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A tool that makes the link between Aids to Navigation, traffic volume and the associated risk

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1 Scope

The purpose of this report is to explain a method that can be used to evaluate the effect of the aids to navigation. An aid to navigation (AToN) is any device external to a vessel specifically intended to assist navigators in determining their position or safe course, or to warn them of dangers or obstructions to navigation. The effect of aids to navigation is here quantified by causation factors explained in the report. The method developed is closely linked to the IALA risk tool software IWRAP and the results from the method are a set of causation factors that can be used when creating a waterway model in IWRAP. The report outlines how the method has been derived and it presents a software tool for using the method. The modeling tool used is Bayesian networks.

2 Background

Today most risk models for estimating the grounding or collision frequency are routed in the approach defined by Fujii et al. and by MacDuff. That is, the potential number of ship grounding or ship-ship collisions is first determined as if no aversive maneuvers are made. This potential number of ship accidents is based on 1) an assumed or prespecified geometric distribution of the ship traffic over the waterway and 2) on the assumption that the vessels are navigating blindly as these are operating at the considered waterway. The thus obtained number of potential accident candidates (often called the geometric number of collision candidates) is then multiplied by a specified causation probability to find the actual number of accidents. The causation probability, which acts as a thinning probability on the accident candidates, is estimated conditional on the defined "blind navigation". The method has been refined by Petersen and Friis and implemented by Gatehouse in the software package IWRAP. The software has been thoroughly tested using models from all over the world, and it has been found to give results similar to historical data. The basic idea of the method is shown in Figure 1.

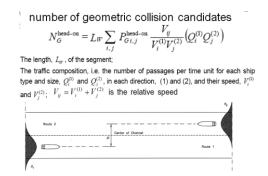


Figure 1: Sketch of the IWRAP method to estimate head-on collision frequency

2.1 Causation factors

One of the cornerstones in the IWRAP model is the use of causation factors. A causation factor is the probability that a person does not act as he or she is supposed to. The causation factor depends on a range of factors from safety culture of the shipping company to the layout of the waterway. Causation factors are in the order of 10⁻⁴, meaning that one out of 10000 times an operation is carried out wrongly. From the waterway authorities point of view it would be valuable to know what the effect of its aids to navigation has on the collision and grounding frequencies. In the figures below some causation factors for different parts of the world are showed. It is important to remember that the smaller a causation factor is the better. A causation factor on 1 means that the navigator always acts incorrectly. 0 means that he always acts correctly.

	Ship-	ship collisions	
Location	Pc	Comment	Reference:
	[×10 ⁻⁴]		see [20] for ref.
Dover Strait	5.18	Head-on, no traffic separation	MacDuff [21]
Dover Strait	3.15	Head-on, with traffic separation	MacDuff [21]
Øresund, Denmark	0.27	Head on	Karlson et al. [19]
Japanese Straits	0.49	Head on	Fujii & Mizuki [9]
Japanese Straits	1.23	Crossings	Fujii & Mizuki [9]
Dover Strait	1.11	Crossings, no traffic separation	MacDuff [21]
Dover Strait	0.95	Crossings, with traffic separation	MacDuff [21]
Strait of Gibraltar	1.2		COWIconsult
Japanese Straits	1.10	Overtaking	Fujii & Mizuki [9]
Great Belt, Denmark	1.30	At bends in lanes	Pedersen et al. [24]
Danish waters	3.0	Head-on and overtaking Crossings also?	COWIconsult Oil and Chemical Spills, 2007

Figure 2: Collision causation factors from the literature.

	Vessel grounding										
Location	Pc	Comment	Reference:								
	[×10 ⁻⁴]		see [20] for ref.								
Japanese Straits	[1.0; 6.3]	Collisions and grounding	Fujii								
Japanese Straits	1.58		Fujii & Mizuki [9]								
Japanese Straits	[0.8; 4.3]		Matsui								
Dover Strait	1.55	No traffic separation	MacDuff [21]								
Dover Strait	1.41	With traffic separation	MacDuff [21]								
Strait of Gibraltar	2.2		COWIconsult								
Øresund, Denmark	2.0		Karlson et al. [19]								

Figure 3: Grounding causation factors from the literature.



3 Method for estimating causation factors

A good and simple method for estimating causation factors is to create a model of the waterway using IWRAP and then relate the number of calculated collisions and grounding to the actual number of historical collisions and grounding. This gives the causation factors. The problem with this method is that it requires the actual number of historical collisions and groundings on each leg and in each crossing point. Because the number of incidents in general is very low, we cannot apply this method leg by leg.

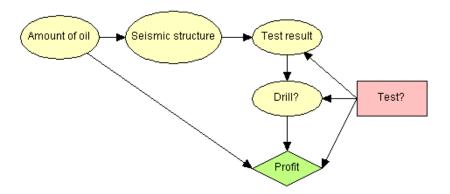
Instead we will try to model what actually goes on, on the bridge of the ship. This is attempted using Bayesian Networks. A Bayesian Network is a graphical representation of uncertain quantities (and decisions) that explicitly reveals the probabilistic dependence between the set of variables and the flow of information in the model. A Bayesian Network is designed as a knowledge representation of the considered problem and may therefore be considered as the proper vehicle to bridge the gap between analysis and formulation.

