

# Climb or Glide?

## Data Mining & Knowledge Discovery in Flight Records

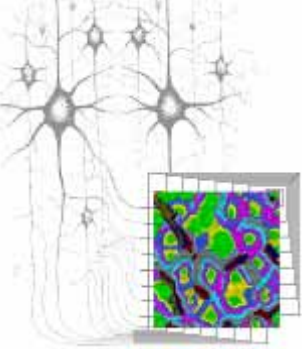
**Question:**

**What is the probability  
to find a thermal of certain strength  
on a typical cross country day?**

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**Arbeitsgruppe Datenbionik**

**Universität Marburg**



## The Data

**1635 Flight Records (IGC files)**

**Source: Online Contest (OLC) of BY, HE and TH  
and Coburg Competition**

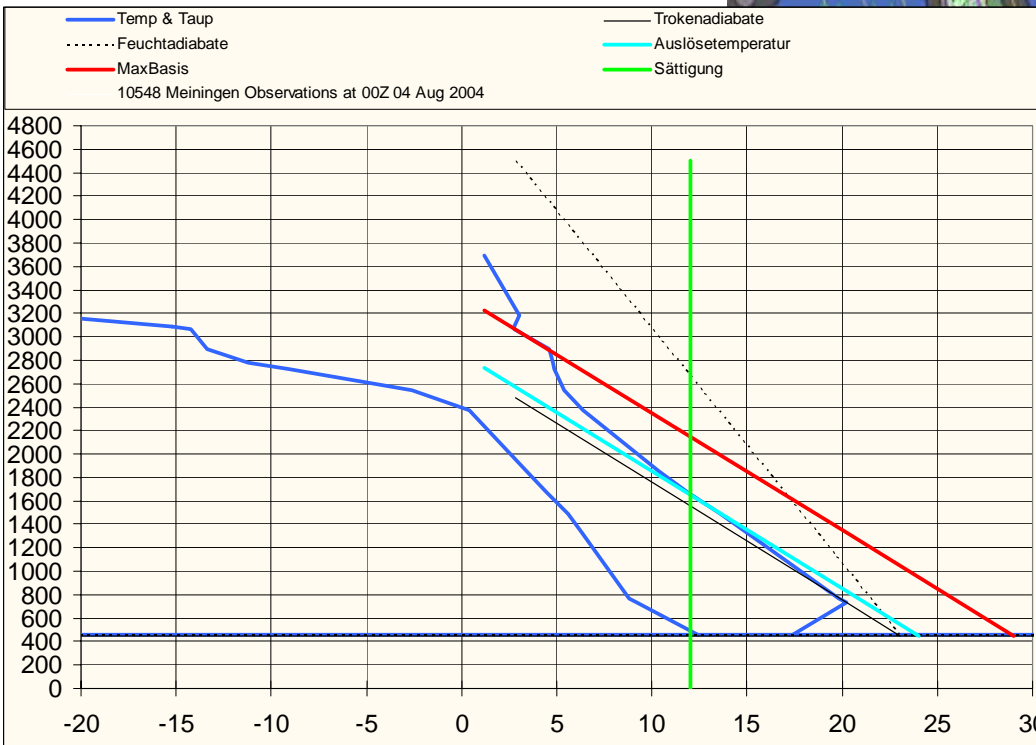
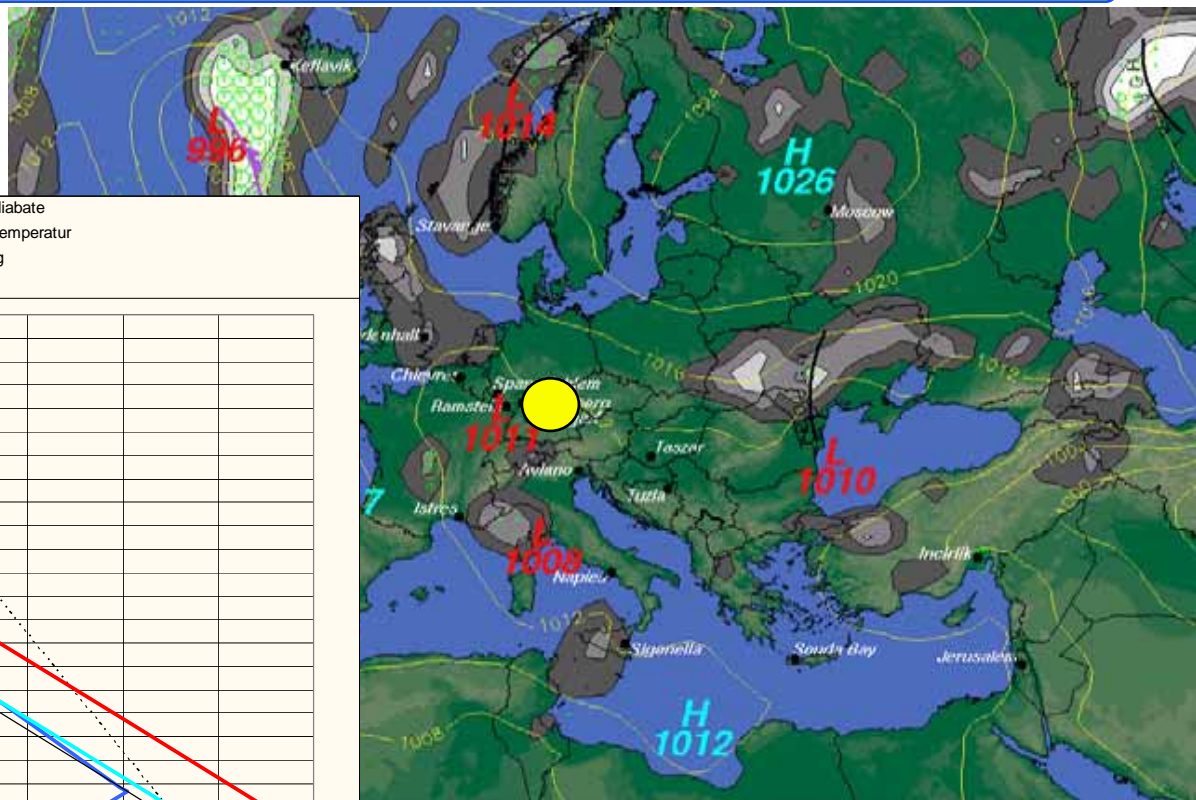
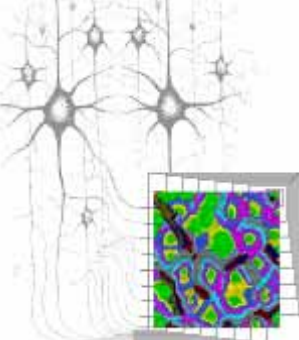
**Recording period :**

