

Underwater Navigation – Autonomous Underwater Vehicles (AUV)

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1) Inertial Navigation System (INS)

a. What are Error Sources and Failure Modes? --> Error Sources and Failure Modes

Maritime underwater navigation employing Inertial Navigation Systems can be impacted by multiple error causes and failure kinds. Detecting these challenges, isolating their sources, and insuring the continuation of operation involves the use of redundancy, sensor fusion, compensation algorithms, and careful sensor selection. By applying these approaches, the accuracy and dependability of integrated navigation systems can be greatly boosted, making underwater navigation more robust and dependable.

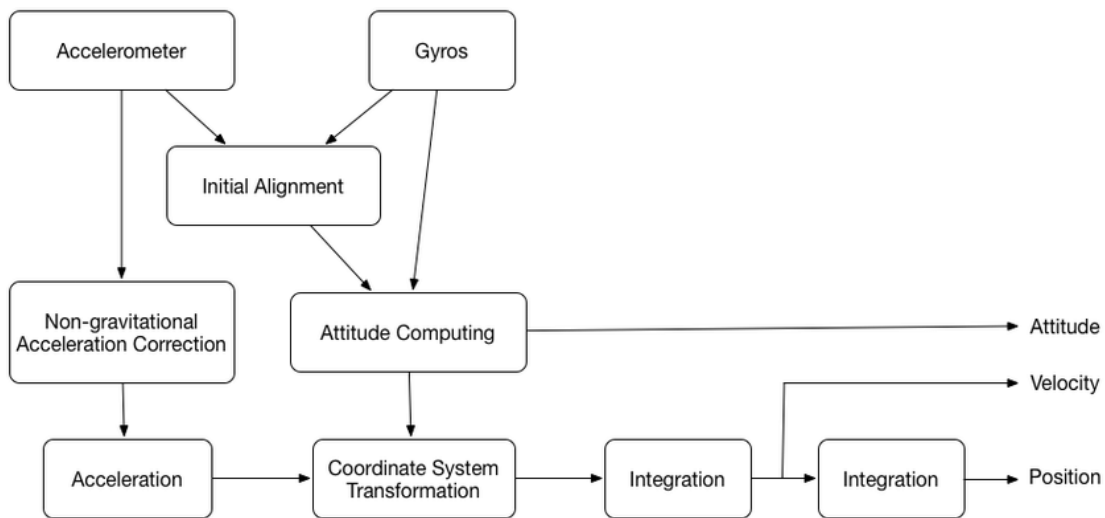


Figure 1: Inertial Navigation System (INS) process

Source: (Research Gate, 2022)

Inertial Navigation Systems (INS) are extensively utilised for underwater navigation in maritime applications. However, they are prone to several error causes and failure modes that could compromise their accuracy and reliability.

- **Gyro Drift:** INS relies on gyroscopes for measuring angular rates. Over time, gyros can experience drift, generating inaccuracies in the determination of orientation. This can lead to inaccurate heading, roll, and pitch angles.
- **Accelerometer Bias:** In the words of Shen *et al.* (2020), accelerometers measure linear accelerations, yet they can suffer from bias when the measured acceleration deviates from the true acceleration. This can introduce inaccuracies in velocity and position computations.

- **Vibration and Shock:** Underwater circumstances can subject INS to strong vibrations and shocks, which can dramatically impair the accuracy of gyroscopes and accelerometers.
- **Temperature Effects:** Temperature variances might induce changes in the properties of the inertial sensors, resulting in inaccuracies in their measurements.
- **Cross-Coupling:** Cross-coupling is the phenomenon where movement in one axis changes the readings in another axis due to mechanical coupling between sensors. This can pose complications in the orientation estimation.

b. How to detect the failure? -→ Detection

To detect these failure modes:

- **Gyro Drift:** Monitor the constancy of angular rate data over time. Significant and consistent deviations from the expected rate of change indicate gyro drift.
- **Accelerometer Bias:** Compare accelerometer findings with known gravitational acceleration when the INS is stationary. Deviations from the projected value show accelerometer bias.
- **Vibration and Shock:** Implement vibration and shock sensors alongside inertial sensors. Sudden variations in these parameters could signify an adverse impact (Huang, 2019).
- **Temperature Effects:** Include temperature sensors to monitor changes in the sensor environment. Significant temperature variations can provide warnings regarding potential sensor issues.
- **Cross-Coupling:** Perform controlled motion studies, moving the system in one axis while watching the measurements in other axes. Deviations from the predicted behaviour can demonstrate cross-coupling effects.

c. How to isolate? -→ Fault Isolation

Once faults are found, it's crucial to pinpoint the origin of the problem:

- **Gyro Drift:** Implement sensor fusion techniques that combine INS data with external sources like GPS or DVL (Doppler Velocity Log) to account for orientation difficulties caused by gyro drift.