3.1 Example of a Bayesian network

The network below is a classic example called oil wildcatter. It tells an oil prospector whether or not to make a seismic test before drilling for oil. The ellipse shaped nodes are the probability nodes. The square nodes are decision nodes and the diamond shaped node is a utility node. The arrows indicate the dependence between the nodes.

The 'Test result'-node depends on whether we make a test or not and the actual structure of the underground.

The profit depends on the actual amount of oil in the underground, whether or not we drill and whether or not we make a test (A test cost money)



This Bayesian network can answer questions such as:

- If we do not make a test our expected profit will be \$10
- If we do make a test then our expected profit will be \$22.5

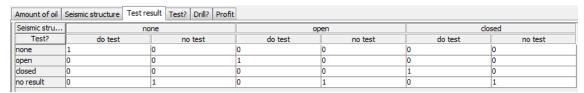
The values inside the nodes look like this:

Amount of oil	Seismic structure	Test result	Test?	Drill?	Profit
dry	0.5				
wet	0.3				
soaking	0.2				

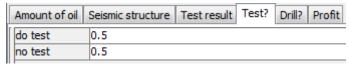
The amount of oil in the ground is only known with a certain probability. For example half the time someone has drilled in the area they have not found anything. 20 % of the time they find oil in abundance.



If the ground is soaking with oil then the ground has a closed structure in 50% of the incidents and there is only 10% chance that it has no structure at all. If the ground is dry then the probability that the structure is closed is only 10%



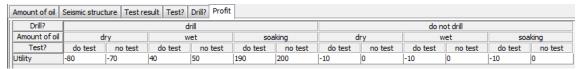
The 'test result' node is a logical node because the values are either true or false. If the seismic structure is open and you do the test then the test result is also an 'open structure'. If you do not do the test then there is of course no result.



The 'Test?' node is a decision node that can be true or false

Test?		do t	test		no test			
Test result	none open		open closed		none	open	closed	no result
drill	0	1	1	0	0.5	0.5	0.5	0.5
do not drill	1	0	0	1	0.5	0.5	0.5	0.5

If we decide to do a test and it shows an open or a closed structure then we drill. If we do not make a test, then we toss a coin whether or not to drill.



The utility node assigns a value to the different outcomes.



3.2 Bayesian network software

A number of software tools for creating Bayesian Networks exist. Here we use the open source tool GeNIe¹ developed at the University of Pittsburgh. This tool is good as long as you do not need to describe the probability tables using equations. GeNIe is only able to use equation nodes if the entire network consists of equation nodes. If equation nodes are required the commercial software package Hugin is a very good option.

3.3 Method

In a number of workshops navigators and risk analysts came up with the network showed in Figure 4. The nodes have been colored so that:

- Orange nodes: Inputs that can be influenced by the maritime authorities.
- Dark orange nodes: Inputs that cannot be influenced by the maritime authorities.
- Yellow nodes: estimated probability tables.
- Green nodes: Final probability tables for the causation factors.
- Dark green nodes: calculation nodes for the causation factors.

By changing the orange input nodes the causation factors can be read in the green nodes.

Overall description of the network

The network calculates the causation factors for a single leg or single junction point. It can be divided into three parts. The left part models how well the ship copes with the situation. The outcome of this side is the node 'Bridge team management'. The right side of the network models the complexity of the network and the outcome of this side is the node 'Waterway complexity'. The two outcome nodes are combined in the causation factor nodes. High waterway complexity increases the causation factors. High 'Bridge team management' decreases the causation factors.



