

auto**N**omous, self-Learning, **OPT**imal and comp**L**ete **U**nderwater **S**ystems **NOPTILUS**

FP7-ICT-2009.6: Information and Communication Technologies

1st Project Review

WP6 (Situation Understanding) Progress Report

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Outline

- ▶ **WP6: Situation Understanding**
- ▶ **Probabilistic Context-Free Grammars**
- ▶ **PCFG Event Recognition**
 - Proof-of-Concept Example
 - Multiple Experiments
- ▶ **WP6 Progress and Plans**

WP6

Situation Understanding

Situation Understanding

► **Definition**

- ❑ cognitive ability of inferring high-level descriptions and representations of the current state of the environment

► **Abilities**

- ❑ analysis of timed data
- ❑ event recognition

► **Observations**

- ❑ stream of discrete symbols

► **Events**

- ❑ patterns in observations

WP6 Tasks

▶ **Task 6.1: Probabilistic Context-Free Grammars**

- ❑ formal models for specifying sequences
- ❑ appropriate for recognizing patterns in observations

▶ **Task 6.2: Structured Prediction**

- ❑ step-by-step approach to building structured objects
- ❑ appropriate for automatically learning grammars

▶ **Task 6.3: Integration**

- ❑ PCFG event recognition using learned grammars

▶ **WP6 current status**

- ❑ significant progress in Task 6.1
- ❑ first steps in Task 6.2
- ❑ no work yet in Task 6.3

PCFG

Probabilistic Context-Free Grammars

Probabilistic CFGs

► **Context-Free Grammars (CFG)**

- ❑ formal models for specifying syntax
- ❑ components
 - ❑ terminal symbols
 - ❑ non-terminal symbols
 - ❑ production rules
 - ❑ start symbol

► **Probabilistic CFGs (PCFG)**

- ❑ CFG with a probability value to each production rule

A Simple PCFG

Grammar	Prob
$S \rightarrow NP VP$	0.8
$S \rightarrow Aux NP VP$	0.1
$S \rightarrow VP$	0.1
$NP \rightarrow Pronoun$	0.2
$NP \rightarrow Proper-Noun$	0.2
$NP \rightarrow Det Nominal$	0.6
$Nominal \rightarrow Noun$	0.3
$Nominal \rightarrow Nominal Noun$	0.2
$Nominal \rightarrow Nominal PP$	0.5
$VP \rightarrow Verb$	0.2
$VP \rightarrow Verb NP$	0.5
$VP \rightarrow VP PP$	0.3
$PP \rightarrow Prep NP$	1.0

Lexicon	
$Det \rightarrow the \mid a \mid that \mid this$	0.6 0.2 0.1 0.1 $\Sigma=1.0$
$Noun \rightarrow book \mid flight \mid meal \mid money$	0.1 0.5 0.2 0.2 $\Sigma=1.0$
$Verb \rightarrow book \mid include \mid prefer$	0.5 0.2 0.3 $\Sigma=1.0$
$Pronoun \rightarrow I \mid he \mid she \mid me$	0.5 0.1 0.1 0.3 $\Sigma=1.0$
$Proper-Noun \rightarrow Houston \mid NWA$	0.8 0.2 $\Sigma=1.0$
$Aux \rightarrow does$	1.0 $\Sigma=1.0$
$Prep \rightarrow from \mid to \mid on \mid near \mid through$	0.25 0.25 0.1 0.2 0.2 $\Sigma=1.0$

PCFG Parsing

► Derivation

- sequential application of rules to the start symbol
- *probability*: product of the probabilities of the rules used

► Sequence

- sequence of terminal symbols derived from start symbol
- *probability*: sum of the probabilities of all its derivations

► Sequence parsing

- given a sequence, find a derivation, if one exists
- useful for uncovering the structure of the sequence

► Sequence likelihood

- compute the probability of a derivable sequence
- useful for classifying and ordering sequences

PCFG Benefits

► Representation

- ❑ compact and hierarchical representation of sequences
- ❑ self-explanatory production rules

► Algorithms

- ❑ a variety of parsers for various types of sequences
- ❑ algorithms for learning the probabilities of the rules

► Applications

- ❑ natural language processing
- ❑ visual human activity recognition

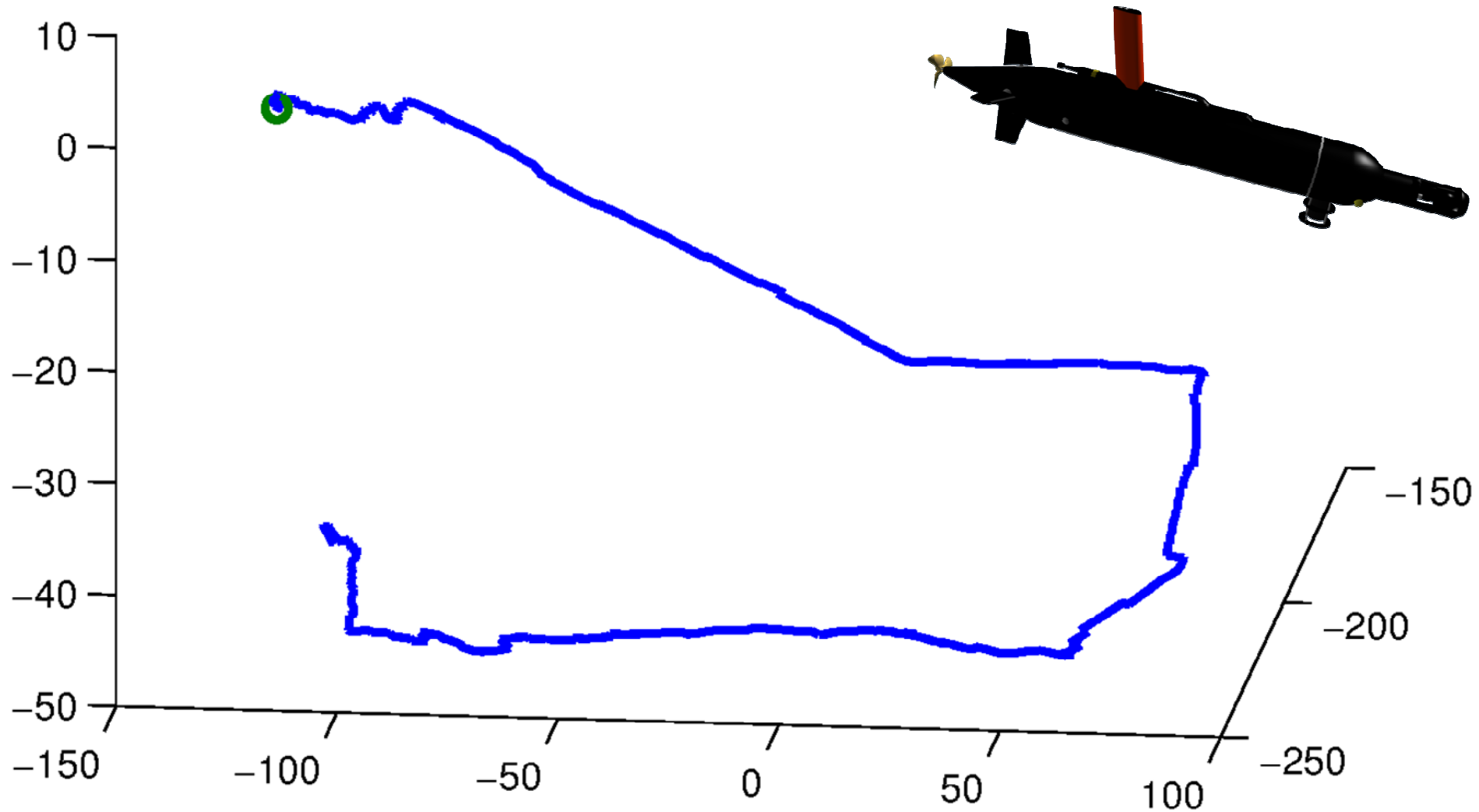
► Noptilus

- ❑ PCFGs for AUV event recognition
- ❑ abstract state description, reduction of communication
- ❑ causal grammar specification, diagnostic sequence parsing

