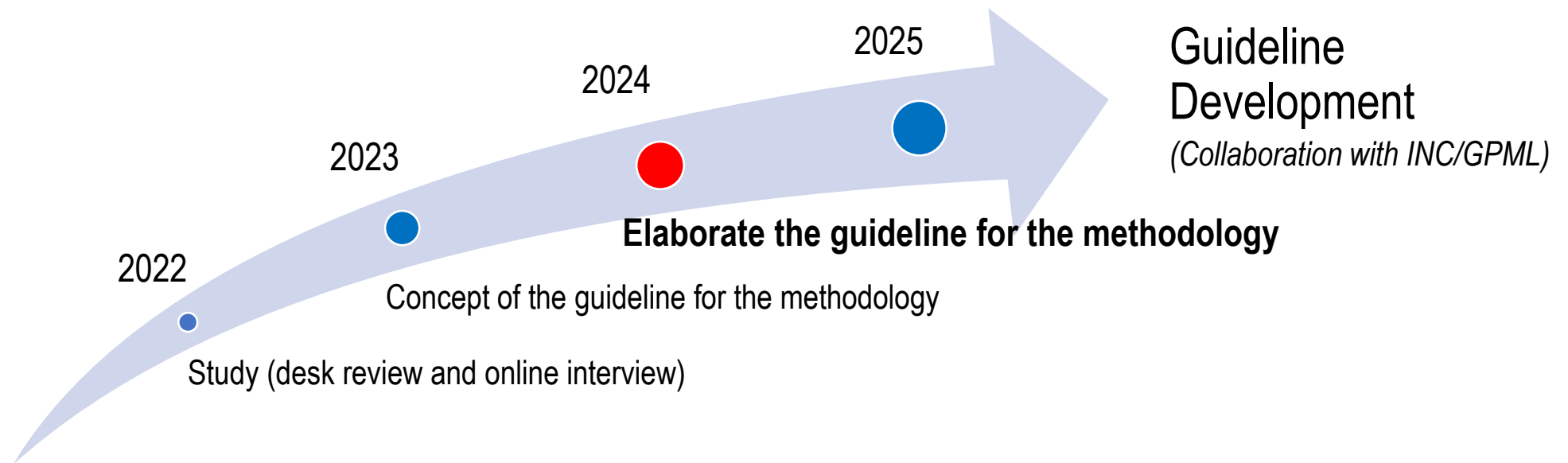


# **Project on Inventory Development Methodology for a Plastic Leakage into the Environment, Including the Marine Environment (FY 2024)**

## 1-1. Progress of the project on plastic inventory development methodology conducted by the Ministry of the Environment, Government of Japan

- **Global/National plastic pollution issue:** Project on understanding of the amount of plastic discharged into the ocean in Japan conducted by the Ministry of the Environment, Japan (MOEJ) **since 2020**
- **Overall project objective:** Develop and Elaborate the Plastic Discharge Inventory (macro/microplastic) with the estimation and evaluation methodologies in Japan
- **One of the component added in 2022:** Study on harmonized methodology for development of Plastic Discharge Inventory
- **Progress in 2023:** Develop the Classification of existing estimation methodologies for Plastic Discharge Inventory Development
- **Plan in 2024:** Elaborate the Classification of estimation methodology for Plastic Discharge Inventory Development



### 1. Background and Objectives

- Current status and challenges of global plastic pollution
- INC discussion on plastic inventory, leakage estimation, research, scientific evidence, NAP, and monitoring, etc.
- Japan's knowledge, experience, and international cooperation

### 2. What is Plastic Leakage Inventory

- Current status and challenges of inventory, monitoring, and estimation
- Science-based policy decisions and interventions

### 3. Target, Scope, and Scale

- Policy maker, Practitioner, Researchers, etc.
- Objectives of leakage estimation, estimation scale, dataset for estimation
- Definition of terms

### 4. Categorization of Estimation Methods

- Introduction of Japan's inventory and methods
- Categorization of existing plastic leakage estimation methods
- Overview of existing plastic leakage estimation methods