**July 30th and August 8 2004**

**= during Coburg Gliding Competition (Bayerische  
Segelflugmeisterschaften der Club- und  
Doppelsitzerklasse)**

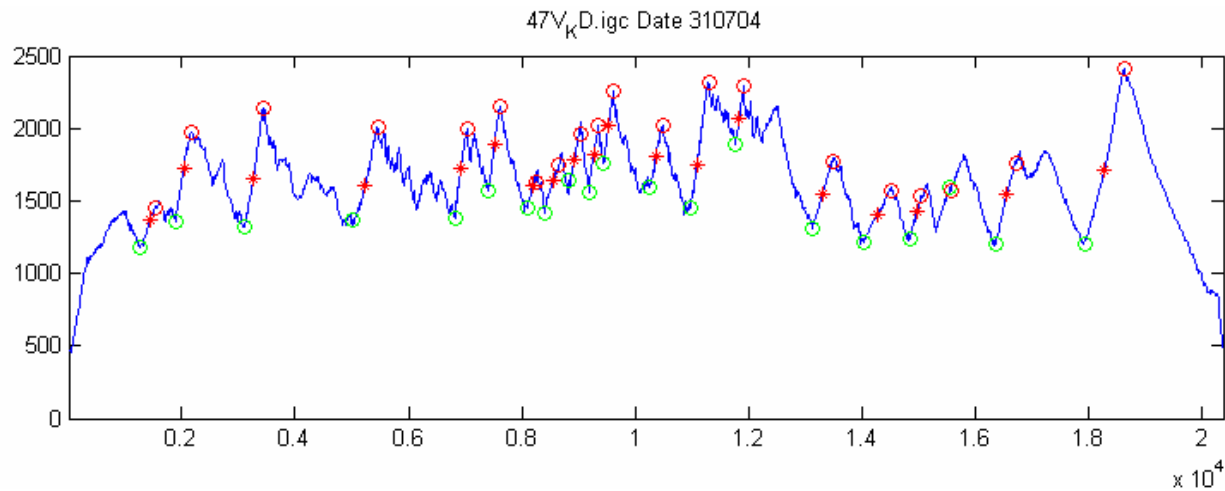
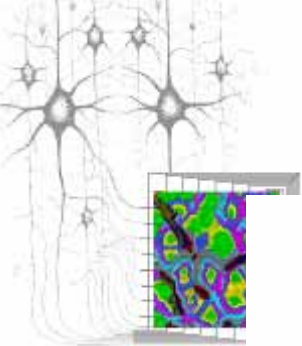
**Area: restricted to 49.5-51.2°N / 10-12.8°E**

# Weather during period



- mainly High pressure regime,
- dominated by East situation
- thermal strengths prognosed at Coburg (A.Ultsch): typ. 2m/sec and more
- competition tasks of 300-500km fulfilled by most pilots on all days

# Identification of thermals



## Problems:

- noisy height due to turbulence
- „dolphin“- flight stile
- Minimum altitute gain as paramter (250m)

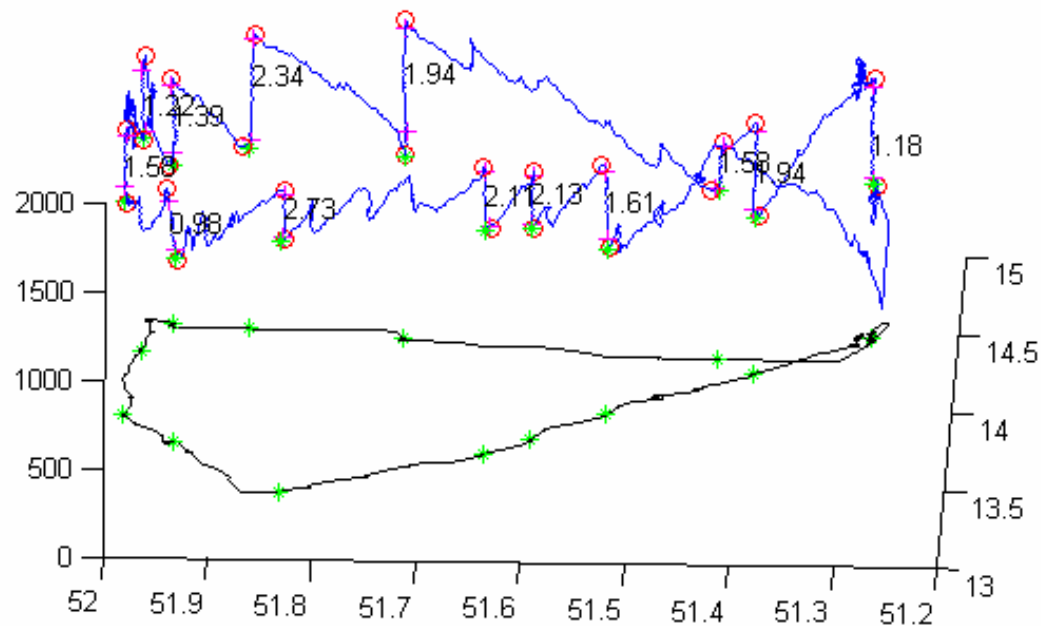
# Thermal Strength

Footpoints of 9677 thermals of the total 21695 thermals in area around Coburg

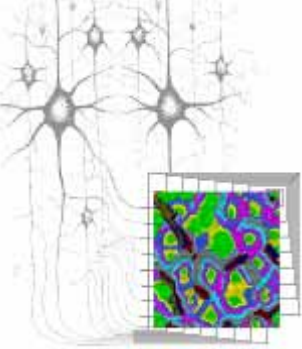
Filtered for uniqueness (no two pilots in the same thermal)