PCFG Event Recognition

A Proof-of-Concept Example

AUV Mission Log



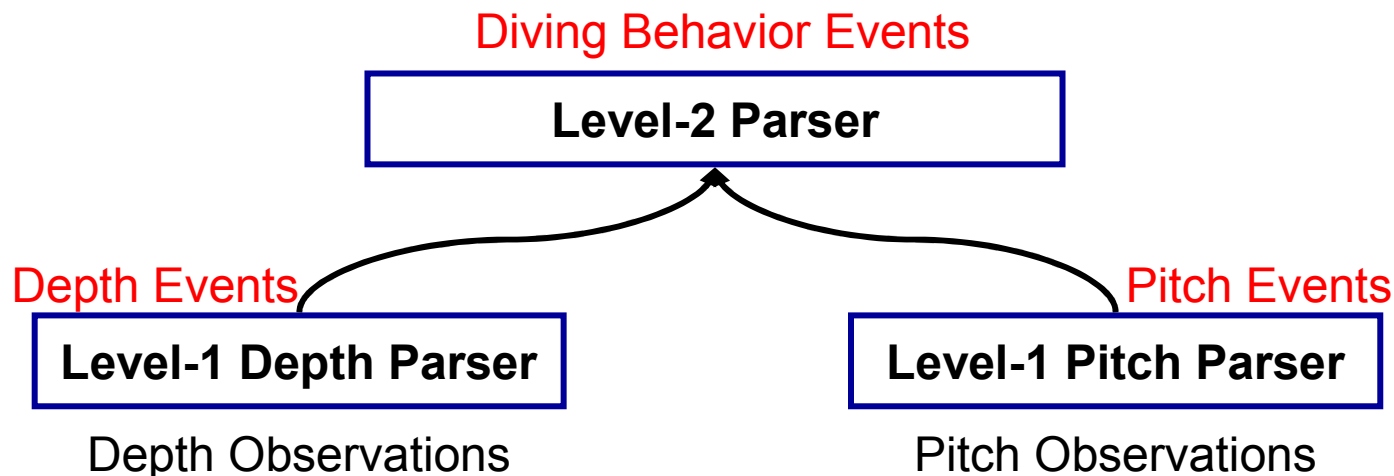
PCFGs for Noptilus

► Goal

- simple event recognition regarding diving behavior
- focus on joint patterns in depth and pitch

► Hierarchy

- level 1: independent grammars for depth and pitch events
- level 2: grammar for the combination of Level-1 events



Generating Level-1 Observations

► Raw data

- ❑ depth measurements
- ❑ pitch measurements

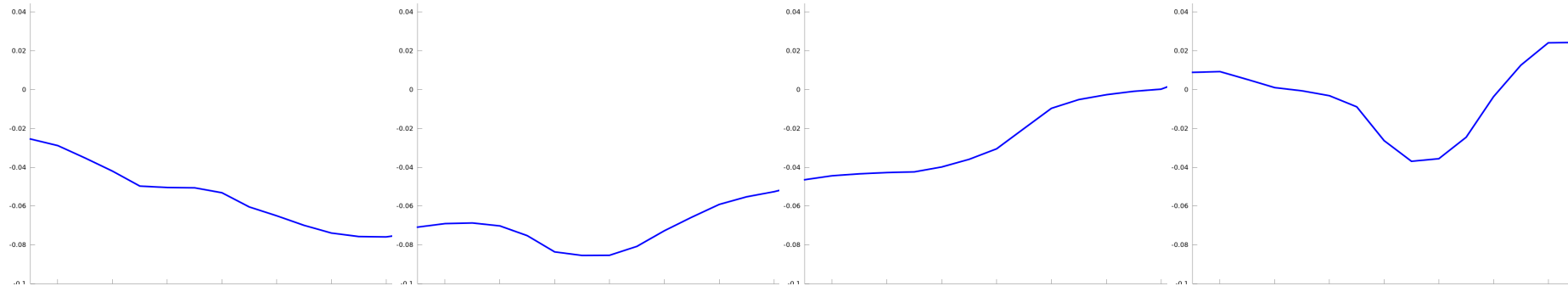
► Abstraction

- ❑ depth: compute the rate of change, average over a window
- ❑ pitch: average over a window

► Level-1 observations

- ❑ $-$: negative ($\text{value} < -\varepsilon$)
- ❑ 0 : zero ($|\text{value}| < \varepsilon$) for some $\varepsilon > 0$
- ❑ $+$: positive ($\text{value} > +\varepsilon$)

Depth Observation Generation



+

0

-

0

0.037474
0.032219
0.028115
0.025039
0.022243
0.022903
0.026767
0.035741
0.037393
0.034305
0.028398
0.023691
0.024289
0.026708
0.027268

0.024209
0.017864
0.014133
0.013886
0.015512
0.016860
0.014031
0.009555
0.003580
-0.000692
-0.002861
-0.007767
-0.012192
-0.017634
-0.019469

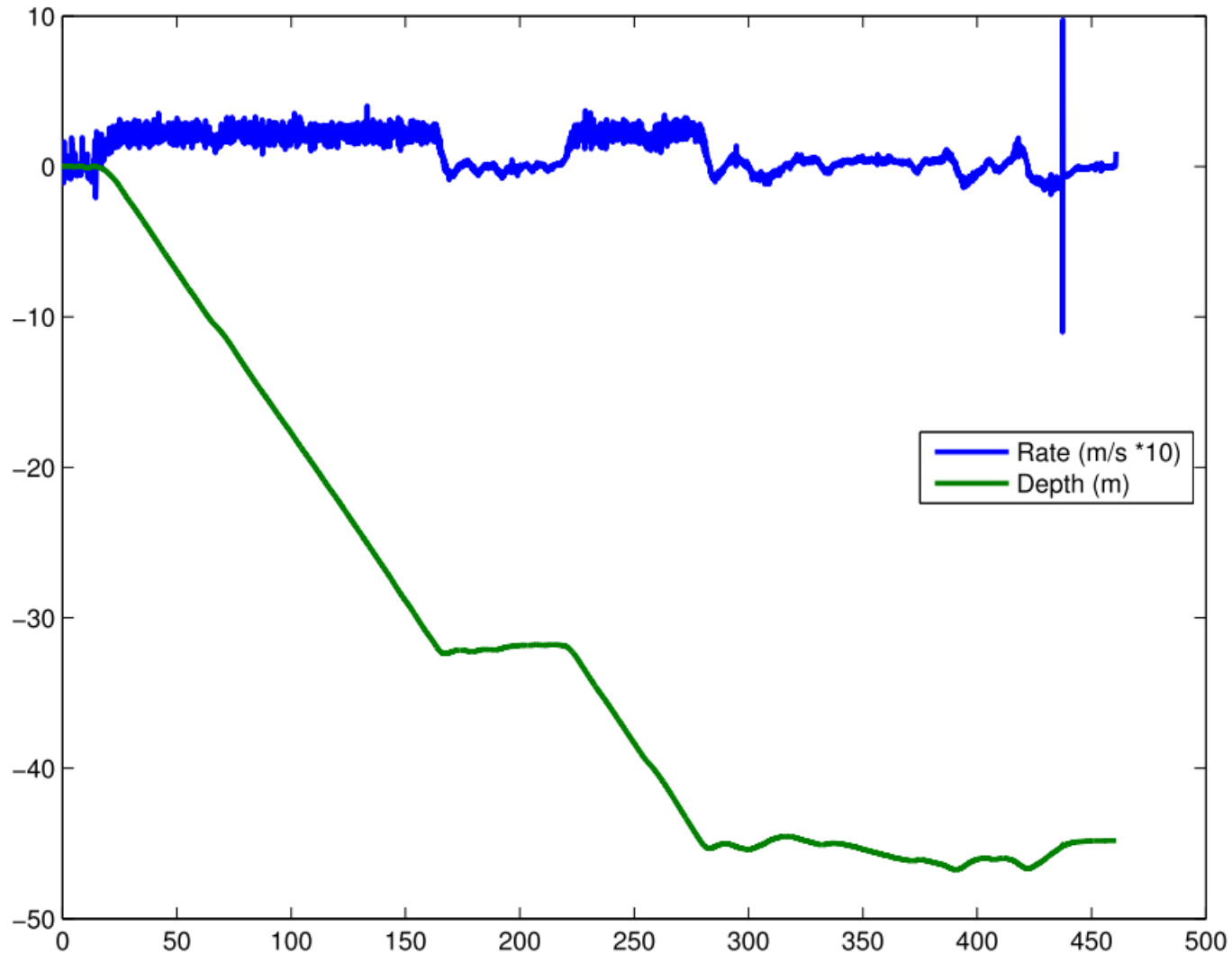
-0.019689
-0.017810
-0.016547
-0.022303
-0.027118
-0.032096
-0.033769
-0.030667
-0.019416
-0.000125
0.008617
0.017316
0.025421
0.028816
0.035222

