



# Evaluation of accelerometer as an effective tool to measure sheep behavior in a pastoral context

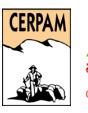
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- ▶ <u>Issue</u>: adapt technologies(GPS, accelerometer) to the needs of pastoral farmers in order to support their activity and improve the utilization of rangelands.
- ▶ <u>Objective</u>: determine the technical and functional specifications of a tool combining GPS and accelerometer, to be put on the animals
- In the framework of: UMT Pasto & RMT Travail en élevage







### Context



- Accelerometer technology not a new tool
  - ▶ Since the middle of the 90's, used to characterise animal behavior
  - ▶ Different animal species especially since 2000 (Shepard et al. 2008)
  - ▶ On sheep in several studies (Mason et al., 2013; Marais et al., 2014...):
    - ▶ Up to 5 behavior identified, on grassland
- Pastoral farming issues
  - ► Efficient pastoral resource management
  - Strong and compact device
  - Good battery operating time
  - Efficient in communications
  - Which individuals to equip?







### An applied project, with a variety of partners





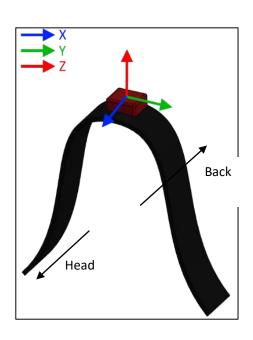
## Pastoral farming issues

- ► Efficient pastoral resource management
- Strong and compact device
- ► Good battery operating time
- ► Efficient in communications
- ► Which individuals to equip?





## WPK 2.1: Evaluation of accelerometer as a tool to characterize animal behavior



#### Accelerometer

- 3 axis
- $g = 9.81 \text{m.s}^{-2}$
- Acquisition frequency: 100Hz







1.5 days on grassland + 1.5 days on rangeland

**3 different animals** for each ½ day





#### Acquisition of data in the field



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3 different animals for each ½ day

#### **Direct observation** of 9 types of behaviors

Lying -	Standing -	Standing -
Sleeping	motionless	ruminating
Lying -	Standing -	Standing -
motionless	walking	eating brush
Lying - ruminating	Standing - running	Standing - grazing

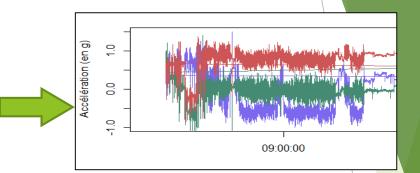






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ruminating	running	grazing



Time	Behavior
09:00:07	Standing - Grazing
09:02:00	Standing - Walking
09:02:10	Standing - Grazing
09:02:38 to 09:03:12	Standing - Walking
09:03:29	Standing - Walking
09:04:03	Standing - Grazing
09:05:15	Standing - Grazing





Statistical

treatments

Acquisition of data in the field



-1.0 0.0 1.0 0.00:000

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### Method: combination of Data sets



IUCIC					
Time		Acceleration axis x (g)	Acceleration axis y (g)	Acceleration axis z (g)	Behavior
09:00:05	000	0.53125	0.125	0.75	GRAZING
09:00:05	001	0.53125	0.125	0.75	GRAZING
09:00:05	002	0.5625	0.125	0.78125	GRAZING
•••					
09:00:05	099	0.5625	0.09375	0.78125	GRAZING
09:00:06	000	0.5625	0.09375	0.78125	GRAZING
•••					
09:00:06	099	0.53125	-0.125	0.7735	GRAZING
09:00:07	000	0.53125	-0.125	0.75	GRAZING
•••					
09:00:09	099	0.53125	-0.125	0.7735	GRAZING
09:00:10	000	0.53125	-0.125	0.75	GRAZING
•••					
09:01:59	099	0.475	-0.09375	0.78125	GRAZING
09:02:00	000	0.475	-0.225	0.75	WALKING
09:02:00	001	0.325	0.375	0.75	WALKING
09:02:09	099	0.5625	0.375	0.75	WALKING



09:02:09

099

0.5625

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•••	•••				•••

0.375

0.75

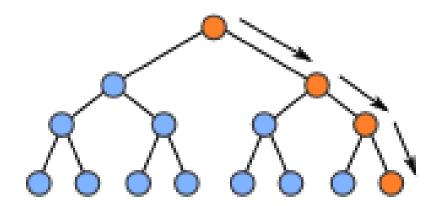
**WALKING** 



# Method: Statistical treatments with the Random Forest algorithm



- ✓ Treatment of a pool of 3500 x 5-second segments with only 1 behavior.
- ✓ Data analysed: 100Hz and 25Hz
- CART method: Classification and Regression Tree
  - Classifies segments in subsets to discriminate them
  - ✓ At every node of the tree, the algorithm looks for the best statistical variable to discriminate segments and to divide them in 2 subsets (according to this variable)





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#### ✓ Random Forest:

- ✓ 500 regression trees, 3500 segments per tree (sampled with replacement), 6 variables per tree
- ✓ 18 variables to discriminate segments (median, average, standard deviation, min, max, ...)
- Output: rate of success in segment classification



### Results: rate of success in segment classification



#### Grassland and rangeland data, 25Hz frequency

Behavior observed	Number of analysed segments	% of good predictions	Confusions with
Lying - sleeping	1415	94.8	Lying - motionless
Lying - motionless	2362	92.9	Lying - ruminating
Lying - ruminating	1658	90.7	Lying - motionless
Standing - grazing	5024	98.0	
Standing - ruminating	292	67.5	Lying - ruminating
Standing - eating brush	9	0.0	Standing - grazing
Standing - motionless	244	28.7	Standing - grazing and Lying - motionless
Standing - walking	132	62.1	Standing - grazing
Standing - running	117	73.5	Standing - walking

- → 92.4 % of segments correctly predicted
- → Confusions between some behaviors



### Results: rate of success in segment classification



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### **Conclusions**

- ▶ 25 Hz vs 100 HZ: same results → reduce the data volume
- ▶ 92.4 % of good prediction → very good prediction for lying and grazing behaviors
- ▶ To be done next:
  - ► Improvement of the algorithm with new sequences (rangeland pasture)
  - Validation of the algorithm on other flocks and other kinds of pastures
  - Determination of the number of segment per minute (or per hour) necessary to predict correctly the behavior





### Perspectives

- ► Efficient pastoral resource management
  - ▶ For pastoral farmers, combined to GPS data:
    - ► Better rangeland management
    - ► Alerts: dangerous areas or abnormal behavior (predator attack)
- Strong and compact device
- Good battery operating time
- Efficient in communications
- Which individuals to equip?



## Thank you!