**ThermalCoreStrength = ThermalCoreHeightGain/ThermalCoreDuration;**

**Core = central 80% of thermal time**



## Estimation of data density



**Pareto Density Estimation (PDE)** a kernel based density estimation with fixed kernel (Parzen window).

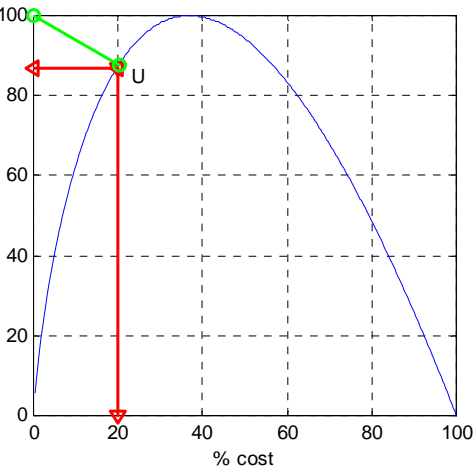
**Kernel such that entopic yiel of subset is optimized**

**Properties of Pareto Density Estimation PDE:**

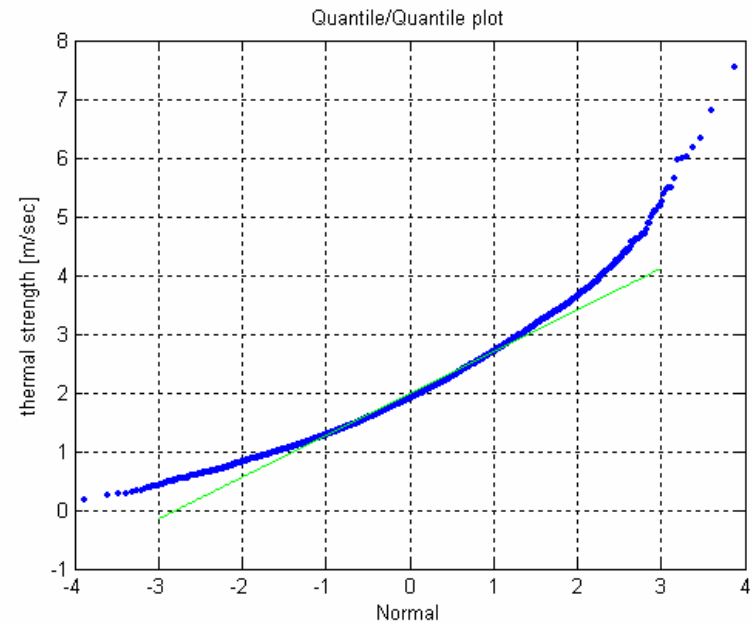
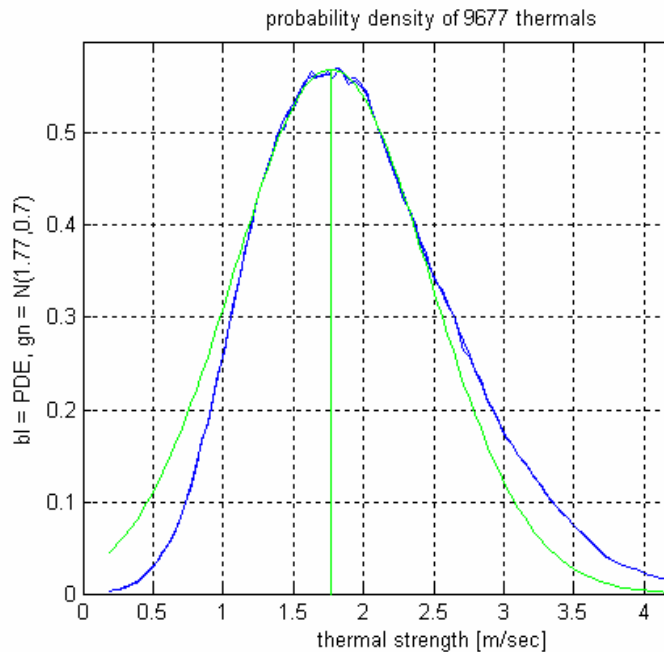
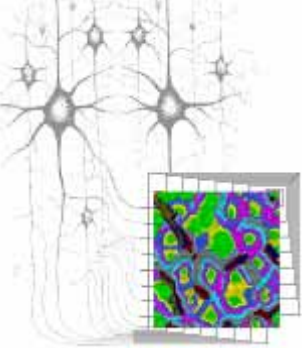
- **Optimal density estimation of Gauss mixture models (GMM)**
- **very good to analyze overlapping**
- **shows modal points**
- **in particular good for the detection of clusters**

**Applications:**

- **one dimensional data: PDEplot** –approximation of probability density
- **two dim data PDEscatter**
- **high dimensional data: P-Matrix**