0.028816
0.035222
0.041991
0.049693
0.050414
0.050569
0.053115
-0.006621
0.007136
0.021382
0.029188
0.040740
0.043735
0.042784
0.042784

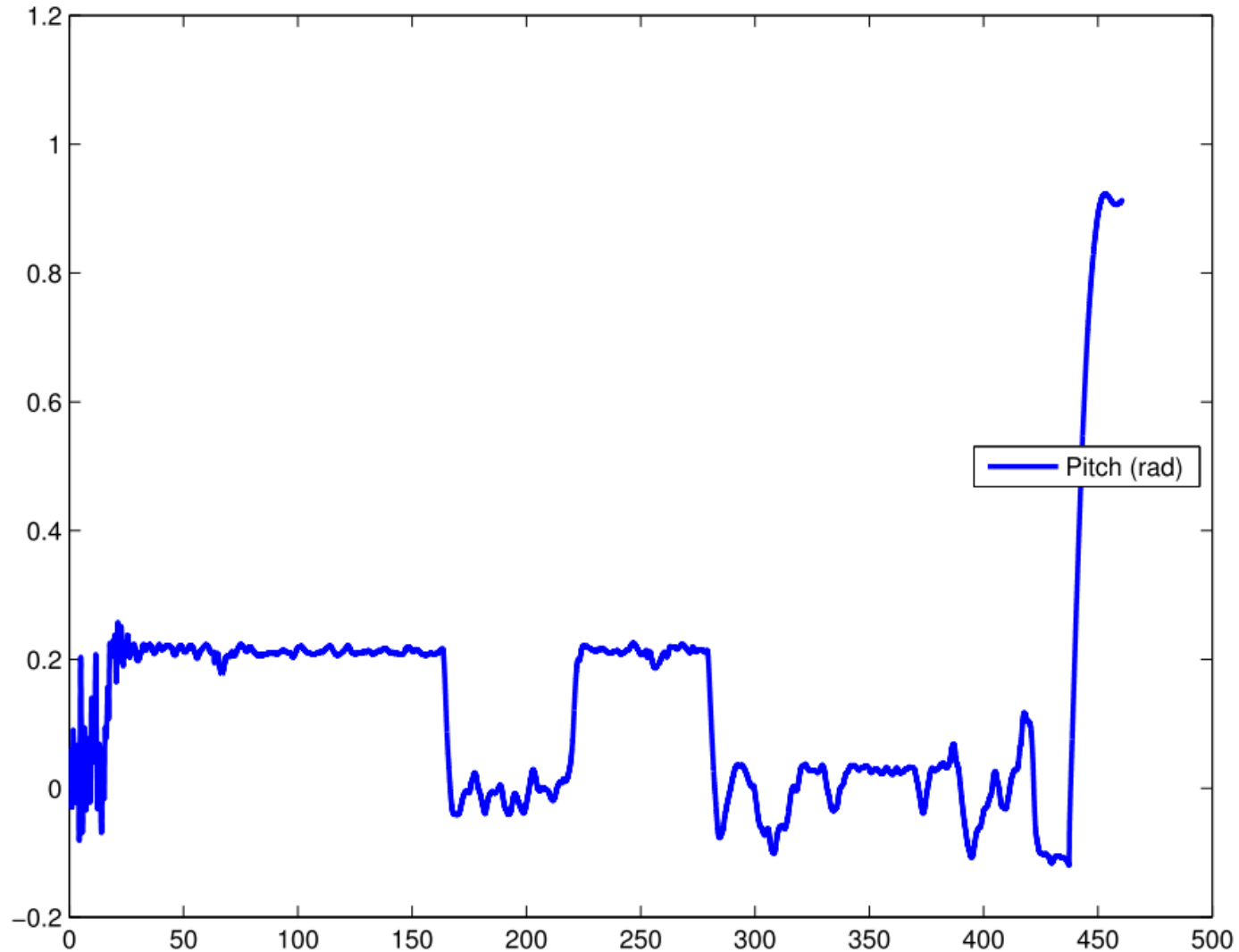
averaging window $s=15$



Depth Sensor Data



Pitch Sensor Data



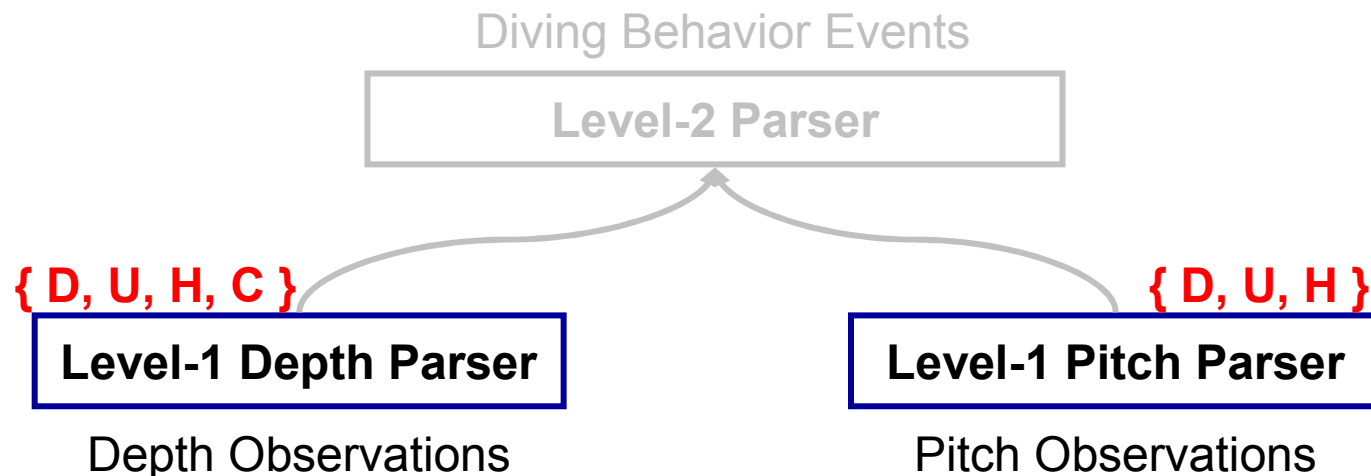
Level-1 Event Recognition

► Level-1 events

- ▣ depth : **D**own, **U**p, **H**over, **C**hange
- ▣ pitch : **D**own, **U**p, **H**over