http://genie.sis.pitt.edu/





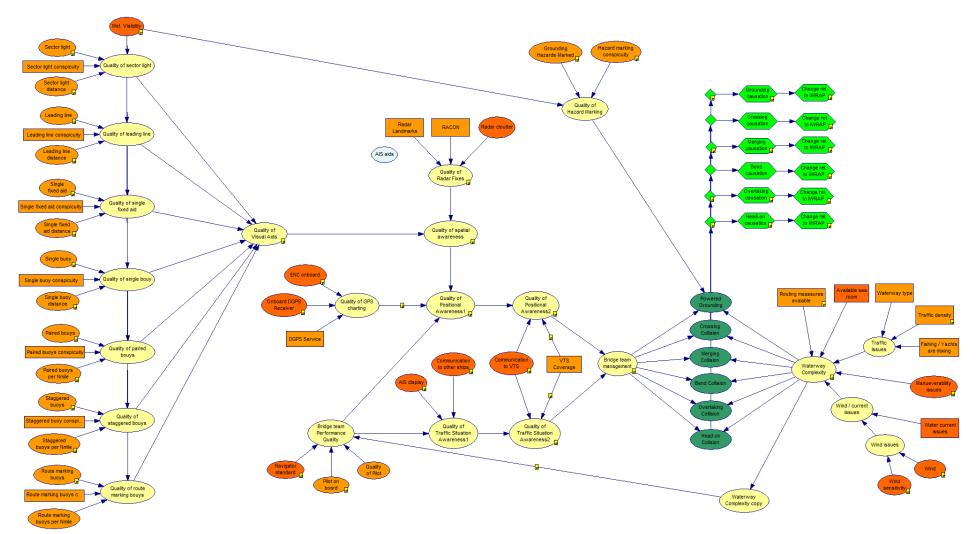


Figure 4: The Bayesian network for estimating the causation factor

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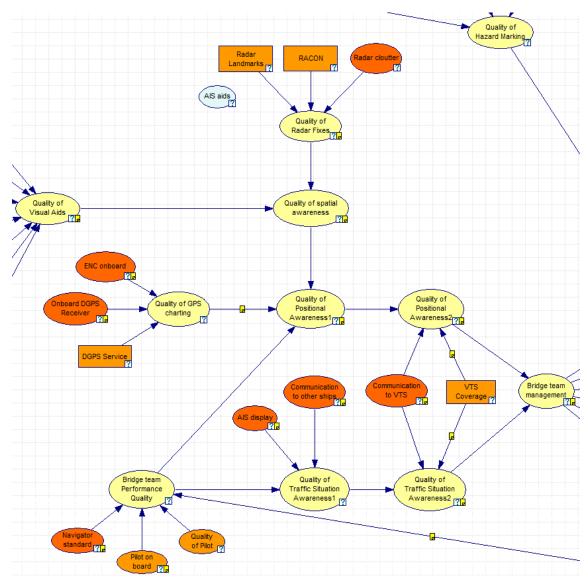


Figure 5: Left side of the network. When the 'Bridge team management' node is in a high state the causation factors are decreased (Good) and vice versa

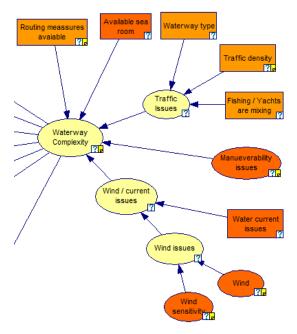


Figure 6: The right side of the network where the water way properties are defined. When the waterway complexity is high the causation factors are increased (bad) and vice versa

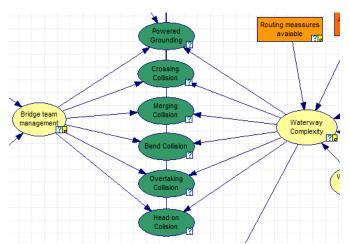


Figure 7: The central part of the network where the two sides are combined and the six causation factors are calculated

3.4 Verifying the network

The network and the results from it, has only been verified as to their overall value. The actual values which are valid for a single leg are quite difficult to verify. A workshop in which several persons do a number of cases is required to verify it.

4 Implementing the network

In order to test the network for people not familiar with GeNIe, a Windows application has been developed. This allows the user to play with the input nodes and see how the causation factors are affected.

The program is started by executing the file Effect of AtoNs.exe. Each tab in Figure 8 represents a part of the network's input nodes. After giving the input, return to the Menu-tab and press the Calculate-button. Now the causation factors are shown. The case can be saved by pressing the save-button. The grid at the bottom shows The Iterate-button makes it possible to choose one or more input nodes. The program then automatically assign values to the nodes and calculate the causation factors. The result is saved in a csv-file.

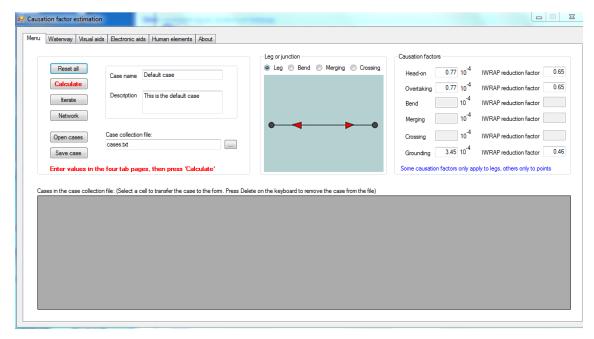
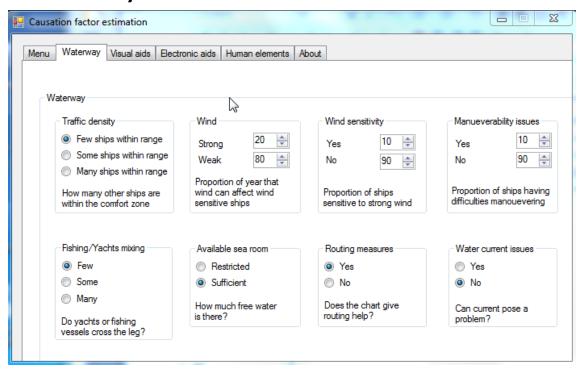


Figure 8: User interface for testing the network

5 Detailed description of the input parameters

In the following each input node/parameter is described. The node/parameter name is underlined.

5.1 Waterway tab



Traffic density

Few ships within range: So few ships comes within the Baltic, Kattegat

comport zone that the navigators

might relax too much.

Some ships within range: 1-3 ship comes within the ships Great Belt

comport zone every hour

Many ships within range: Almost constantly others ships Dover, Malacca,

within the comport zone Oresund, Bosperus

Wind

Proportion of the year that wind may be a problem for wind sensitive ships.