- **Accelerometer Bias:** Calibrate accelerometers routinely to correct for bias. Implement redundant accelerometers and compare their outputs to locate the biased sensor.
- **Vibration and Shock:** Implement sensor fusion techniques to filter out the influences of vibrations and shocks. This can need combining complementing sensor types, such as pressure sensors, to accommodate for height variations induced by underwater disturbances (Huang, 2019).
- **Temperature Effects:** Apply temperature compensation techniques that alter sensor data depending on temperature variations. Additionally, isolate sensors from strong temperature changes.
- **Cross-Coupling:** Develop mathematical models to forecast cross-coupling effects and use these models in the sensor fusion algorithms to restrict their influence.

d. How to continue operation? → Continuity of Operation

To ensure ongoing operation despite failures:

- **Gyro Drift:** Combine INS with external navigation sources, such as GPS or DVL, to keep proper heading information when gyro drift occurs.
- **Accelerometer Bias:** Implement redundant accelerometers and switch to the one with the least bias. Regularly calibrate sensors to correct bias-related inaccuracies.
- **Vibration and Shock:** Use sensor fusion to filter out vibrations and shocks. If necessary, consider adaptive filtering algorithms that adjust filter parameters based on sensor activity.
- **Temperature Effects:** Morales and Kassas (2021), advocated applying temperature compensation algorithms and creating a system to perform within a certain temperature range to minimize temperature-induced faults.
- **Cross-Coupling:** Incorporate cross-coupling compensation into the navigation algorithms. Consider employing different sensor types to check the readings and eliminate cross-coupling effects.

2) Doppler Velocity Log (DVL)

a. What are Error Sources and Failure Modes? -→ Error Sources and Failure Modes

Underwater Navigation systems play a vital role in marine operations, ensuring the safe and accurate positioning of underwater vehicles and vessels. Maritime Underwater Navigation technologies, particularly the Doppler Velocity Log, provide crucial velocity data for integrated navigation systems. However, multiple error sources and failure kinds might degrade their accuracy.

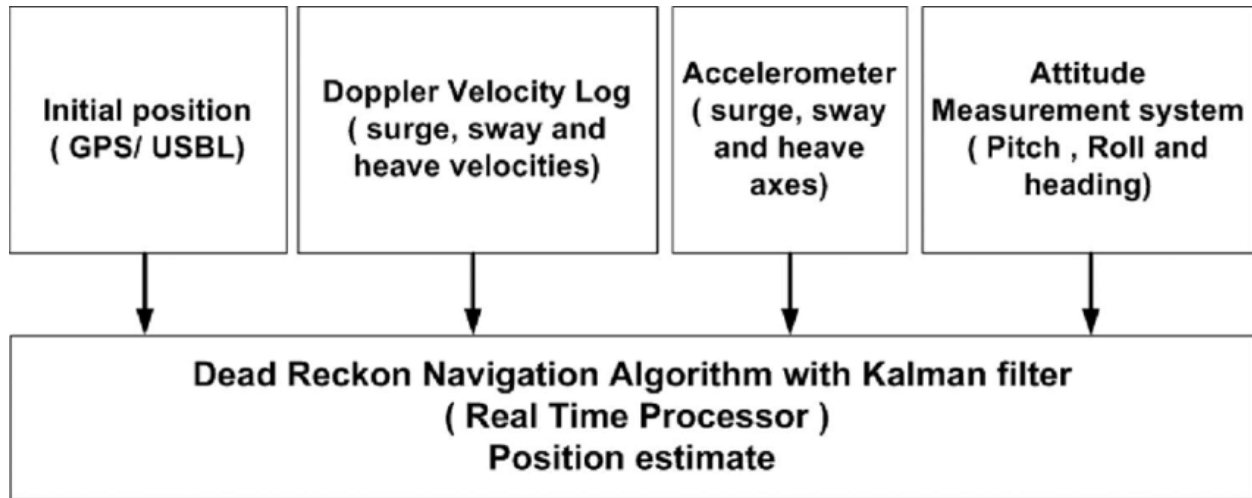


Figure 2: Architecture of Doppler Velocity Log (DVL)

Source: (Research Gate, 2022)

By incorporating powerful detection techniques, effective fault separation tactics, and rules for sustaining the continuity of operation, these technologies can minimize errors and contribute to safer and more reliable maritime operations. It's vital to continuously develop signal processing techniques, sensor redundancy, and integration with other navigation sensors to increase the overall performance of underwater navigation systems. In the words of Fukuda and Kubo (2022), The Doppler Velocity Log (DVL) is a critical component of integrated navigation systems that delivers real-time velocity measurements by utilising the Doppler effect of acoustic signals reflected off the seafloor. Despite its advantages, the DVL is subject to various error sources and failure mechanisms that could decrease the accuracy of navigation solutions.