► Level-1 parsing

- ▣ input: observations over a rolling window
- ▣ output: most probable depth/pitch event occurred



Grammar for Depth Events

$E \rightarrow C$ [0.10]
 $E \rightarrow U$ [0.30]
 $E \rightarrow D$ [0.30]
 $E \rightarrow H$ [0.30]

$d \rightarrow '+'$ [0.85]
 $d \rightarrow '0'$ [0.15]
 $u \rightarrow '-'$ [0.85]
 $u \rightarrow '0'$ [0.15]
 $h \rightarrow '0'$ [1.00]

$D \rightarrow D D$ [0.50]
 $D \rightarrow d$ [0.50]

$U \rightarrow U U$ [0.50]
 $U \rightarrow u$ [0.50]

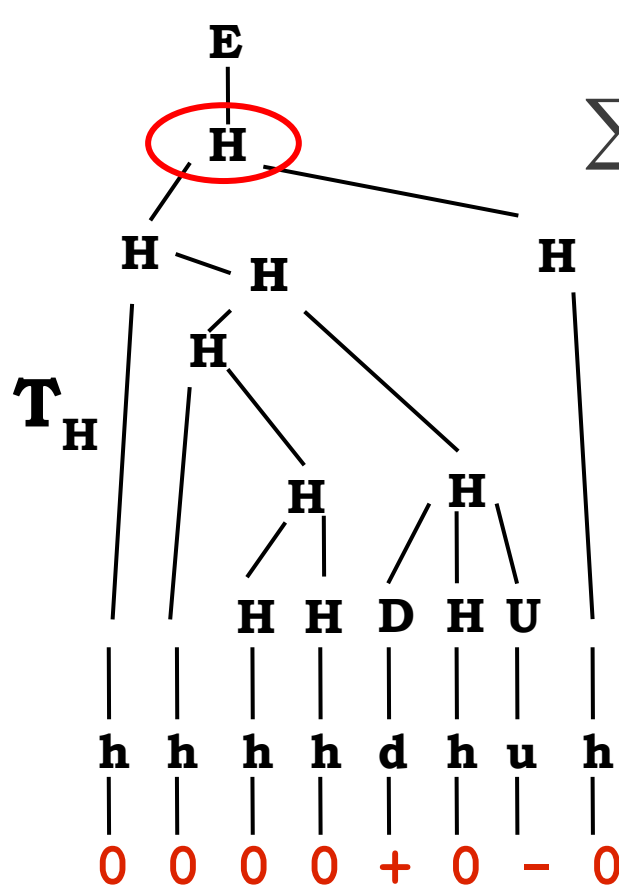
$C \rightarrow U H D$ [0.50]
 $C \rightarrow D H U$ [0.50]

$H \rightarrow H H$ [0.34]
 $H \rightarrow U H D$ [0.16]
 $H \rightarrow D H U$ [0.16]
 $H \rightarrow h$ [0.34]

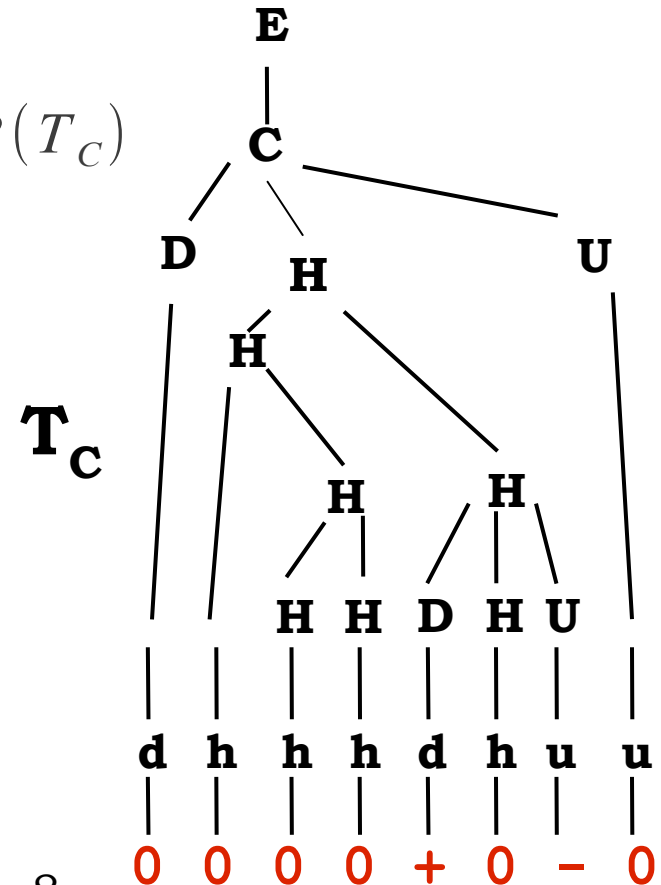
Recognize event E by choosing the derivation that maximizes production probability

Level-1 Parsing: Depth

0 0 0 0 0 0 + 0 - 0 + + + + + + + + + + + + + + +
 H H **H** C C C H C D D D D D D D D D D D D D D D



$$\sum_{T_H} P(T_H) > \sum_{T_C} P(T_C)$$



parsing window w=8

Level-1 Annotated Depth



Level-2 Event Recognition

► Level-2 Observations

- ❑ combined events from Level-1
- ❑ DD, DU, DH, UD, UU, UH, HD, HU, HH, CD, CU, CH

► Level-2 events

- ❑ **OK**, **Collision**, **UpDraft**, **DownDraft**

► Level-2 parsing

- ❑ input: synchronized Level-1 outputs
- ❑ output: most probable diving event

► Patterns

- ❑ the two Level-1 parsers “disagree” for a significant time period

E → OK	[0.889]
E → Co	[0.001]
E → UpD	[0.050]
E → DownD	[0.050]