Wind sensitivity

Proportion of ships that are sensitive to strong wind. Typical ships with a high super structure.

Maneuverability issues

Proportion of ships that have difficulties in deviating from their course.



Fishing/Yachts mixing

Few: Almost never an issue Kattegat

Some: Can be an issue Great Belt

Many: Constantly an issue Oresund

Available sea room

Restricted: Ships will not be able to make a turn

Routing measures

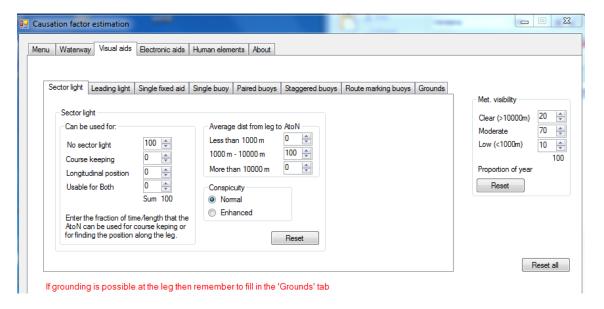
Does the chart offer help in navigating the waterway?

Water current issues

Is it necessary for the ships to account for current in the area?

5.2 Visual aids

The idea here is that visual aids can be used for course keeping or for fixing the position if sailing along a row of buoys.

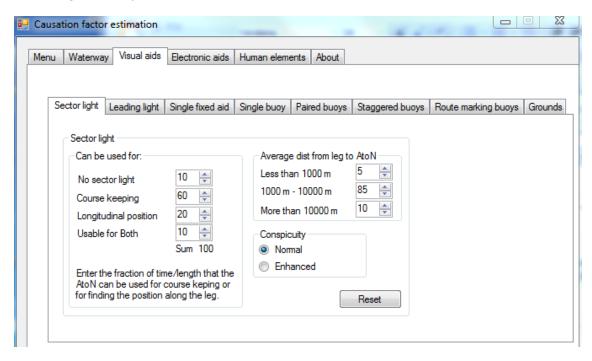


Met. visibility

The proportion of the year that the meteorological visibility in the area is more than $10\ km$, $1\text{-}10\ km$, less than 1km.



Sector light example



Can be used for

In the example above the sector light cannot be used/seen for 10 % of the leg/point. During 60 % length of the leg it can only be used for course keeping. During 20 % of the leg it can only be used for fixing the position. In 10 % of the length of the leg it can be used for both. This is a constructed example. In most cases one of the text boxes can be assigned 100.

Average distance

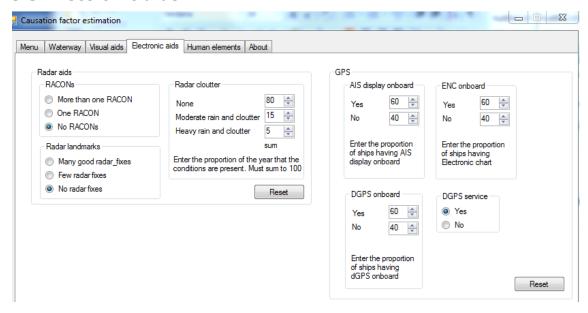
During 5 % of the length of the leg, the ship is less than 1000 m from the sector light. During 85 % of the length of the leg the ship is between 1000 and 10000 m from the sector light. In 10 % of the length of the leg the ship is more than 10000 m. from the sector light.

Conspicuity

Conspicuity is here set to Normal. If something extra has been done to the sector light it can be set to Enhanced.



5.3 Electronic aids



RACONS's

The number of RACONs that can be used at this leg/point.

Radar landmarks

This indicates how easy the coast line is seen on the radar screen.

Radar cloutter

This indicates how often the atmospheric conditions can yam the radar.

AIS display onboard

On how many ships are the AIS signals showed graphically on a monitor?

ENC onboard

How many ships have electronic charting onboard?

DGPS onboard

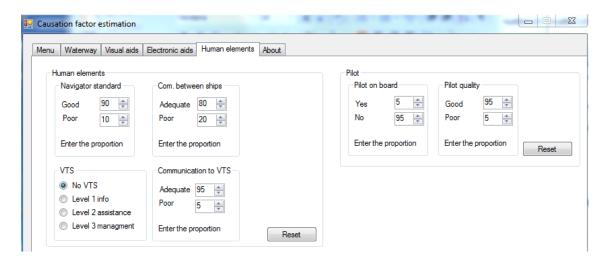
How many ships are able to use a DGPS service?

DGPS service

Is there a Differential Global Positioning System service in the area. DGPS can improve the precision from 20 m to 1 m.



5.4 Human elements



Navigator standard

The proportion of navigators that cannot navigate safely.

Communication between ships

The proportion of ships that is able to understand each other.

Communication to VTS

The proportion of ships that can communicate with the VTS.

VTS

Is there a VTS service in the area?

Pilot

How many ships have a pilot onboard

Pilot quality

How qualified are the pilots in the area?