Error sources and failure modes in DVL-based underwater navigation devices can come from several variables, ranging from ambient situations to technology restrictions. These sources can be broadly categorized as follows:

- ***Doppler Shift Variation due to Scattering and Attenuation:*** One large error source derives from fluctuations in Doppler shift produced by scattering and attenuation of acoustic signals in the underwater environment. These variations can result in erroneous velocity computations and navigation data.
- ***Sediment & Particulate Interference:*** Suspended particles and sediment in the water column can interfere with the DVL's acoustic signals, leading to signal degradation and reduced measurement accuracy (Yona and Klein, 2023). This is especially evident in murky waters or regions with high silt concentrations.
- ***Multi-Path Propagation:*** Reflections and refractions of acoustic signals from underwater structures or the seabed can introduce multi-path propagation, where signals take numerous paths before reaching the transducer. This can result in erroneous velocity estimates due to signal interference.
- ***Sensor Misalignment:*** Misalignment of the DVL sensor with the vessel's motion can lead to angular errors in velocity data. These offsets might propagate navigation solutions and generate navigation problems.
- ***Acoustic Interference from Other Systems:*** Other underwater acoustic systems, such as sonars, can interfere with DVL measurements, causing signal distortion and a decrease in velocity estimates.
- ***Temperature and Pressure Variations:*** Changes in water temperature and pressure can alter sound speed, leading to mistakes in Doppler shift calculations and ultimately affecting velocity measurements.
- ***Calibration Drift:*** Over time, calibration parameters of the DVL can drift, resulting in erroneous velocity estimates and navigation solutions.
- ***Sensor Noise and Resolution Limitations:*** Sensor noise, inherent in any measurement system, can contribute to imprecise velocity measurements, particularly at low velocities. Moreover, limits in sensor resolution can impair the accuracy of small-scale movements.
- ***Connection Loss:*** Data connection between the DVL and other components of the integrated navigation system might be disrupted owing to electromagnetic interference or physical damage to communication cables. This can result in a lack of real-time velocity and location updates.

b. How to detect the failure? --> Detection

- **Signal Reflection and Multipath:** Acoustic signals could bounce off various surfaces before reaching the DVL, resulting in erroneous velocity predictions. Detection comprises assessing the consistency of the received signals with the expected ones. Advanced algorithms can distinguish abnormal signal patterns, indicating multipath interference.
- **Sensor Drift:** Over time, sensors within the DVL can experience drift, contributing to erroneous velocity data. Monitoring sensor outputs against a constant reference, such as a gyroscope or external positioning system, can help detect severe drift (Xu, Guo and Hu, 2021).
- **Cavitation and Bubble Interference:** Bubbles formed by the motion of the vehicle or other reasons could interfere with acoustic signal reception, resulting in lower accuracy. Monitoring the signal strength and its consistency over time helps indicate the presence of cavitation effects.
- **Sediment Scattering:** Acoustic waves can scatter off sediment particles in the water column, generating inaccuracies in velocity estimation. Detecting anomalous fluctuations in velocity data relative to a known reference can imply sediment scattering effects.

c. How to isolate? --> Fault Isolation

- **Sensor Malfunction:** If a sensor within the DVL fails or produces irregular data, isolating the problematic sensor is crucial. Cross-comparing the outputs of redundant sensors (if available) and comparing them with previous data can assist pinpoint the faulty sensor.
- **Multipath Interference:** In the case of detected multipath interference, the DVL should utilise advanced signal processing methods to discriminate between the direct and reflected signals. Filtering methods like beamforming and adaptive signal processing can extract erroneous signals.
- **Cavitation and Bubble Interference:** When cavitation or bubble interference is seen, altering the vehicle's depth or orientation could decrease the effects. Additionally, Luo, Chen and Luo (2020), opined that signal-processing algorithms can identify periods of high interference and discard matching data points from velocity computations.

- **Sediment Scattering:** To lessen sediment scattering effects, altering the DVL's parameters or putting more sensors to detect sediment density can help. Algorithms can then adjust for the scattering effects based on the sediment information.

d. How to continue operation? → Continuity of Operation

- **Sensor Failure:** If a sensor fails, switching to redundant sensors or relying on other navigation sensors like Inertial Navigation Systems (INS) can assure ongoing operation. A well-designed navigation system should feature seamless sensor swap routines.
- **Multipath Mitigation:** Advanced signal processing techniques, such as adaptive beamforming, can be utilised to minimise multipath interference. By dynamically altering the receive beam pattern, the system may focus on the direct signal and decrease reflected signals.
- **Cavitation and Bubble Interference:** In the presence of cavitation or bubble interference, the DVL can temporarily switch to other velocity sources like INS. The system can then restart utilising DVL data once interference levels fall.
- **Sediment Scattering Compensation:** Algorithms that simulate sediment scattering effects can be integrated into the navigation solution. These algorithms modify velocity data based on expected silt density, allowing the system to deliver accurate velocity calculations despite scattering.