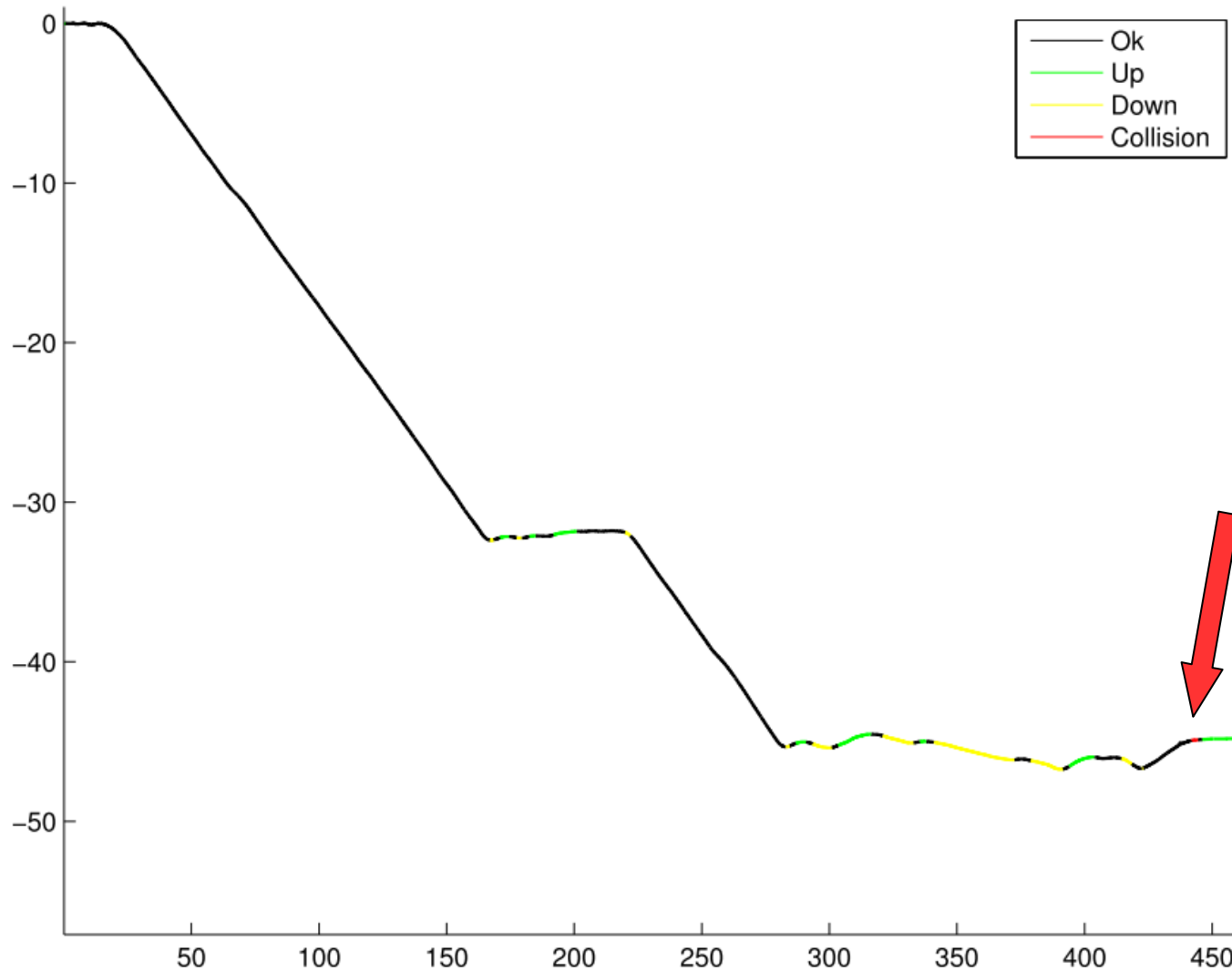
Co → Co co	[0.500]
Co → co	[0.500]

co → o	[0.01]
co → 'UD'	[0.90]

o → 'DD'	[...]
------------------------	--------------

.	.
.	.
.	.
o → 'UU'	[...]

Level-2 Annotated Depth



PCFG Event Recognition

Multiple Experiments

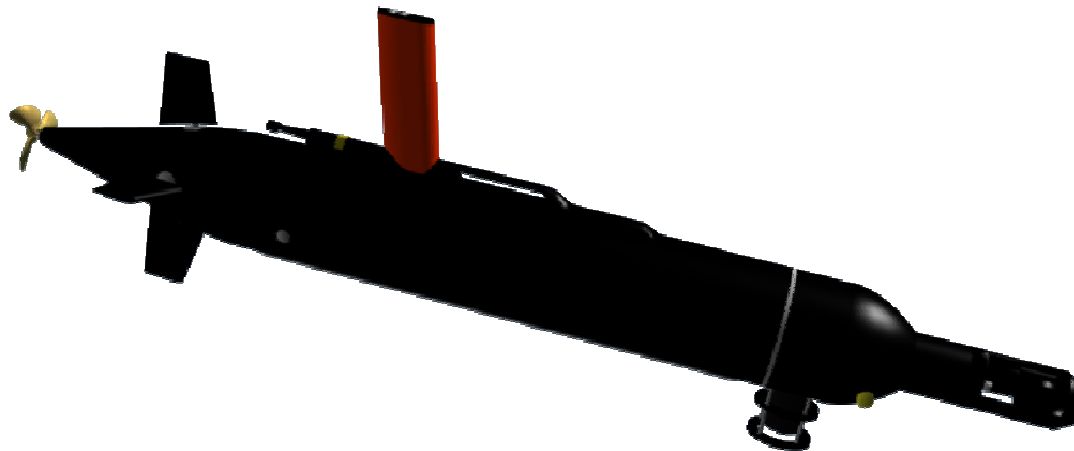
Mission Data

► Data logs

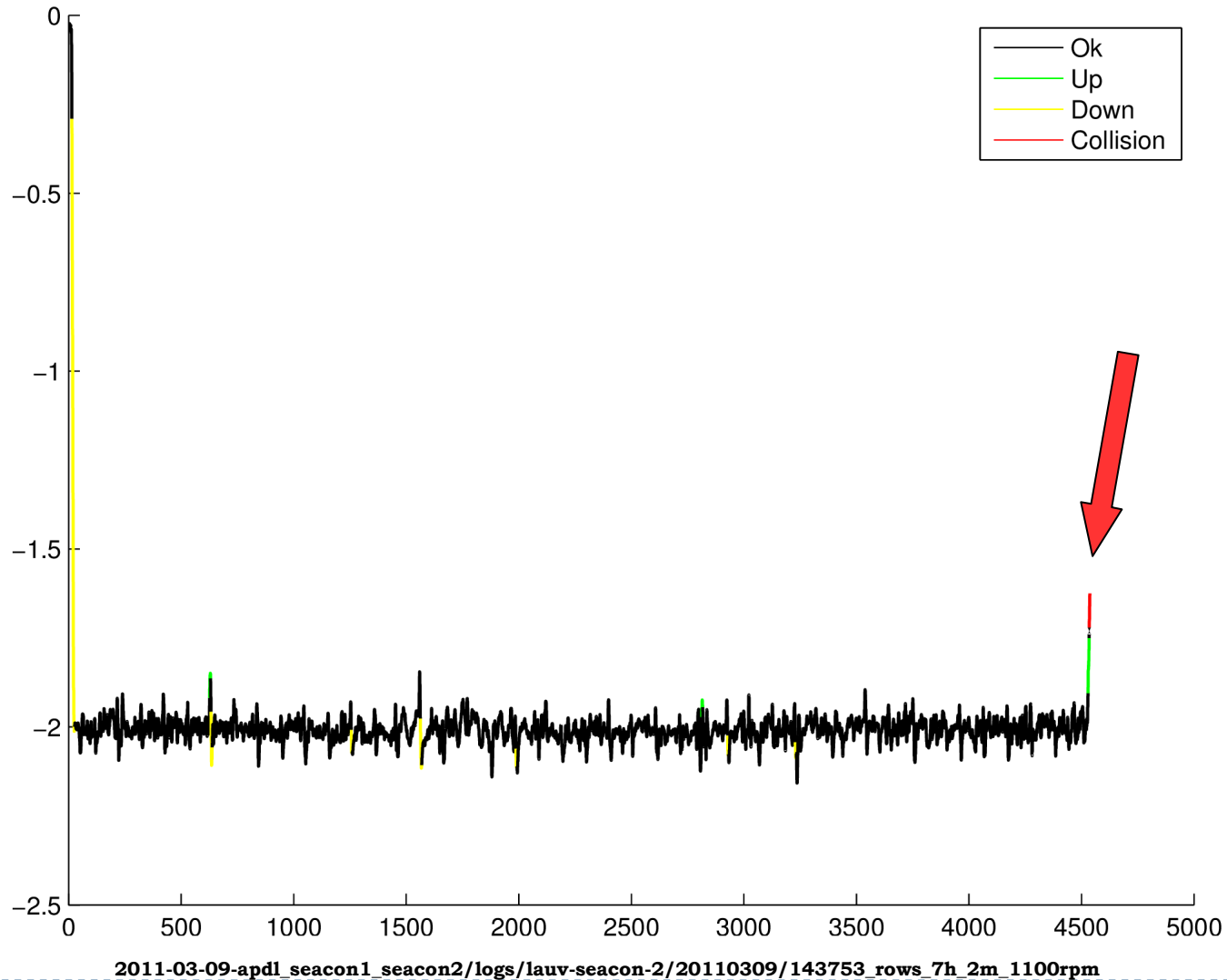
- ❑ seacon-1, seacon-2, swordfish
- ❑ more than 20 hours of mission time