### 5. Development of Estimation Method

- Purpose of estimation, Definition of terms, Confirmation of necessary data, Feasibility of actual surveys, Setting of baseline year, Estimation frequency, and Scientific policy development and interventions
- Concept of "Step" approach / Decision tree / Estimation options (including "Step" approach)

### 6. Pilot Cases

- Indonesia
- Thailand
- Vietnam

2-1. Elaboration of the Classification Table for Plastic Leakage Estimation Method (M1: Production and consumption)				
Name		UNITAR, 2024, Statistical Guideline for Measuring Flows of Plastic Throughout the Life Cycle	OECD, 2020, ENV-Linkages Modelling Framework	Nakatani et al., 2020, Revealing the Intersectoral Material Flow of Plastic Containers and Packaging in Japan
Category		M1. Production and Consumption	M1. Production and Consumption	M1. Production and Consumption
Scale of Estimation	Spatial	Global, National, Regional	Global, Regional, National	National
	Temporal	Long-term, Continuous	Long-term (2019–2060), Continuous	Long-term (2000-2015)
Input Data (Variables)	Data Types	Plastic production and consumption data, waste management statistics, trade data	Economic flows, plastics use by sector, product lifespans, recycling rates, mismanaged waste rates	Resin type, processed forms, sector-specific usage
	Sources	SEEA CF, CPC, ISIC classification, national trade & waste datasets	OECD databases (e.g., GTAP 10), Ryberg et al., 2019、Geyer et al., 2017, regional economic data	Japan's input-output tables, industrial production and shipment statistics
	Plastic Granularity	Plastic Granularity: Polymer- and sector-based	Plastic Granularity: Polymer- and sector-based	Plastic Granularity: Resin-based, sector-based
	Sector Details	Packaging, transport, construction, electrical, textiles	Packaging, textiles, construction, consumer goods, transportation	Food packaging, industrial packaging, retail
Estimation Approach	Type of Model	Physical Supply and Use Tables (PSUT)	Computable General Equilibrium (CGE) Model	Input-Output Material Flow Analysis (IO-MFA)
	Description	Tracks plastic flows across production, use, and waste phases using the SEEA framework	Links sectoral economic projections to plastics use, waste, and environmental impacts; incorporates feedback from external models	Tracks intersectoral flows of plastic containers and packaging using IO tables
	Leakage Pathways	Managed and unmanaged waste, informal and illegal waste handling	MPW to terrestrial and aquatic environments, emissions from lifecycle stages	N/A (focus on waste streams and recycling)
	Assumptions	Homogeneity of classification systems; consistent unit measures across datasets	Homogeneity in polymer use, static material composition, fixed GDP-plastic waste relationships	Consistent resin usage across sectors; alignment with IO table precision
Output (Estimation Result)	Results:	Quantified plastic flows by sector and lifecycle stage	Plastics use (Mt), waste management (recycled, landfilled, incinerated, mismanaged), environmental impacts	Material flows of 4.8 Mt of containers/packaging; sectoral inflow and recycling rates
	Temporal Scale	Annual	Annual	Annual
	Accuracy, Validation	Alignment with SEEA standards, validated through expert reviews and pilot testing	Calibrated using observed data, verified via expert reviews and literature benchmarks	Validated against survey data and historical trends
Details of applied model	Name of the model	PSUT Framework	ENV-Linkages Model	IO-MFA Model
	Description of the model, equation,	Integrates environmental and economic data to map plastic lifecycle	Dynamic CGE model integrating sectoral and regional economic activities with plastic flows	Estimates plastic flows into demand sectors via product lifecycles
	Calibration	National statistical datasets	Based on GTAP 10 database, historical and regional waste data	Uses historical IO data for accuracy
	Verification	Expert reviews, alignment with international classifications	Benchmarked against observed recycling and waste Mgt. trends	Compared with waste generation and recycling data
Analysis of Practicality and Versatility	Granularity	Supports detailed sector-specific insights	Enables detailed polymer- and sector-specific analysis	Effective for detailed sector analysis, limited to micro-intervention
	Data Collection	Dependent on accessibility to consistent classification (Plastic Key)& trade data	High dependency on regional economic and waste statistics; sensitive to data availability	Dependent on IO table precision and sectoral statistics
	Impact Verification	Allows tracking of plastic lifecycle, waste flows, and recycling rates	Supports identification of plastic leakage sources, waste management efficiencies, and policy development	Identifies key sources and recycling potentials
Additional Notes	Strengths	Standardized methodology for global comparability; modular structure for phased implementation	Comprehensive integration of economic and environmental dimensions; long-term projections	Comprehensive sectoral flow insights; supports policy formulation
	Weaknesses	Reliance on data availability; complex for countries with limited statistical infrastructure	Complexity of calibration, reliance on static assumptions for polymer distribution	Relies on data aggregation; limited granularity for resin-specific flows
	Tool kit	Yes (webpage)	No	No
4				