3) Ultra Short Baseline (USBL)

a. What are Error Sources and Failure Modes? → Error Sources and Failure Modes

Underwater Navigation technology, notably Ultra Short Baseline (USBL) systems, play a vital role in maritime operations, including subsea exploration, offshore construction, and marine research. These technologies provide accurate placement and tracking of underwater vehicles and equipment. However, like any intricate technology, Ji, Nguyen and Kim (2019), observed that USBL systems are prone to multiple error causes and failure mechanisms that could decrease their performance.

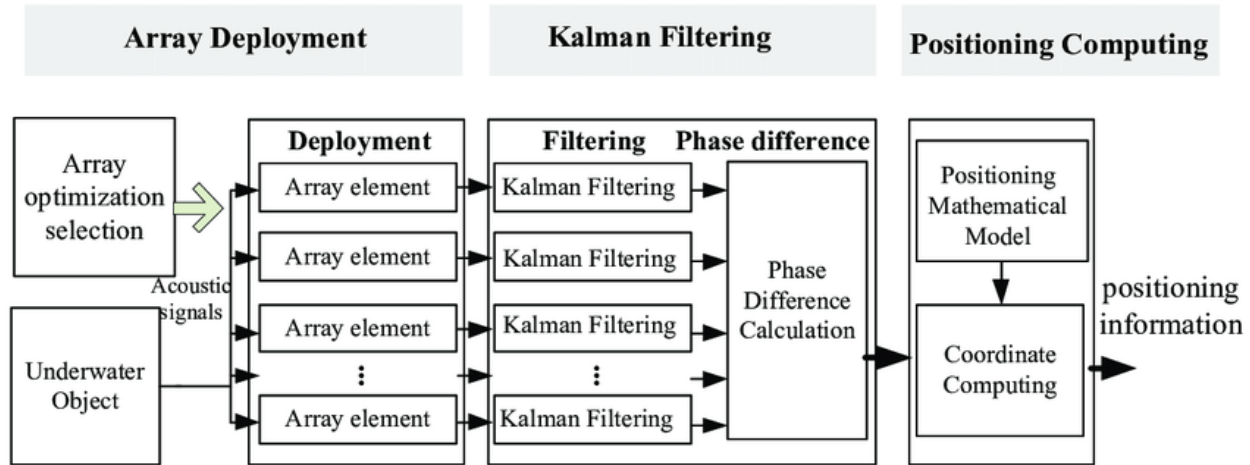


Figure 3: The framework of (USBL)

Source: (Research Gate, 2022)

Understanding these error sources and failure modes is crucial for devising effective strategies to detect, isolate, and eliminate them, ensuring the reliability and accuracy of integrated navigation systems. Error origins and failure situations in Ultra Short Baseline (USBL) underwater navigation systems can come from several variables. These can be roughly classified into environmental, technical, and operational sources.

- **Environmental Sources:** Underwater navigation is influenced by environmental factors such as water density fluctuations, sound speed profiles, temperature gradients, and salinity shifts. These factors can cause acoustic signal refraction, leading to erroneous range and location.
- **Technical Sources:** Technical failures can stem from equipment deficiencies, sensor drift, electronic noise, and hardware difficulties. Transducer misalignment, signal interference, and multipath effects (when signals bounce off barriers and arrive at the receiver by various channels) are prominent concerns.
- **Operational Sources:** Errors can result from erroneous system setup, bad calibration, inappropriate installation, and inadequate operational processes. Human errors, misunderstanding of data, and inadequate training might contribute to these issues (Ji, Nguyen and Kim, 2019).

- **Acoustic Interference:** Acoustic interference can occur from sources such as other vessels, marine life, or ambient noise. This interference can lead to distorted or delayed signals, lowering the accuracy of USBL measurements.
- **Multipath Effects:** Multipath originates when audio waves reflect off underwater barriers before reaching the transceiver. This can result in erroneous distance measures, leading to inaccuracies in position calculations.
- **Sensor Drift:** Over time, sensors employed in USBL systems can exhibit drift owing to temperature changes, pressure differences, or ageing. This could lead to gradual inaccuracies in position estimations.
- **Signal Attenuation:** Acoustic transmissions might face attenuation as they flow through the water column. This attenuation reduces the signal intensity, sometimes leading to inaccurate distance estimations.
- **Geomagnetic Disturbances:** Geomagnetic anomalies can impair the accuracy of USBL systems that rely on magnetic compasses for heading information. These disturbances can lead to heading mistakes and ultimately compromise position accuracy.

b. How to detect the failure? --> Detection

To detect problems in USBL underwater navigation systems, different approaches can be employed:

- **Redundant Sensors:** Implement redundant sensor designs, such as many transducers or hydrophones. If one sensor fails, others can yield trustworthy measurements.
- **Outlier Analysis:** Continuously monitor ranging and location data. If quick and substantial deviations are discovered, it could signal a sensor malfunction.
- **Data Comparison:** Compare the output of different sensors or navigation systems. If there is a considerable gap between readings, it may suggest a problem.
- **Signal Quality Metrics:** Monitor the quality of received auditory signals. Poor signal-to-noise ratios or abnormalities in transmission patterns could imply interference or hardware difficulties.