► Human expertise

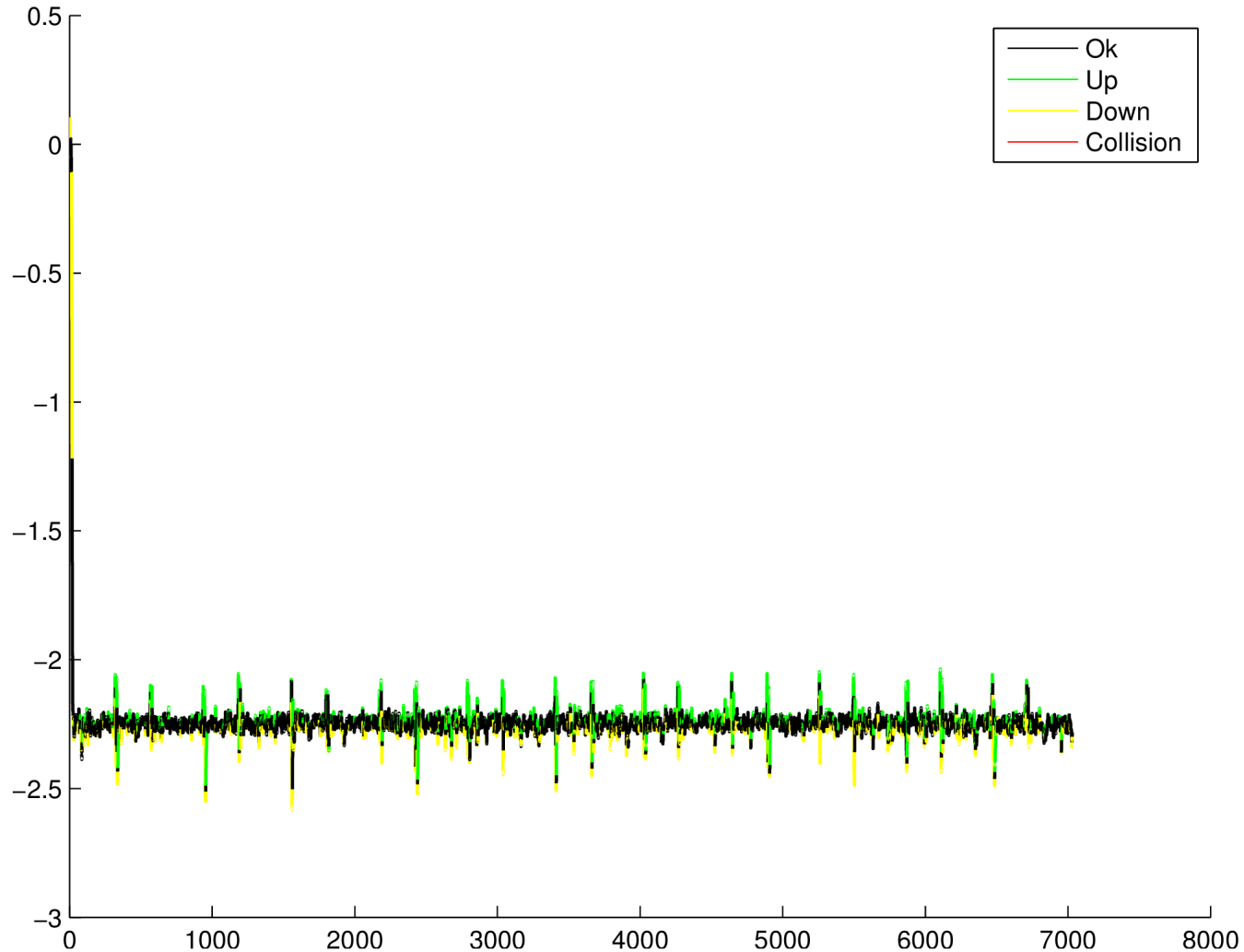
- ❑ descriptive annotation of events
- ❑ related sensor streams