# Distribution of thermal strength



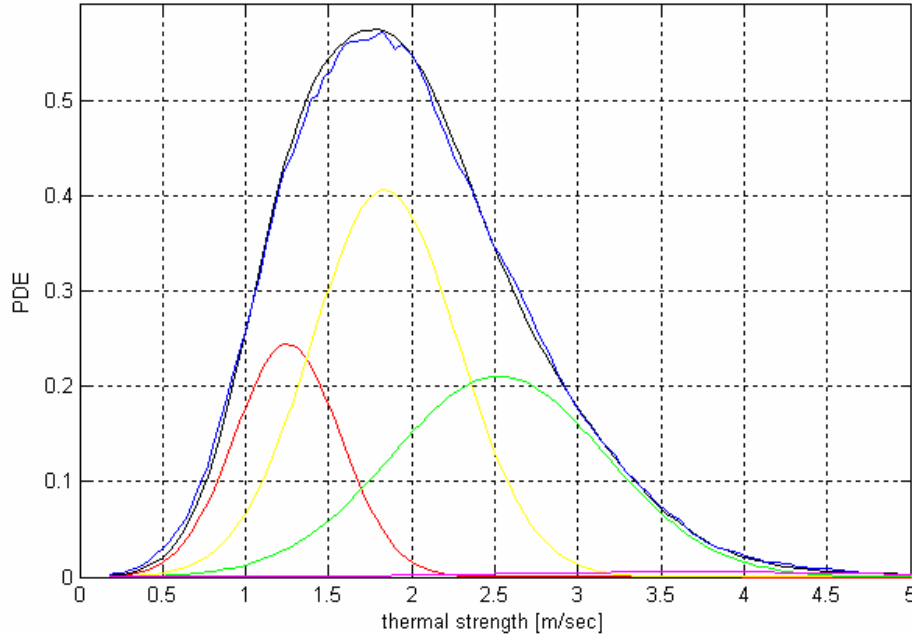
**Comparison to Normal distribution:**

⇒ **Definitive not a Normal (Gaussian) distribution**

⇒ **To many big ones , to few small ones**

⇒ **Smooth and systematic deviation from Normal distribution**

# Gauss Mixture Model

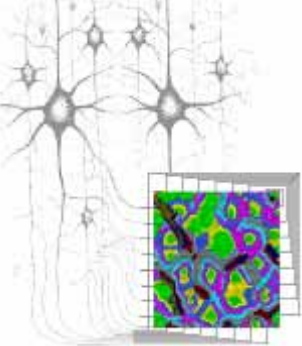


From 1...10 Mixtures wird tried´:

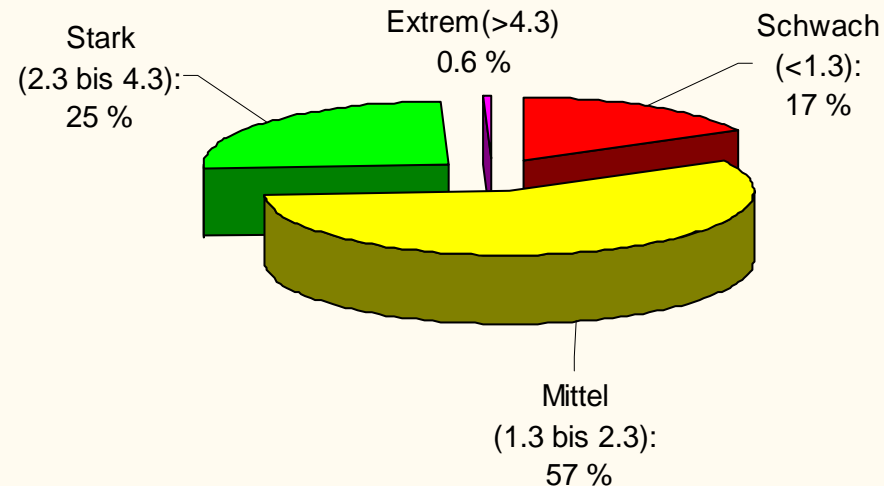
Best model has 4 Gaussians.

Mean	STD	Weight
1.25	0.32	0.19
1.84	0.44	0.45
2.52	0.64	0.34
3.69	1.18	0.02





## 4 Classes of Thermals?



**Maximum likelihood decision defines 4 classes of thermals:**

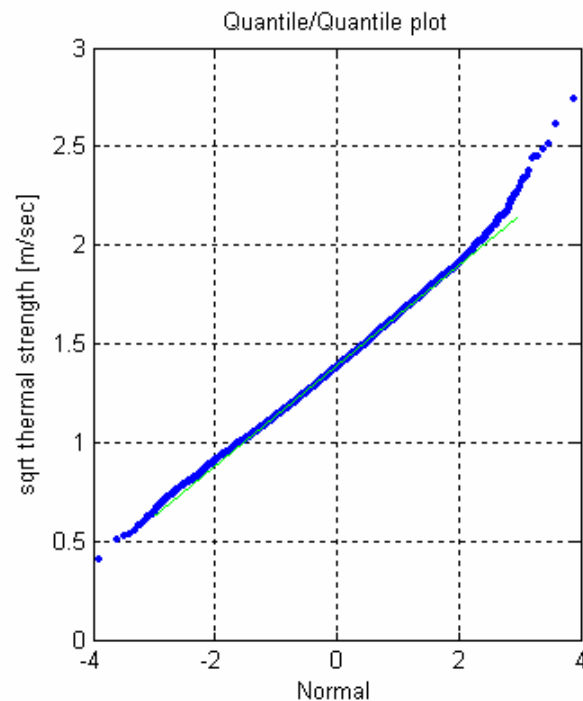
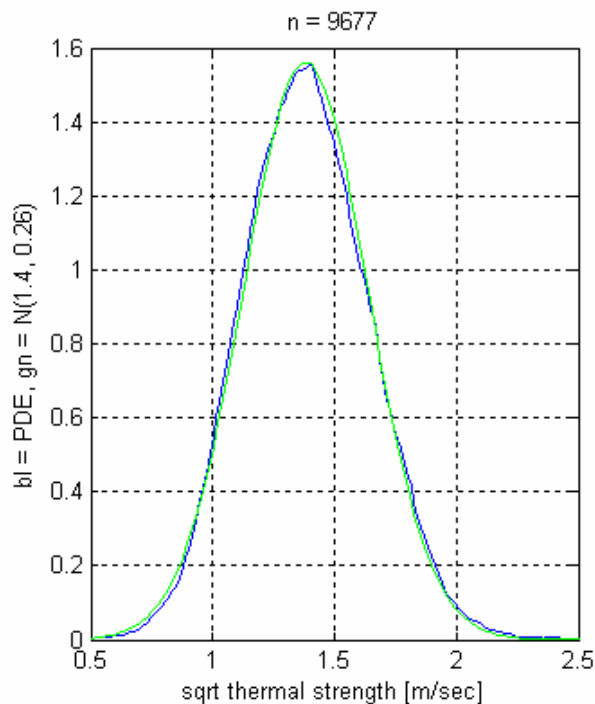
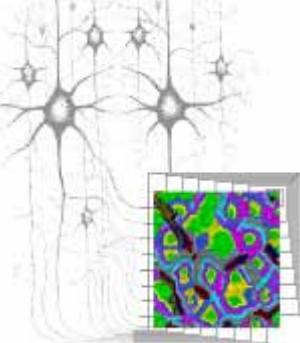
**Weak** <1.3 [m/sec]