c. How to isolate? --> Fault Isolation

When a breakdown is found, it's vital to determine the cause of the problem to prevent future degradation of the navigation system. This involves:

- **Sensor-by-Sensor Analysis:** Temporarily deactivate or isolate questionable sensors while keeping the rest of the system operating. By analysing the behaviour of the remaining sensors, you can narrow down the problematic component (Graça, 2020).
- **Calibration Check:** If possible, execute on-the-spot calibration checks for sensors. Misaligned or badly calibrated sensors can be identified and recalibrated.
- **Signal Analysis:** Examine acoustic signal patterns for multipath interference. Adjust the signal processing methods to limit the influence of multipath effects.

d. How to continue operation? → Continuity of Operation

In the presence of failures, it's necessary to preserve operational continuity to ensure safe navigation. Strategies include:

- **Sensor Fusion:** Utilize sensor fusion techniques to aggregate data from numerous sources, such as USBL, inertial navigation systems (INS), and depth sensors. This boosts robustness and accuracy even if one sensor fails.
- **Fallback Modes:** Implement preset fallback modes that rely on alternate navigation methods or lesser accuracy modes. These modes can give rudimentary navigation skills while the issue is fixed (Graça, 2020).
- **Manual Intervention:** Design the system to allow manual input for important navigation decisions. Operators can manually update position estimates based on visual cues or alternative sensor data.
- **On-the-fly Reconfiguration:** Develop the system to dynamically reconfigure itself in response to sensor failures. The system may modify its navigation method to the available sensors.

4) Long Baseline (LBL)

a. What are Error Sources and Failure Modes? → Error Sources and Failure Modes

Underwater navigation technology, particularly Long Baseline (LBL) systems, plays a significant role in assuring precise location and navigation for varied maritime activities.

However, these systems are prone to several error sources and failure modes that could compromise their accuracy and reliability. These errors might occur from environmental factors, hardware failures, signal interference, and human errors. Understanding these error sources and failure modes is crucial for creating effective strategies to mitigate their influence on integrated navigation systems.

- ***Acoustic Signal Propagation Variability:*** Denton and Gehrlein (2023), claimed that LBL systems utilize acoustic signals to estimate distances between transponders and the receiver. Variations in sound speed owing to varying water temperature, salinity, and pressure can produce inaccuracies in distance estimations. These deviations create erroneous range and location estimates.
- ***Transponder Deployment Errors:*** Incorrect deployment of transponders on the seafloor could lead to inaccurate baseline measurements. Miscalculations in baseline lengths affect the accuracy of location triangulation.
- ***Transponder Hardware problems:*** Brahma and Giri (2023), opined that Transponders may face hardware problems, resulting in the transmission of erroneous acoustic signals or no signals at all. This can lead to missing or inaccurate data in the positioning calculations.
- ***Acoustic Signal Interference:*** Noise from other underwater sources, such as ships, marine life, or natural phenomena, could interfere with the acoustic signals utilised in LBL systems. This interference might distort the received signals and induce mistakes.
- ***Multi-path Effects:*** Acoustic signals can bounce off underwater structures or the seafloor, providing many signal paths. These multi-path effects can contribute to signal delays and mistakes in distance estimations.
- ***Calibration inaccuracies:*** Improper calibration of transponder hardware, sound velocity profiles, or sensor alignment might generate systematic inaccuracies in the location calculations.

b. How to detect the failure? -→ Detection

Detecting malfunctions and errors in Long Baseline (LBL) systems is critical for maintaining the integrity of underwater navigation systems.

- **Real-time Monitoring:** Continuous monitoring of acoustic signal quality, transponder responses, and baseline consistency helps detect anomalous behaviour.
- **Redundant Measurements:** Employing redundant transponders and sensors enables cross-validation of measurements. Inconsistent data can reveal potential errors.
- **Signal Quality Metrics:** Analyzing signal-to-noise ratios and signal strength can help discover signal interference and multi-path effects.

c. How to isolate? → Fault Isolation

Once errors are found, pinpointing their sources is crucial to avoid the transmission of inaccurate information.

- **Diagnostic Algorithms:** Implementing diagnostic algorithms can pinpoint the individual transponder or hardware component responsible for the issue.
- **Comparative Analysis:** Comparing measurements from multiple transponders and sensors can uncover inconsistencies and aid in isolating the defective item.
- **Environmental Analysis:** Integrating data on water conditions and environmental characteristics might assist distinguish between errors caused by propagation variability and device faults.

d. How to continue operation? → Continuity of Operation

Ensuring the system can continue functioning in the event of failures is vital, especially for safety-critical applications.