New Mission #1 Depth Annotation

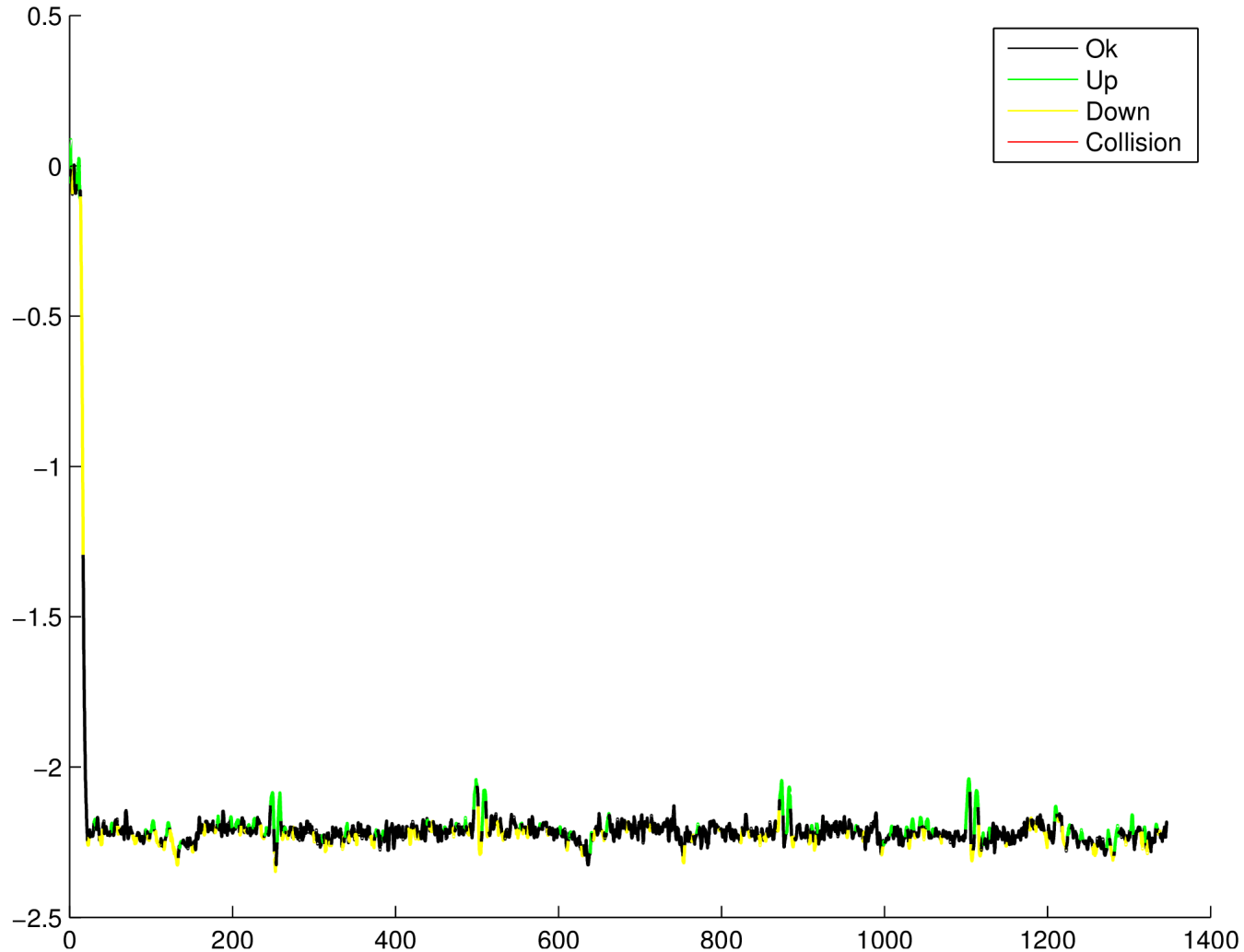


New Mission #2 Depth Annotation



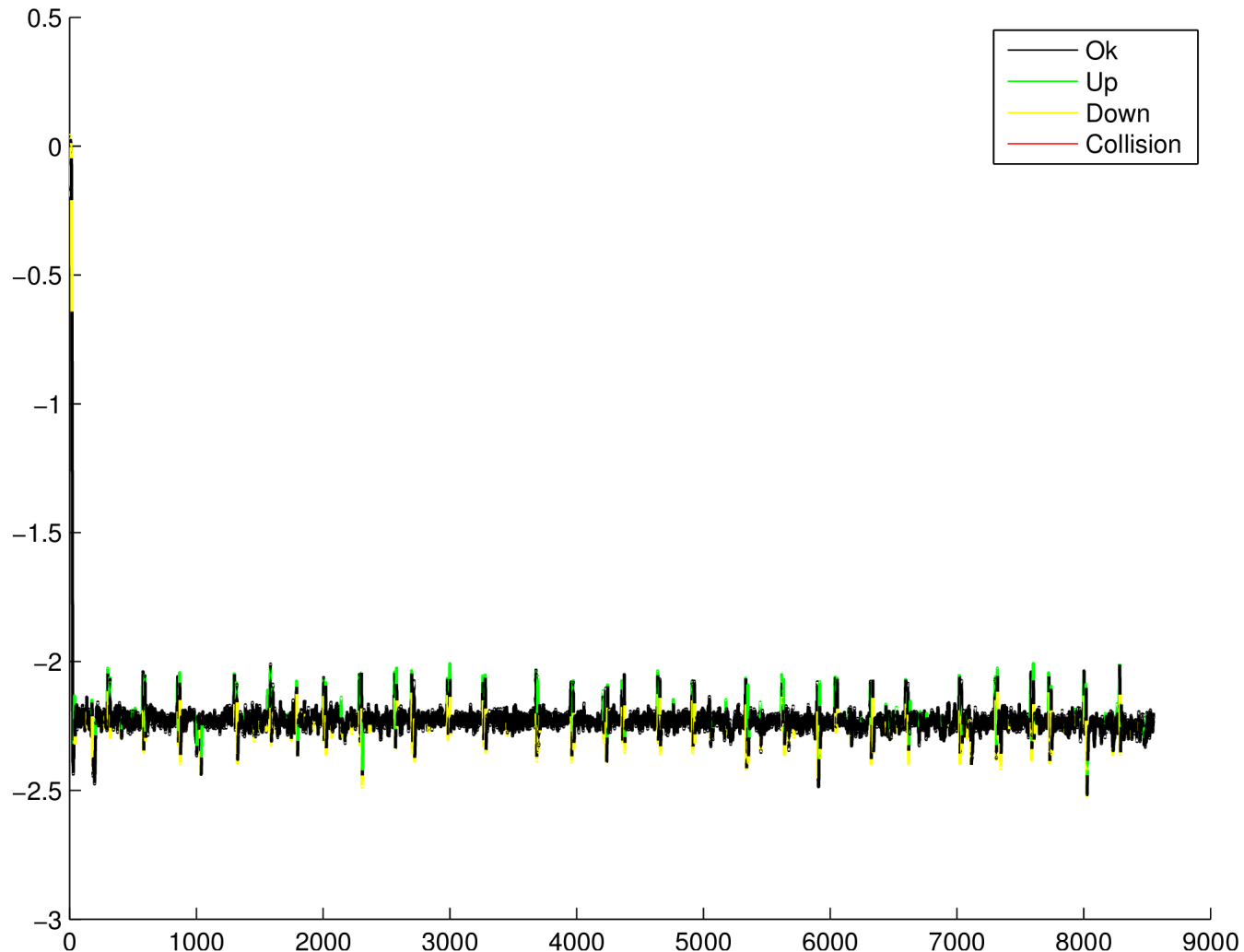
2011-02-08-apdl_season1_season2/logs/lauv-season-1/20110210/141917 rows 2h 2m

New Mission #3 Depth Annotation



2011-02-08-apdl_season1_season2/logs/lauv-season-1/20110209/110213 mission 900rpm 2m 3h30

New Mission #4 Depth Annotation



2011-02-08-apdl seacon1 seacon2/logs/lauv-seacon-1/20110208/141628 long_mission_900rpm 2 30H

Experimental Results

► Grammars

- ❑ minor refinement of the grammars
- ❑ plain tuning of probabilities

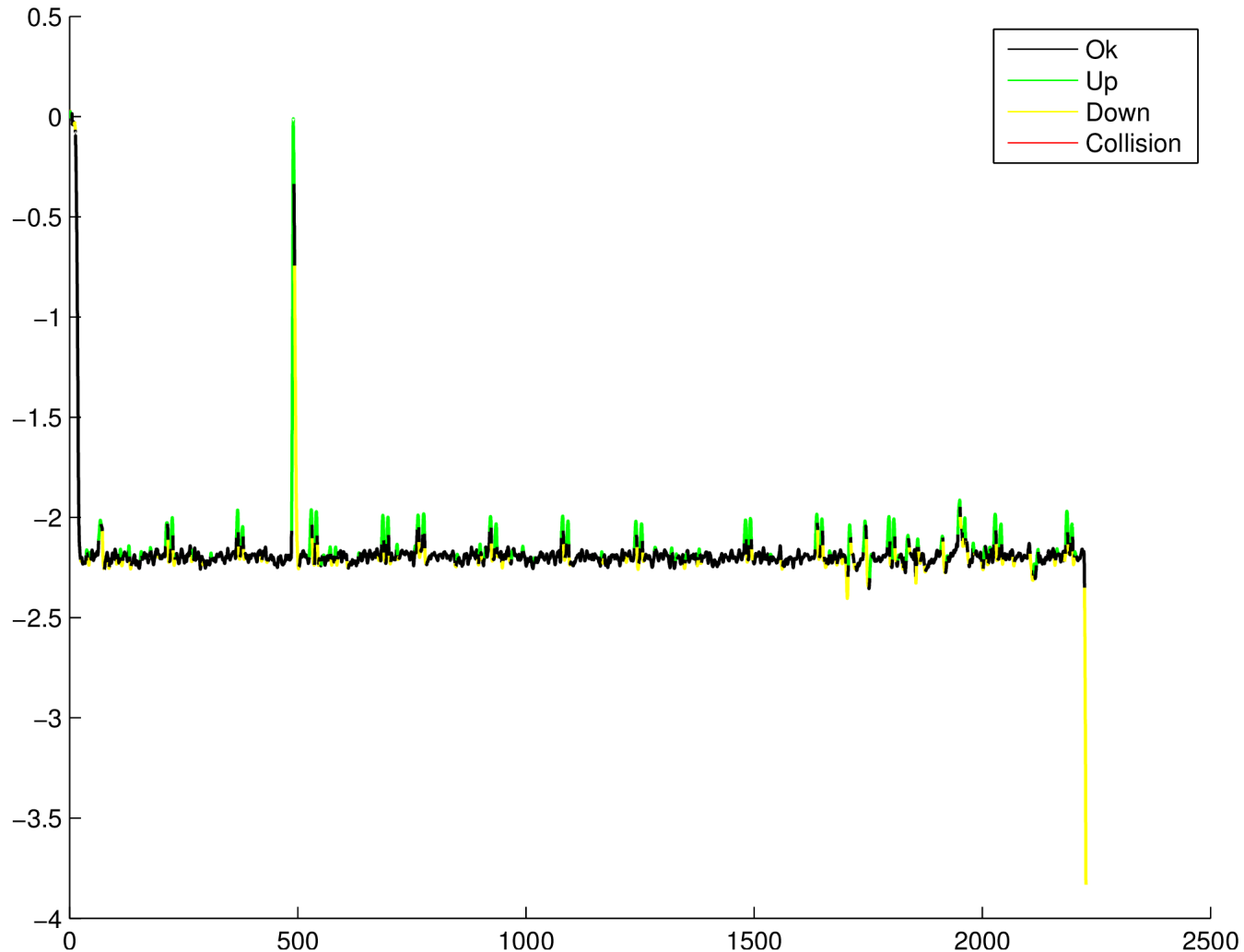
► Event recognition

- ❑ collision is correctly recognized, where present
- ❑ collision is correctly not recognized, where absent
- ❑ initial grammar not an overfit to the initial log!