**Average** 1.3 ... 2.3[m/sec]

**Strong** > 13 [m/sec]

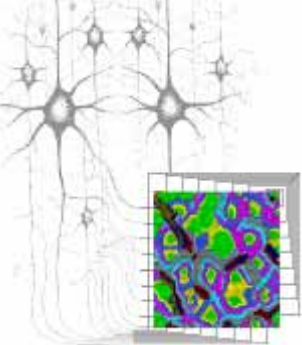
**Extreme** > 4.3 [m/sec]

## Square root of thermal strength

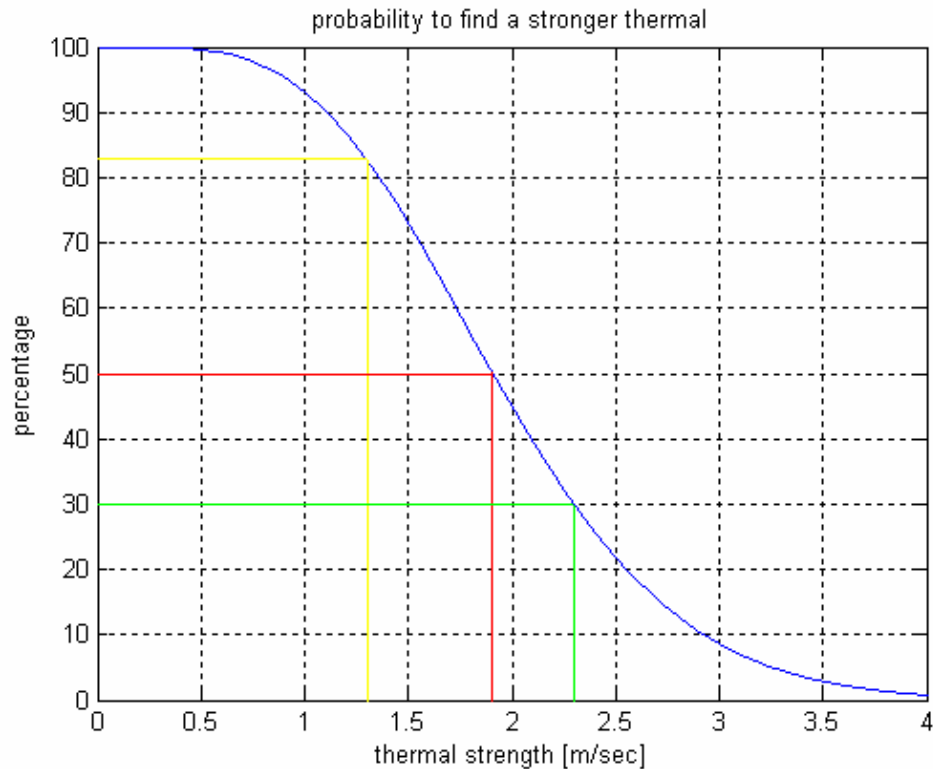


**Surprising fit to Normal distribution !**

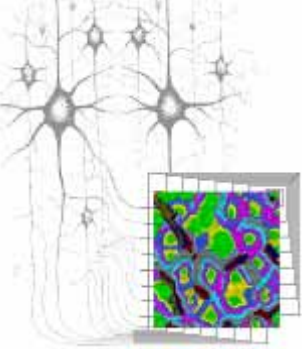
**=> Simple model of probability to find a thermal**