- **Adaptive Algorithms:** Algorithms that dynamically adjust for changing signal conditions or eliminate outlier measurements can maintain accurate location even in the presence of occasional faults.
- **Redundancy and Diversity:** Redundant transponder networks, multiple navigation systems (such as LBL paired with inertial navigation), and sensor fusion approaches boost system robustness.
- **Fallback Modes:** Systems can feature fallback modes that rely on alternative navigation sources or reduced accuracy criteria when primary measurements are disrupted (Fernández, 2019).

- **Human Intervention:** Operators can manually intervene, choosing to omit or correcting erroneous measurements based on real-time analysis.

5) Sparse LBL or Single Beacon Technique

a. What are Error Sources and Failure Modes? → Error Sources and Failure Modes

Inaccurate navigational solutions may result from error causes and failure mechanisms in marine underwater navigation systems, particularly Sparse Long Baseline (LBL) or Single Beacon Techniques. Among these problems are:

- **Acoustic Propagation Variability:** Luo and Ko (2022), claimed that acoustic signals might diverge from their anticipated trajectories due to changes in water's sound speed, which can result in range errors.
- **Multipath Interference:** Acoustic signals may be scattered by submerged objects, providing many arrival pathways and complicating range estimation.
- **Environmental disturbances:** Range inaccuracies are introduced by ocean currents, temperature gradients, and salinity changes, which impact sound speed profiles.
- **Sensor Alignment Error:** Misinterpretation of bearing and range readings might result from improper sensor alignment.
- **Sensor Noise:** Unpredictable noise in sensor data might cause range and bearing predictions to be inaccurate.

Strategies for mitigation include:

- **Diverse Sensor Fusion''** may increase robustness by fusing information from various sensors, such as Inertial Navigation Systems (INS), Doppler Velocity Log (DVL), and Ultra-Short Baseline (USBL) systems.
- **Environmental modelling:** Adapting to environmental changes in real-time may lessen the influence they have on navigational accuracy.
- **Kalman Filtering:** Extended Kalman Filter (EKF) filtering methods may be used to combine sensor data and forecast precise vehicle locations.

- **Error Budgeting:** By allocating error budgets to sensors, Masmitja, Gomariz and Navarro (2022), observed that the system can identify and correct the measurements that are most prone to mistake.

Monitoring discrepancies between sensor readings, contrasting sensor outputs with expected behaviour, and highlighting anomalies are all steps in the detection process. Isolation comprises locating the defective sensor using voting or outlier detection techniques. The system may change to alternative sensors, modify its navigation plan, or give priority to trustworthy sensors while degrading less reliable ones to continue operating.

b. How to detect the failure? --> Detection

Monitoring multiple system components to spot abnormalities or inconsistencies is necessary to detect faults in the Sparse LBL or Single Beacon Technique. Important techniques for spotting failures include:

- **Residual Analysis:** Residual Analysis is a continuous comparison of predicted and observed ranges between the AUV and sparse LBL beacons. Potential failure is indicated by significant variances.
- **Divergence from projected Trajectory:** Investigating if the actual trajectory of the AUV considerably differs from its projected route based on LBL data, which may point to a problem with beacon location or AUV navigation.
- **Redundancy Check:** Verifying consistency by comparing data from many sparse LBL beacons. A malfunction in one of the beacons or the ranging mechanism of the AUV may be indicated by discrepancies in beacon ranges (Rypkema, Schmidt and Fischell, 2021).
- **Statistical Outliers:** Recognising range readings that deviate from the predicted statistical distribution, which may point to a malfunctioning beacon or an incorrect measurement.
- **Sensor Fusion:** combining information from several sensors, including depth sensors, Doppler velocity records, and inertial navigation systems (INS). LBL failure may be indicated by discrepancies between LBL and other sensor data.
- **Communication Health Monitoring:** Observing the communication connection quality between the AUV and LBL beacons.

Communication deterioration or abrupt decreases may be an indication of a failed beacon or underwater impediments (Rypkema, Schmidt and Fischell, 2021).

- **Environmental monitoring:** keeping an eye on factors including temperature, salinity, and water currents. Extreme alterations might cause range measurements to be interpreted incorrectly and signal a probable failure.
- **Comparison of Prediction and Reality:** contrasting the actual AUV location as measured by other sensors with the expected AUV position based on LBL observations. Significant departures could indicate LBL failure.
- **Position Consistency:** Examining how consistently the locations of the AUVs derived from several LBL beacons are consistent. Inconsistencies might be a sign of a malfunctioning navigation system or a beacon.

These techniques may be used to identify faults in the Sparse LBL approach while continually monitoring the data. A prompt reaction, separation of the impacted components, and maintenance of overall navigation accuracy are all made possible by such detection.

c. How to isolate? → Fault Isolation

A small number of acoustic beacons set on the bottom are employed in the sparse LBL, also known as the single beacon technique, to establish the location of an autonomous underwater vehicle (AUV). The accuracy of Sparse LBL may be impacted by a variety of error causes and failure mechanisms, much like any other navigational device. The following list of possible failure types is followed by some suggestions for mitigating them:

Failure modes include:

- **Beacon Malfunction:** Position estimations may be off if one or more beacons malfunction or send out the wrong signals.
- **Acoustic Signal Interference:** The signals that the AUV's hydrophone picks up may be distorted by environmental elements such as underwater noise, multipath reflections, and acoustic interference (Rypkema, Schmidt and Fischell, 2021).