► Concern

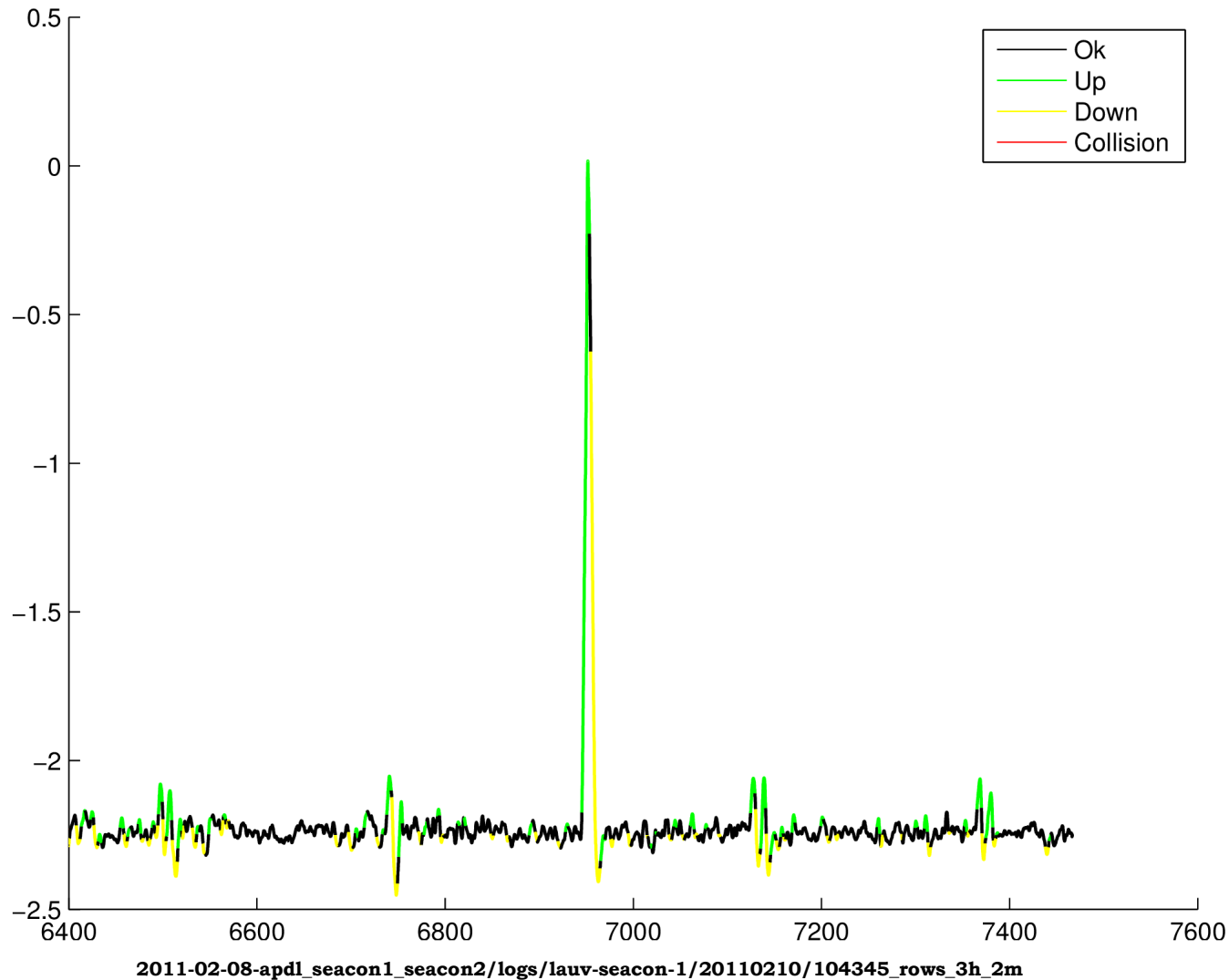
- ❑ lateral collisions were not recognized
- ❑ current grammar unable to detect such events

Unrecognized Lateral Collision #1



2011-02-22-apd1_seacon1_seacon2_nauv/logs/lauv-seacon-1/20110224/103428_rows_4h_2m_1100rpm

Unrecognized Lateral Collision #2



WP6 Progress and Plans

What's there? What's next?

WP6 Progress

► **Preprocessing**

- ❑ abstraction from raw data to observations

► **Event recognition**

- ❑ *events*: diving behavior
- ❑ *technology*: probabilistic context-free grammars
- ❑ successful recognition on several past mission logs
- ❑ demonstration of feasibility and potential

► **Software**

- ❑ optimized parser for event recognition in Matlab

WP6 Plans

► **Software**

- ❑ optimized parser for real-time event recognition in Dune

► **Event recognition**

- ❑ *new events*: getting caught in net, vehicle faults, ...

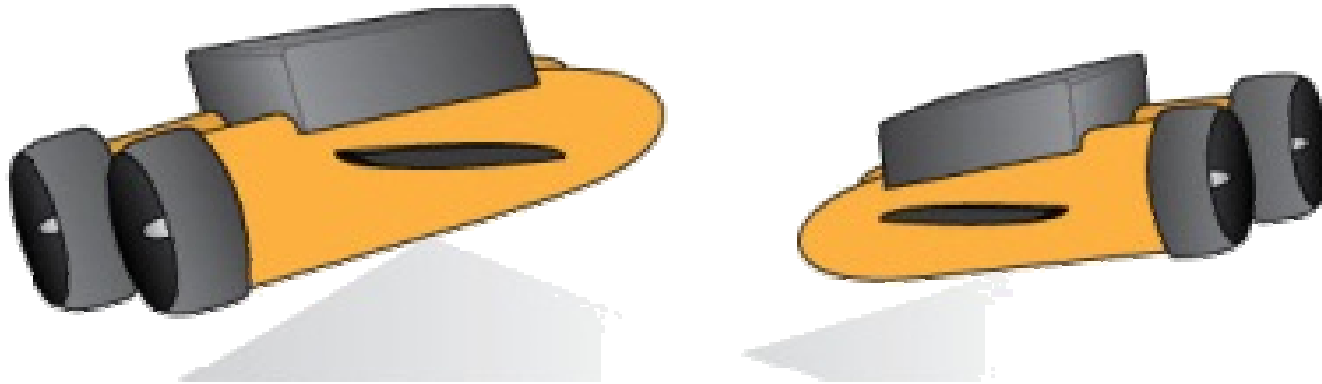
► **Rethinking events**

- ❑ normal vs. abnormal AUV operation
- ❑ *idea*: instead of looking for the abnormal and rare, ...
- ❑ ... why not look after the normal and frequent?

► **Structured prediction**

- ❑ automatic construction of grammars
- ❑ learning rules and probabilities from past missions

Thank you!



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