Catching Drift and Anomalies in Food sales Series

Jorn Bakker





Netherlands Organisation for Scientific Research

Technische Universiteit **Eindhoven** University of Technology

Where innovation starts

TU

Stock balancing





/ name of department

Overview

- Food sales prediction..
- ...and its problems
- Online learning cycle
- Outlier detection
- Adaptive windowing
- Experimental setup
- Results
- Future













t	History	Temp	Holiday	Promo	
1					External Features
					r catares
i	{y(i-n) y(i-1)}				
n					U/e Technische Universiteit Eindhoven University of Technology
			•		

Problem





Problem statement

"Find out whether the behavior of the series change with respect to history, as accurate as possible and as soon as possible"



Online Learning Cycle



Online Learning Cycle



Online Learning Cycle





If for any new instance **y** at time **t**:

$$y_t > |\mu_{1..t-1} + 3\sigma_{1..t-1}|$$

It is replaced by: $\mu_{1..t-1}$



Adaptive windowing (ADWIN A. Bifet and R. Gavaldá. '07)



27-10-2010

PAGE 13

/ name of department

Experimental setup

- Set of 439 products
- Manually labeled
 - behavior is drifting or not
 - Timestamp of occurence
- Warmup period of 60 weeks, the rest is used as test
- Each series is transformed to binary series by :





Results : Outliers





Results : ROC











Future

- Need for less lag
- Need for better outlier detection
- What about seasonal patterns?
 - •
- "Closed loop" evaluation



This slide is intended as the last slide



Some false positives





Some false negatives

