purity for 9 GeV



Classification Rate versus Purity for 14 GeV



Premium Events with noise





Summary and Conclusion

- Random Decision Forest with about 30 Features
 - Properties of distribution of time residual
 - Tensor of inertia
 - Quality parameter of reconstruction
 - ...
- It is possible to identify track-like events
 - 80% Classification rate with over 80% purity rate
 - Without noise 95% classification rate with over 90% purity

Thanks for your attention

ecap



Bundesministerium
für Bildung
und Forschung



FRIEDRICH-ALEXANDER
UNIVERSITÄT
ERLANGEN-NÜRNBERG