6 Example. Drogden channel in the Sound, Denmark

In the chart below we will calculate the causation factors for the leg (blue) going through the Drogden channel. On the following pages the input parameters are shown. The results are that the collision causation factors are increased by 8 % compared to the IWRAP default causation factors. The grounding causation factor is increased by 56 %. When comparing these factors to the causation factors in table, we see that we get within 50% of the observed values. It is important to note that we should not expect to get the same values as the values in table x is an average value of the area, where the calculated values are specific for a single leg.



Figure 9: Example leg through Drogden channel

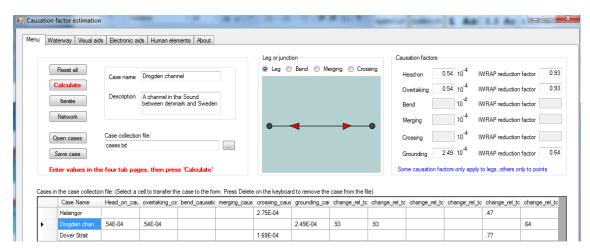
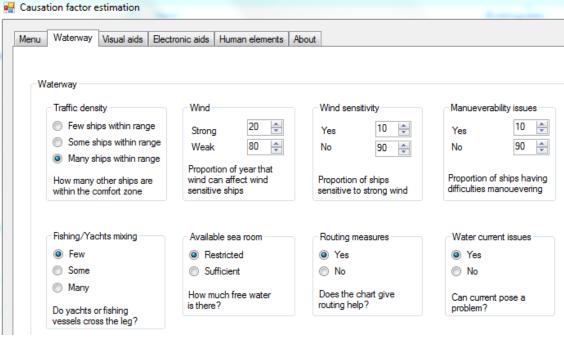


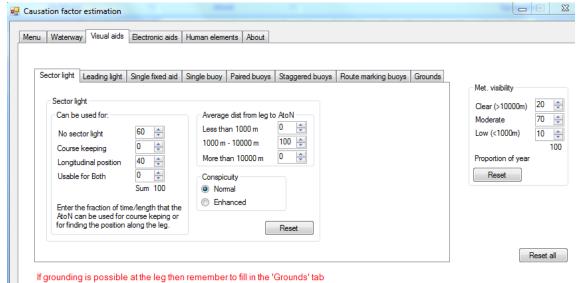
Figure 10: example results. The factors fit well with Figure 2 and Figure 3



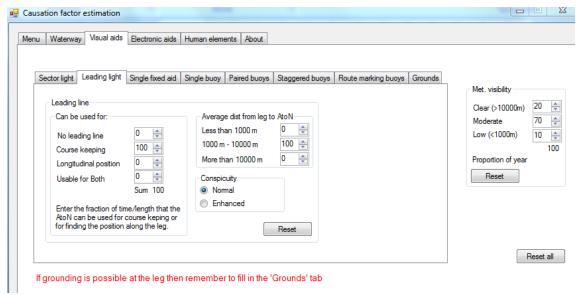
Waterway



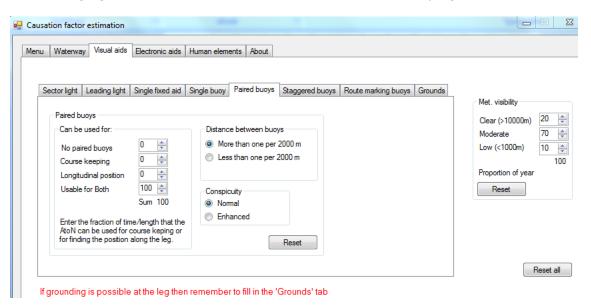
Visual aids



Sector light can be used 40% of the leg for positioning

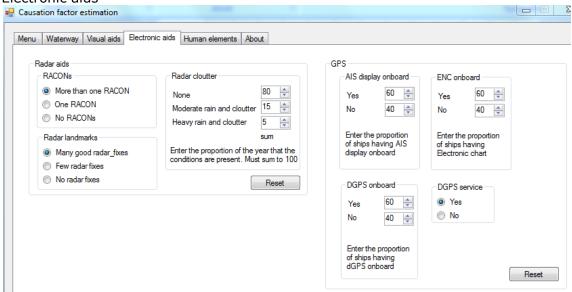


The leading light can be used 100% of the time for course keeping

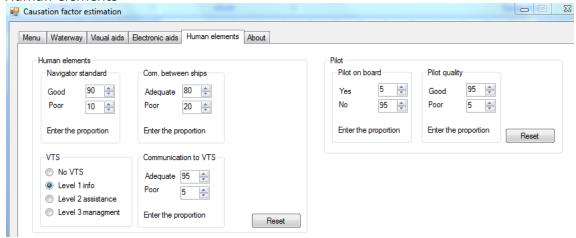


Paired buoys can be used 100 % of the leg

Electronic aids



Human elements



7 Detailed description of the network

The input nodes of the network has already been described in section x. In this section the internal nodes are describes. Many nodes have the states high, adequate and low. These are of course not very precise descriptions. In general low means that the situation is not acceptable. Adequate means that the situation is acceptable and high means that the situation is above what is required.