# Probability Distribution of Thermals

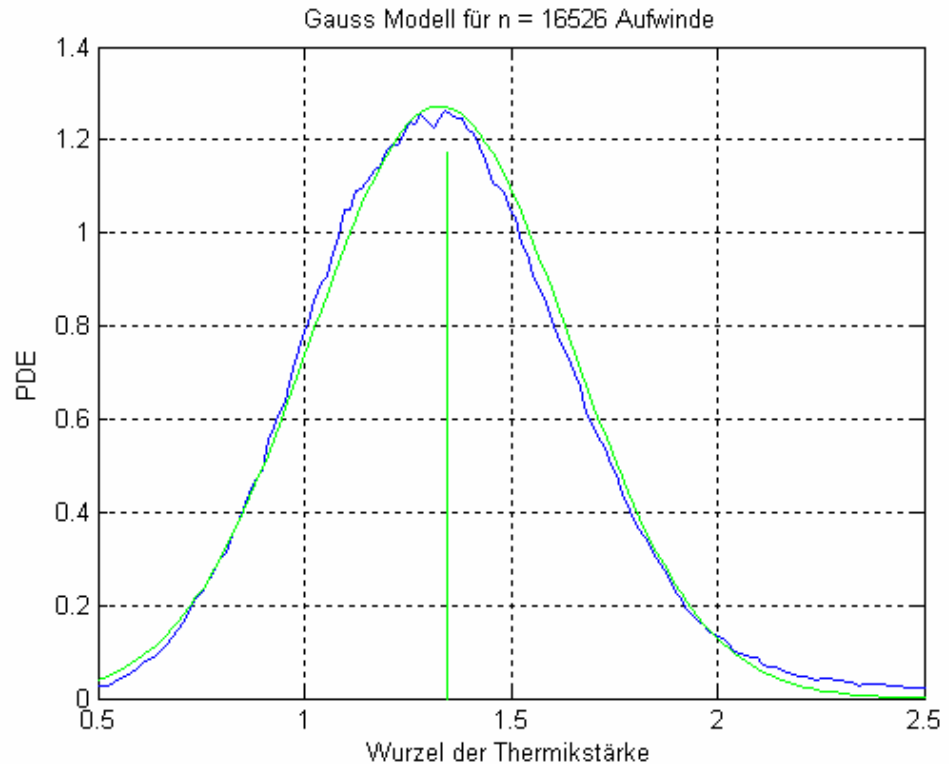


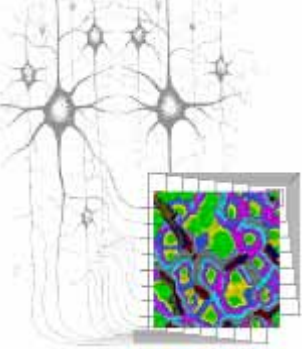
**Model: square root Normal distribution**



## Discussion: What is the reason behind square normal distribution?

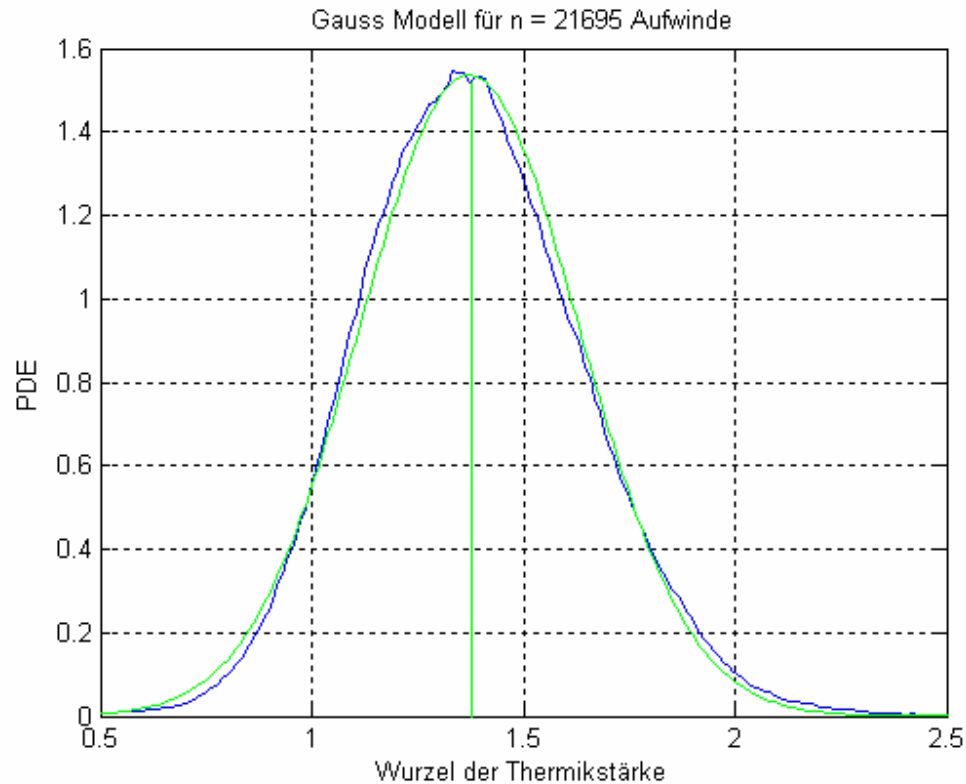
- 1) data processing:** other authors (e.g. Jon Meis Lüsse 1996) used 80 m as min height gain to identify thermals in IGC files
  - **Distribution of  $\sqrt{\text{CoreStrength}}$  using this limit:  $n=16,526$**



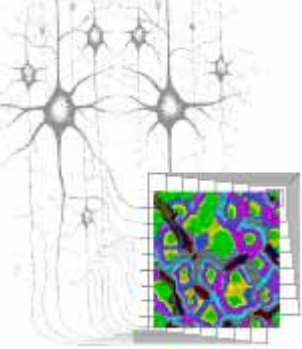


## Discussion: What is the reason behind square normal distribution?

- 2) **Bias by glider pilots:** they use only the strong thermals
  - **Distribution of all 21,695 thermals  $\sqrt{\text{CoreStrength}}$**