Mitigation and isolation of the fault:

Beacon Malfunction

- **Detection** Constantly keeps an eye on all beacon signals. A problem may be revealed by differences from anticipated signal properties.
- **Isolation:** Take the faulty beacon out of the equation for determining location.
- **Continued Operation:** Maintain navigation by using the locations from the remaining operational beacons.

Acoustic Signal Interference -

- **Detection:** Examine signal properties for irregularities brought on by interference or multipath effects.
- Implement signal processing algorithms to identify and eliminate distorted signals during isolation.
- **Continued Operation:** To lessen the effects of interference, use signals from unaffected beacons and sophisticated algorithms.

The Sparse LBL system can detect problems, pinpoint their causes, and adjust to keep giving the AUV precise navigational information by putting these fault isolation techniques into practice. These steps help to provide safe and reliable underwater navigation by reducing mistakes in the integrated navigation system.

d. How to continue operation? → Continuity of Operation

Sparse Long Baseline (LBL) navigation, sometimes referred to as the Single Beacon Technique, is a technique for underwater acoustic beacon navigation. Despite its benefits, the integrated navigation system is prone to several error sources and failure modes that might compromise its accuracy. The following steps may be done to reduce these mistakes and guarantee operational continuity:

Sources of errors and failure modes

- **Acoustic Signal Attenuation:** Acoustic signals may weaken in the underwater medium, resulting in lower signal intensity and erroneous distance estimations.
- **Multipath Propagation:** Acoustic waves may be scattered by submerged objects, creating several signal pathways and erroneous range estimations.
- **Environmental Disturbances:** Temperature changes and underwater currents may skew measurements of range by altering the speed of sound in water.

Mitigation Techniques

- **Signal processing strategies:** It is possible to reduce the impacts of signal attenuation and multipath propagation by using advanced signal processing methods. These techniques may assist in deriving precise range data from the signals that were received.
- **Environmental Compensation:** To account for fluctuations in the sound speed in the water, real-time monitoring and compensation for changes in water temperature and currents may be used.
- **Redundant Sensors:** Including redundant sensors, such as Doppler velocity logs (DVL) and inertial navigation systems (INS), may offer more data sources for redundancy and error correction.

The continuation of operations:

- Data Fusion Includes information from additional sensors, such as INS and DVL, in the navigation system. As a result, the system can continue to estimate positions accurately even if the sparse LBL beacon signal starts to falter.
- Use Kalman filtering methods to aggregate readings from several sensors and control the inherent uncertainties in each sensor. This may aid in maintaining a reliable and precise navigation system.
- Implement a system of continuous sensor health monitoring for the sparse LBL beacon. The system may automatically transition to relying more heavily on other sensors while

preserving the sparse LBL as a secondary input if the signal quality declines beyond a specific level.

- **Dynamic reconfiguration:** Make the navigation system capable of changing its settings on the fly depending on the quantity and calibre of sensor data. The system can modify and re-calibrate its navigation solution if the sporadic LBL beacon signal is lost or unreliable.
- **Human Intervention** Human guidance may be used to direct the autonomous underwater vehicle (AUV) back to a known reference point or a safe area in severe circumstances if the sparse LBL beacon entirely fails.

By putting these tactics into practice, it is possible to keep an integrated navigation system using sparse LBL operational even when error sources and failure modes are present. For autonomous underwater vehicles and other marine applications, this guarantees accurate underwater navigation.

6) Depth Pressure Sensor

a. What are Error Sources and Failure Modes? --> Error Sources and Failure Modes

Depth Pressure Sensor Error Sources and Failure Modes in Underwater Navigation

- **Sensor Drift:** Inaccurate depth readings might result from gradual changes in sensor calibration or sensitivity over time.
- **Temperature Effects:** Depth sensors are susceptible to temperature fluctuations, which may lead to changes in pressure readings when the density of the water varies.
- **Pressure Sensor Malfunction:** Internal pressure sensor component failures might lead to inaccurate depth measurements.
- **Variations in water density:** According to Song (2020), changes in water salinity may have an impact on pressure readings, which can result in inaccurate depth estimates.
- **Water Currents:** The sensor might experience extra pressure from strong currents, which results in a divergence from the true depth.
- **Biofouling:** Marine creatures may build up on the sensor, impairing its functionality and producing false depth readings.