Bridge team management

This models the combined effectiveness of the people on the bridge

Quality of Positional Awareness2		High			Enough		□ Inadequate			
Quality of Traffic Situation Awareness2	High	Enough	Inadequate	High	Enough	Inadequate	High	Enough	Inadequate	
▶ High	1	0.5	0	0.5	0	0	0	0	0	
Enough	0	0.5	0.5	0.5	1	0.3	0.5	0.3	0	
Inadequate	0	0	0.5	0	0	0.7	0.5	0.7	1	

Quality of positional awareness2

This node captures the effect of VTS. Even though the positional awareness1 is low, then a VTS can correct this.

	VTS Coverage			No	VTS			□ Level1_info					
Qu	uality of Positional Awareness 1	□	ligh	□ Enc	ough	□ Inade	equate	□ ⊦	ligh	□ Enc	ough	□ Inade	equate
	Communication to VTS	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor
F	High	1	1	0	0	0	0	1	0.9	0.5	0	0	0
	Enough	0	0	1	1	0	0	0	0	0.5	0.8	0.5	0.1
	Inadequate	0	0	0	0	1	1	0	0.1	0	0.2	0.5	0.9

VTS Coverage	Ξ		Level2_a	ssistance			□ Level3_management					
Quality of Positional Awareness 1	□ Hi	gh	□ End	ugh	□ Inade	quate	□	ligh	□ Enc	ough	□ Inade	quate
Communication to VTS	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor
▶ High	1	0.9	0.75	0	0	0	1	0.9	0	0	0	0
Enough	0	0	0.25	0.75	0.75	0.1	0	0	1	0.5	1	0.1
Inadequate	0	0.1	0	0.25	0.25	0.9	0	0.1	0	0.5	0	0.9

Quality of positional awareness1

This node gives the probability that the bridge team knows where they are.

Bridge team Performance Quality		High										
Quality of spatial awareness	Position_c	an_be_establis	hed_easily	☐ Position_ca	nnot_be_estab	olished_e	─ Position_cannot_be_established					
Quality of GPS charting	ENC_high	ENC_low	Position_se	ENC_high	ENC_low	Position_se	ENC_high	ENC_low	Position_se			
▶ High	1	0.95	0.9	1	0.9	0	1	0.85	0			
Enough	0	0.05	0.1	0	0.1	0.8	0	0.15	0.6			
Inadequate	0	0	0	0	0	0.2	0	0	0.4			

E	Bridge team Performance Quality		Average									
Г	Quality of spatial awareness	─ Position_c	an_be_establis	hed_easily	─ Position_ca	nnot_be_estab	olished_e	Position_	cannot_be_es	tablished		
Г	Quality of GPS charting	ENC_high	ENC_low	Position_se	ENC_high	ENC_low	Position_se	ENC_high	ENC_low	Position_se		
▶	High	0.9	0.85	0.8	0.85	0.8	0	0.8	0.75	0		
Г	Enough	0.1	0.15	0.2	0.15	0.2	0.7	0.2	0.25	0.5		
Г	Inadequate	0	0	0	0	0	0.3	0	0	0.5		

В	ridge team Performance Quality	Ξ				Below_average	•			
Г	Quality of spatial awareness	─ Position_c	an_be_establis	hed_easily	─ Position_ca	nnot_be_estal	olished_e	Position_	_cannot_be_es	tablished
	Quality of GPS charting	ENC_high	ENC_low	Position_se	ENC_high	ENC_low	Position_se	ENC_high	ENC_low	Position_se
┢	High	0.8	0.75	0.7	0.75	0.7	0.2	0.75	0.7	0
	Enough	0.2	0.25	0.2	0.25	0.3	0.5	0.25	0.2	0.5
Г	Inadequate	0	0	0.1	0	0	0.3	0	0.1	0.5

Quality of Traffic Situation Awareness2

This node captures the effect of VTS. Even though the traffic situation awareness1 is low, then a VTS can correct this.

VTS Coverage	Ξ		No	VTS			□ Level1_info					
Quality of Traffic Situation Awareness 1	⊟ H	igh	Enc	ugh	□ Inade	equate	⊟ Hi	gh	□ Eno	ugh	□ Inade	quate
Communication to VTS	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor
▶ High	1	1	0	0	0	0	1	0.9	0.5	0	0	0
Enough	0	0	1	1	0	0	0	0	0.5	0.5	1	0.1
Inadequate	0	0	0	0	1	1	0	0.1	0	0.5	0	0.9

VTS Coverage	Ξ		Level2_a	ssistance			=		Level3_ma	nagement		
Quality of Traffic Situation Awareness1	⊟ Hi	gh	□ Eno	ugh	□ Inade	quate	□ H	igh	Eno	ugh	□ Inade	quate
Communication to VTS	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor	Adequate	Poor
▶ High	1	0.9	0.75	0	0	0	0.9	0.8	1	0	1	0
Enough	0	0	0.25	0.75	0.75	0.1	0	0.2	0	0.8	0	0.1
Inadequate	0	0.1	0	0.25	0.25	0.9	0.1	0	0	0.2	0	0.9