**=> Model fits the data!**



## Discussion: What is the reason behind square normal distribution?

### 3) Meteorological reasons ?

⇒ Left to discuss here

#### Proposal:

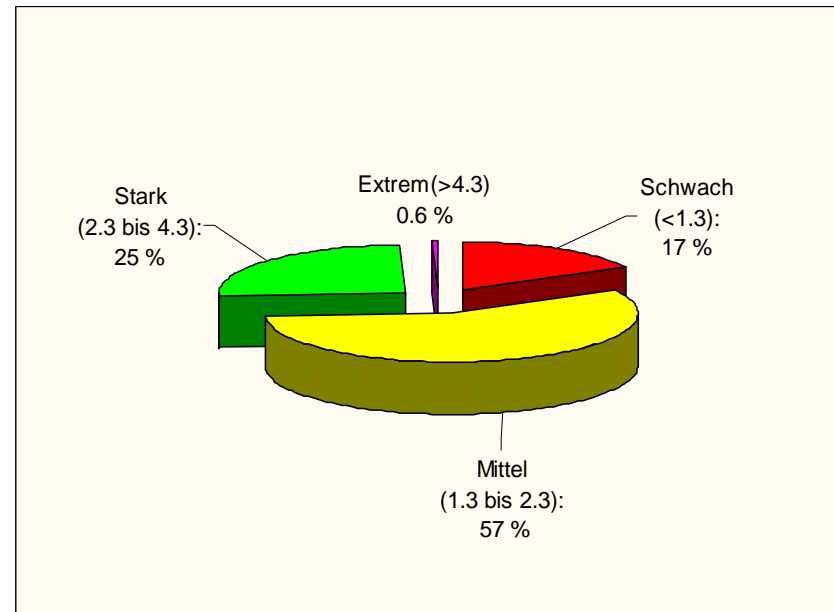
- Nature generates thermals using a solar heating plate .
- The diameter  $D$  of this heating plate is drawn from a Normal distribution Gauss( $m, s$ )
- The strength  $S$  of the thermal in m/sec is direct proportional to area of the heating plate:

$$S = c D^2 \text{ 🙌} \quad \text{with } D \text{ from Gauss}(m, s)$$

## Application

- Compare distribution of thermals of a flight to expected distributions

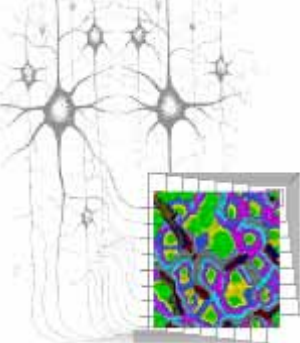
e.g 1/4 should be strong thermals



- give precise semantics to „weak“, „average“, „strong“ in gliding forecasts

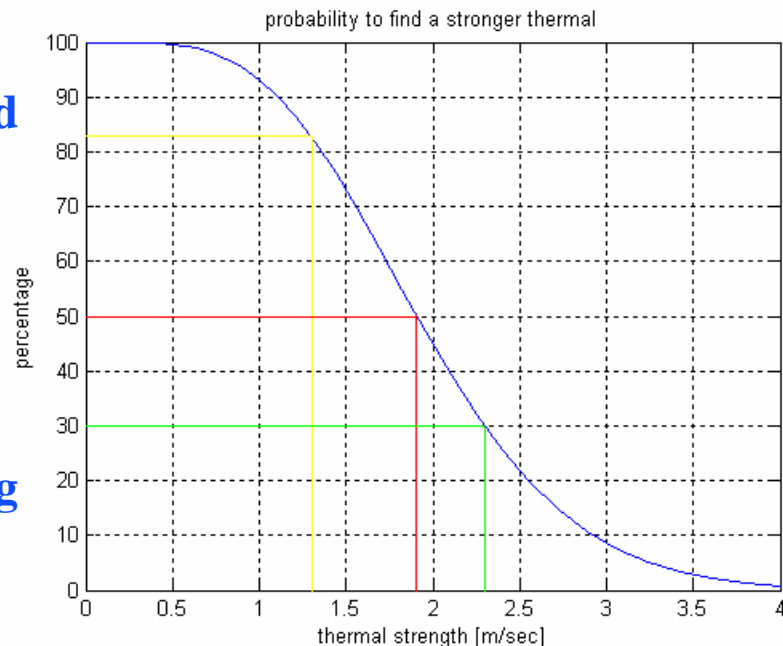
- model can be used for prediction

# Rules of Thumb



derived rules of thumb for cross country soaring:

- Don't use thermals below 1 m/sec. The probability to find a better one is over 90% !
- Go for thermals with at least 1.4 m/sec
- If the integrator shows less than 1.9 m/sec consider flying on. The probability to find a better thermal is > 50%
- Definitivly take thermals with 2.3 or better !
- Use thermals < 1.3 m/sec only in difficult situations (wether/ outlanding)