Monitoring and Reduction:

- **Sensor Health Monitoring:** Conduct routine sensor performance and calibration checks to identify sensor drift and problems.
- **Temperature Compensation:** Adapt pressure readings for temperature effects by integrating temperature sensors.
- **Redundancy:** Use additional pressure sensors to cross-check depth readings and locate broken sensors.
- **Filtering Algorithms:** Use algorithms to eliminate pressure changes that occur quickly due to outside influences.

Continuation and isolation:

- **Redundancy Switchover:** If one sensor malfunctions or drifts, the system switches to data from the redundant sensor while warning the operators.
- **Dynamic Thresholds:** Use dynamic threshold values to find sensor irregularities and sound alerts when limits are crossed.
- **Fallback to Other Sensors:** If the pressure sensor's depth data is inaccurate, depend more on DVL or INS for depth calculation.
- **Operator Override:** In the event of persistent sensor failures, this feature enables operators to manually enter depth data, allowing a cautious operation to continue.

b. How to detect the failure? --> Detection

Failures in a Depth Pressure Sensor (DPS) may be found by:

- **Inconsistent Readings:** Consistently contrasting sensor readings with anticipated values.
- **Zero Pressure Reading:** Maintain a consistent output of zero pressure.
- **Drift:** Monitoring slow, long-term changes in sensor outputs.
- **Pressure Spikes:** Spotting abrupt, significant pressure fluctuations that are unrelated to changes in depth.
- **Communication Loss:** Keeping an eye out for sluggish sensor data updates.
- **Temperature Effects:** Monitoring for unusual pressure fluctuations brought on by temperature changes.

c. How to isolate? -→ Fault Isolation

There are numerous processes involved in isolating problems in a depth pressure sensor for an integrated underwater navigation system:

- **Redundancy:** Using several depth pressure sensors may aid in spotting inconsistencies in sensor results. The most trustworthy reading may be chosen using a voting system if one sensor offers conflicting data (Rahman, Li and Rekleitis, 2019).
- **Comparative Analysis:** Comparing the depth pressure sensor output regularly with other sensor data, such as GPS altitude or water density calculations, might assist in spotting abnormalities or outliers.
- **Threshold Monitoring:** Establish predetermined thresholds for accurate depth readings. A probable failure may be indicated if the sensor reading deviates beyond certain levels and raises an alarm.
- **Sensor Fusion:** Combining data from other sensors, such as USBL and DVL, may improve accuracy and provide a backup depth estimate if the pressure sensor malfunctions.
- **Sensor Health Monitoring:** Keep an eye on the sensor's internal health indicators, such as temperature drift and pressure drift, constantly to spot any slow deterioration that might result in incorrect results.
- **Statistical Analysis:** Examine historical data patterns using statistical approaches. A sensor fault may be indicated by rapid changes or discrepancies in the depth measurements.
- **Cross-Verification:** Use other navigation systems, such as Long Baseline (LBL) systems and Inertial Navigation Systems (INS), to cross-verify depth data. Significant discrepancies between the sensor data and these sources might indicate a malfunction.
- **Dynamic Reconfiguration:** Create the navigation system such that, in the event of sensor failure, it may dynamically reorganise itself to depend more heavily on other available sensor inputs.

The integrated navigation system can efficiently identify, isolate, and reduce errors caused by the depth pressure sensor by integrating various fault isolation techniques, guaranteeing precise underwater navigation even in the event of sensor failures.

d. How to continue operation? → Continuity of Operation

For the vehicle to be able to determine its depth underwater, depth pressure sensors are essential. The following actions may be performed to guarantee operation continuation in the event of failures:

- **Redundancy:** Install several deep pressure sensors on the AUV and compare the findings. To avoid inaccurate depth calculation, a sensor might be separated if it differs considerably from the others.
- **Implement threshold checks on sensor outputs:** The system can mark a sensor reading as potentially failing and switch to a backup sensor if it deviates from a certain range (Kumaresan, Ozioko and Dahiya, 2021).
- **Sensor Fusion:** Combine information from several sensors, such as the Doppler Velocity Log (DVL) or the Inertial Navigation System (INS), to cross-verify depth readings. Combining data from many sensors may improve accuracy and fault tolerance.
- **Error Modelling:** Create error models using data on previous sensor performance. The system can detect and respond to an anomaly if a sensor behaves differently than predicted.
- **Kalman Filtering:** Make use of Kalman filtering methods to calculate the real depth while taking into account the sensor's output and related uncertainty. This may lessen the effects of inaccurate sensor readings.
- **Dynamic Threshold Adjustment:** Apply adaptive threshold changes to the operating circumstances of the AUV. By doing this, false alerts during real fluctuations in sensor readings may be avoided.
- **Online Diagnostics:** Constantly monitor sensor performance and health. It is possible to exclude a sensor from the fusion process until it has been recalibrated or replaced if its dependability has declined beyond a specific threshold.
- **Data Logging and examination:** Record system behaviour and sensor readings for examination after the mission. This may help uncover patterns of sensor failure and improve the fault detection and isolation algorithms used by the system.

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