Quality of Traffic Situation Awareness1

The probability that the bridge team understand the traffic around them

AIS display	⊟		Y	es		
Communication to other ships		Adequate			Poor	
Bridge team Performance Quality	High	Average	Below_ave	High	Average	Below_ave
▶ High	1	0.2	0	0.7	0.1	0
Enough	0	0.8	0.4	0.2	0.7	0.2
Inadequate	0	0	0.6	0.1	0.2	0.8

AIS display		No								
Communication to other ships		Adequate			Poor					
Bridge team Performance Quality	High	Average	Below_ave	High	Average	Below_ave				
▶ High	0.9	0	0	0.5	0	0				
Enough	0.1	1	0.2	0.3	0.7	0				
Inadequate	0	0	0.8	0.2	0.3	1				

Bridge team performance quality

Quality of Pilot	⊟					Go	ood					
Navigator standard	⊟		Go	od			□ Poor					
Pilot on board	Ξ	Yes			No			Yes			No	
Waterway Complexity	High	Nomal	Low	High	Nomal	Low	High	Nomal	Low	High	Nomal	Low
▶ High	0.7	0.8	0.6	0.4	0.5	0.7	0.64	0.59	0.49	0	0	0
Average	0.3	0.2	0.4	0.4	0.5	0.25	0.35	0.4	0.5	0.1	0.2	0.5
Below_average	0	0	0	0.2	0	0.05	0.01	0.01	0.01	0.9	0.8	0.5

Quality of Pilot	⊟					P	oor					
Navigator standard			Go	od			□ Poor					
Pilot on board		Yes			No		Ξ	Yes			No	
Waterway Complexity	High	Nomal	Low	High	Nomal	Low	High	Nomal	Low	High	Nomal	Low
▶ High	0.3	0.4	0.6	0.4	0.5	0.7	0	0	0	0	0	0
Average	0.4	0.4	0.3	0.4	0.5	0.25	0.05	0.1	0.4	0.1	0.2	0.5
Below_average	0.3	0.2	0.1	0.2	0	0.05	0.95	0.9	0.6	0.9	0.8	0.5

Quality of GPS charting

	ENC onboard	Ξ		Y	es			N	lo .	
	Onboard DGPS Receiver	Ξ	Ye	s		lo	Y	es	□ N	lo
	DGPS Service	Yes		No	Yes	No	Yes	No	Yes	No
▶	ENC_high_pos_acc		1	0	0	0	0	0	0	0
Г	ENC_low_pos_acc		0	1	1	1	0	0	0	0
Г	Position_set_in_paperchart		0	0	0	0	1	1	1	1

Quality of spatial awareness

Quality of Visual Aids		Position_can_be_established_easily								
Quality of Radar Fixes	Position_can_be_established_easily	Position_cannot_be_established_easily	Position_cannot_be_stablished_by_radar							
▶ Position_can_be_established_easily	1	1	1							
Position_cannot_be_established_easily	0	0	0							
Position_cannot_be_established	0	0	0							

Quality of Visual Aids	Position_cannot_be_established_easily								
Quality of Radar Fixes	Position_can_be_established_easily	Position_cannot_be_established_easily	Position_cannot_be_stablished_by_radar						
► Position_can_be_established_easily	0.5	0	0						
Position_cannot_be_established_easily	0.5	0.8	1						
Position_cannot_be_established	0	0.2	0						

Quality of Visual Aids	E	Position_cannot_be_established_visua	ally
Quality of Radar Fixes	Position_can_be_established_easily	Position_cannot_be_established_easily	Position_cannot_be_stablished_by_radar
▶ Position_can_be_established_easily	0.3	0	0
Position_cannot_be_established_easily	0.6	0.5	0
Position_cannot_be_established	0.1	0.5	1

Quality of radar fixes

	RACON	⊟			Mo	re_than_one				
	Radar Landmarks	─ Many_good	l_radar_fixes_in_:	several_dire	☐ Few_good	_radar_fixes_on	e_direction		No_radar_fixes	
	Radar cloutter	None	Moderate_rain	Heavy_rain	None	Moderate_r	Heavy_rain	None	Moderate_r	Heavy_rain
Þ	Position_can_be_established_easily	1	0.8	0.5	1	0.8	0.5	1	0.8	0.4
	Position_cannot_be_established_easily	0	0.2	0.4	0	0.2	0.4	0	0.2	0.5
Г	Position_cannot_be_stablished_by_radar	0	0	0.1	0	0	0.1	0	0	0.1

Г	RACON		One									
	Radar Landmarks	─ Many_good	d_radar_fixes_ir	n_several	☐ Few_good.	_radar_fixes_on	e_direction		No_radar_fixes	+		
Г	Radar cloutter	None	Moderate_r	Heavy_rain	None	Moderate_r	Heavy_rain	None	Moderate_r	Heavy_rain		
П	Position_can_be_established_easily	1	0.8	0.5	1	0.8	0.5	0	0	0		
Г	Position_cannot_be_established_easily	0	0.2	0.4	0	0.2	0.4	1	0.9	0.6		
	Position_cannot_be_stablished_by_radar	0	0	0.1	0	0	0.1	0	0.1	0.4		