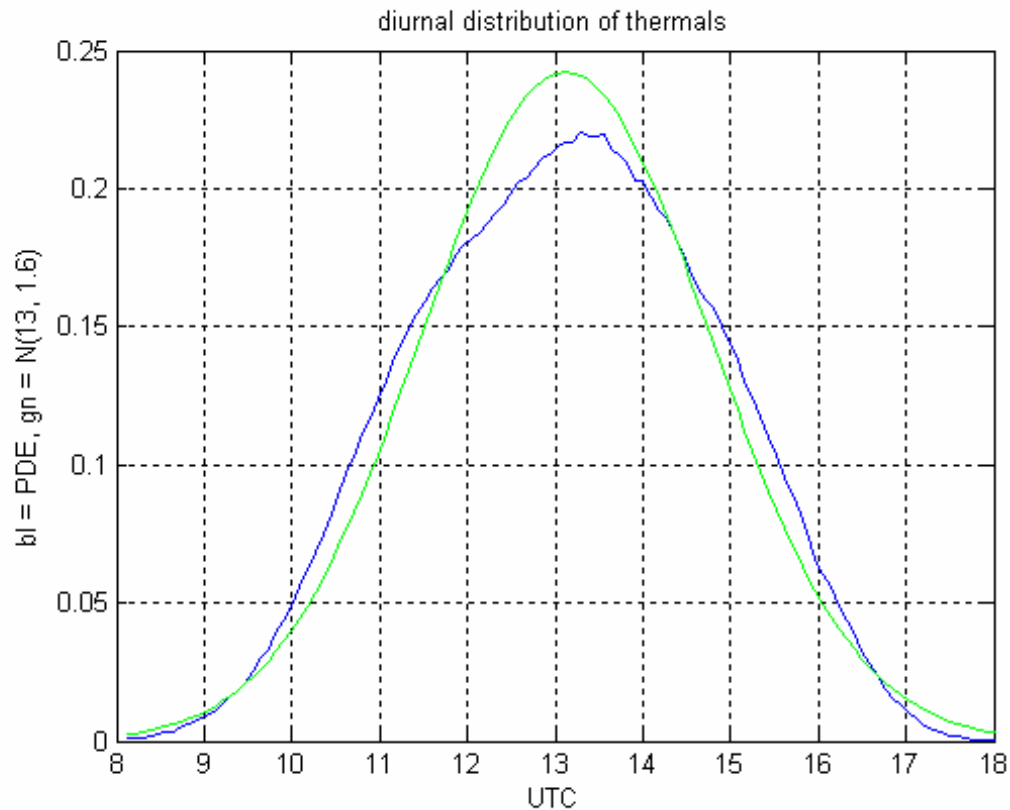




## Future Work

### 1) Diurnal Distribution of Thermals

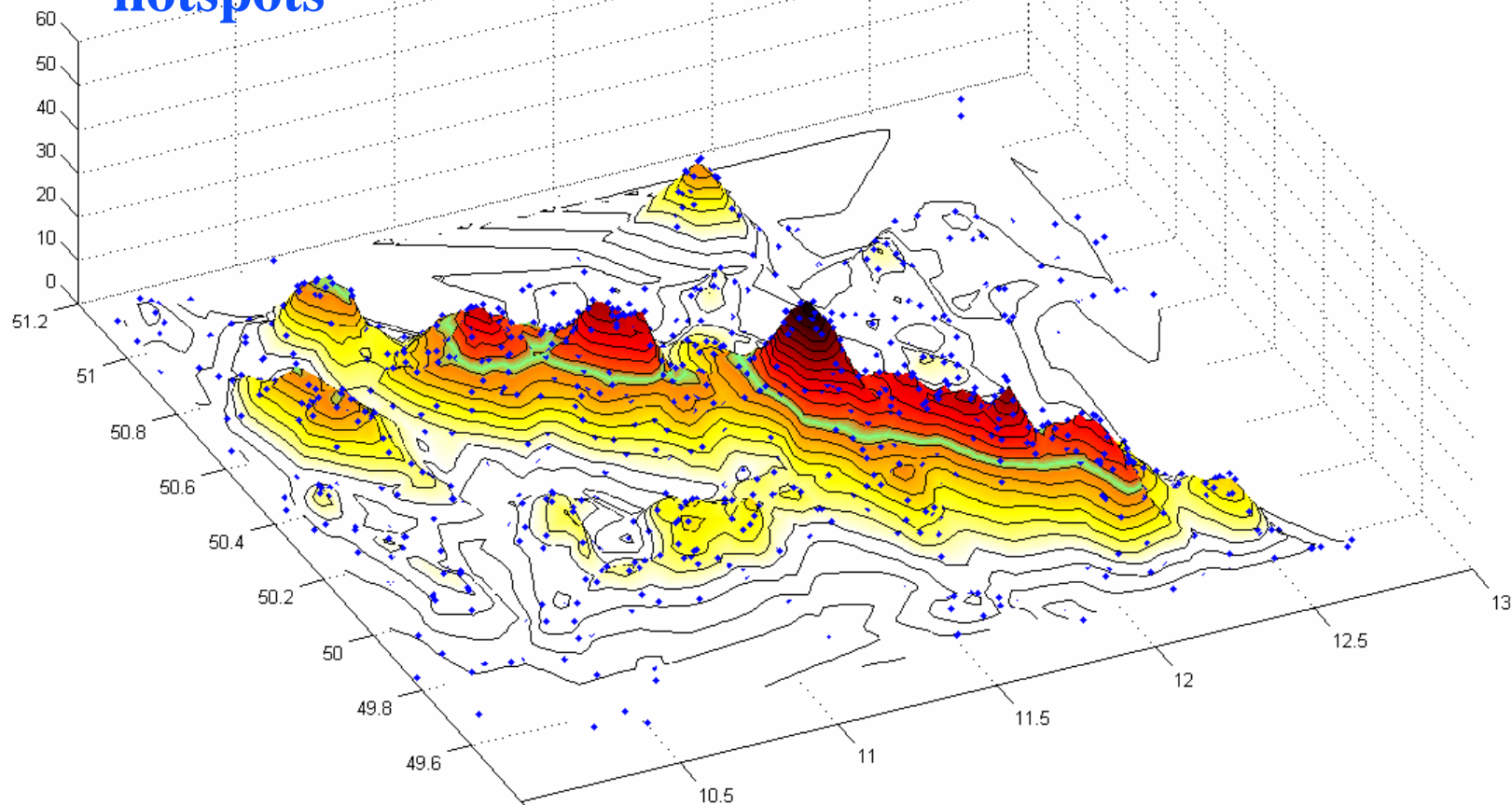
A first look at circadian distribution of occurrences:



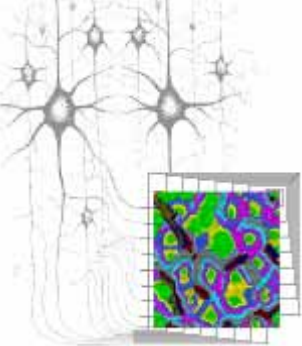
## Future Work 2: Are there Hotspots?

PDE-scatterplot,  $r = 0.1$

**Different approach than Enderle/Leykauf (DWD):  
Measure thermal densities (PDE scatter plots) to find  
hotspots**



**3) Are there Hotspot points in the Alps (sponsors sought)?**



## Summary

The empirical distribution of a large collection of thermals in typical cross county areas /weather is consistent with either

- 1) four types of thermals  
(small, average large, extra) or
- 2) a Sqrt-Normal Distribution

Open question: Is there a meteorological law behind this findings?

(e.g.:  $S = c D^2$   with D from Gauss(m, s))