	RACON	⊟				None				
	Radar Landmarks	─ Many_good	l_radar_fixes_ir	n_several	☐ Few_good.	_radar_fixes_on	e_direction		No_radar_fixes	
	Radar cloutter	None	Moderate_r	Heavy_rain	None	Moderate_r	Heavy_rain	None	Moderate_r	Heavy_rain
D	Position_can_be_established_easily	1	0.8	0.5	0	0	0	0	0	0
	Position_cannot_be_established_easily	0	0.2	0.4	1	0.9	0.6	0	0	0
	Position_cannot_be_stablished_by_radar	0	0	0.1	0	0.1	0.4	1	1	1

Quality of hazard markings

Grounding Ha				VI .		None						
Hazard markin ⊡ Enhanced				Nomal		Enhanced			─ Nomal			
Met. Visibility	Clear	Moderate	Low	Clear	Moderate	Low	Clear	Moderate	Low	Clear	Moderate	Low
▶ Very_Good	1	0.8	0.4	0	0	0	0	0	0	0	0	0
Average	0	0.2	0.5	1	0.9	0.8	0	0	0	0	0	0
Poor	0	0	0.1	0	0.1	0.2	1	1	1	1	1	1



Dark green nodes

Head-on collision (Dark green node)

Waterway Complexity ☐ High				Moderate		□ Low			
Bridge team management	High	Enough	Inadequate	High	Enough	Inadequate	High	Enough	Inadequate
▶ x1	0	0	1	0	0	0	0	0	0
x2	0	0	0	0	0	0	0	0	0
×4	0	1	0	0	0	1	0	0	0
x8	1	0	0	0	1	0	0	0	1
x16	0	0	0	1	0	0	0	1	0
x32	0	0	0	0	0	0	1	0	0
x64	0	0	0	0	0	0	0	0	0

Powered grounding (dark green node)

	Waterway Complexity					High						
B	Bridge team management		High			Enough		☐ Inadequate				
Q	uality of Hazard Marking	Very_Good	Average	Poor	Very_Good	Average	Poor	Very_Good	Average	Poor		
▶	x1	0	0	0.5	0	0	0	0	0	0		
Г	x2	0	0.5	0.5	0	0	0.5	0	0	0		
Г	x4	0	0.5	0	0	0.5	0.5	0	0	0.5		
Г	x8	1	0	0	0	0.5	0	0	0.5	0.5		
Г	x16	0	0	0	1	0	0	0	0.5	0		
	x32	0	0	0	0	0	0	1	0	0		
	x64	0	0	0	0	0	0	0	0	0		

	Waterway Complexity					Moderate						
Br	ridge team management		High			Enough		☐ Inadequate				
Qı	uality of Hazard Marking	Very_Good	Average	Poor	Very_Good	Average	Poor	Very_Good	Average	Poor		
E	x1	0	0	0.75	0	0	0	0	0	0		
Г	x2	0	0.75	0.25	0	0	0.75	0	0	0		
Г	x4	0	0.25	0	0	0.75	0.25	0	0	0.75		
Г	x8	1	0	0	0	0.25	0	0	0.75	0.25		
Г	x16	0	0	0	1	0	0	0	0.25	0		
Г	x32	0	0	0	0	0	0	1	0	0		
Г	x64	0	0	0	0	0	0	0	0	0		

Waterway Complexity	Ξ				Low					
Bridge team management		High			Enough		☐ Inadequate			
Quality of Hazard Marking	Very_Good	Average	Poor	Very_Good	Average	Poor	Very_Good	Average	Poor	
▶ x1	0	0	1	0	0	0	0	0	0	
x2	0	1	0	0	0	1	0	0	0	
x4	1	0	0	0	1	0	0	0	1	
x8	0	0	0	1	0	0	0	1	0	
x16	0	0	0	0	0	0	1	0	0	
x32	0	0	0	0	0	0	0	0	0	
x64	0	0	0	0	0	0	0	0	0	

Bright green Nodes

The bright green nodes contain an initial causation factor which is multiplied by the outcome of the dark green nodes.

8 Conclusions

A Bayesian network for estimating the causation factors at specific legs or junction points has been established. The overall structure of the network is thought to model the dependences of the causation factors well. However the actual probability tables within the network are based on judgment. They can most likely be improved.

An application for testing the network has been created. This means that one does not have to use the GeNie software in order to use the network.

9 References/Literature

Friis-Hansen, P., 2008. IWRAP MK II. Basic Modeling principles for prediction of collision and grounding frequencies. Technical document. 59 p.

Friis Hansen, P and Pedersen, P.T.: "Risk Analysis of Conventional and Solo Watch Keeping" Submitted to Int. Maritime Organisation (IMO) Maritime Safety Committee by Denmark at the 69th Session. 1998.

Fujii, Y. Yamanouchi, H and Mizuki, N.: "Some Factors Affecting the Frequency of Accidents in Marine Traffic. II: The probability of Stranding, III: The Effect of Darkness on the Probability of Stranding". Journal of Navigation, Vol. 27, 1974.

MacDuff, T.: "The Probability of Vessel Collisions". Ocean Industry, September 1974. pp. 144-148.