2-2. Elaboration of the Classification Table for Plastic Leakage Estimation Method (M2: Waste Flow)					
Name		Lau et al., 2020, P2O Model	Cottom et al., 2024, SPOT Model	University of Leeds, 2019, ISWA Plastic Pollution Calculator	GIZ, 2020, Waste Flow Diagram
Category		M1. Production & Consumption, M2. WM	M2. Waste Management	M2. Waste Management	M2. Waste Management
Scale of Estimation	Spatial	Global	Global, sub-national (50,702 municipalities)	Global, regional, city, and district levels	Municipal/City Level
	Temporal	Long-term (2016-2040)	Temporal: Continuous, focusing on the year 2020	Long-term, with capacity for scenario analysis	Short-term to Long-term (annual estimates)
Input Data (Variables)	Data Types	Population size, per capita macroplastic MSW, microplastic product use, loss rates, collection rates	Municipal waste generation rates, waste composition, collection coverage, socioeconomic indicators	Waste generation rates, waste composition, collection coverage, socio-economic conditions, littering prevalence	Population, MSW generation rates, waste composition, collection efficiency, leakage influencers
	Sources	World Bank, UNEP, industry reports, scientific literature	Global and national databases, including WaCT, WABI, WaW2.0, and UNSD city data	Local waste management data, literature defaults, expert-reviewed leakage factors	Primary data collection, SDG 11.6.1 assessment, existing waste management studies, local waste authorities
	Granularity	Macro (MSW) and Micro (textiles, tires, pellets, PCPs)	Macroplastic (>5mm) by rigid and flexible types	Item-specific analysis: type, shape, size, and material	Macroplastic by material types: plastic film, dense plastic
	Sector Details	Packaging, consumer goods, municipal waste systems	Municipal solid waste management	MSWM, informal and formal waste collection	MSWM, informal and formal waste collection sectors
Estimation Approach	Type of Model	Coupled ODE Model, Monte Carlo Simulation	MFA combined with Machine Learning	Material Flow Analysis (MFA), GIS tool	Material Flow Analysis (MFA), Decision Trees
	Description	Calculates plastic flows through stocks and pathways using differential equations, accounting for feedbacks and uncertainties	Utilizes a bottom-up approach with municipal-level data, employing Monte Carlo simulations to handle uncertainties	Models plastic movement from waste generation through collection, treatment, and potential leakage pathways	Maps waste flows through the MSWM system, identifies stages where plastic leaks, and estimates quantities using predefined leakage factors
	Leakage Pathways	Land-based mismanagement, aquatic and terrestrial leakage pathways	Open burning, debris emissions from uncollected waste, collection systems, uncontrolled disposal, and mismanaged sorting reject	Open burning, on-land dispersion, entry into water bodies	Open burning, on-land dispersion, drains, entry into water systems
	Assumptions	Constant growth rates, stable informal sector impact, region-based differentiation	Assumes constant waste generation rates, regional socioeconomic influences, and informal sector impact	Local waste infrastructure influences leakage; uniformity in socio-economic conditions in each district	Simplified linear flows, no tracking between leakage fates, dependent on local conditions and data reliability
Output (Estimation Result)	Results:	Quantities of plastic entering ecosystems (e.g., 710 million metric tons cumulative by 2040)	Estimated 52.1 Mt/year of macroplastic emissions, with 57% from open burning and 43% from debris	Quantifies plastic leakage per item type and ranks pathways for targeted interventions	Quantified plastic leakage amounts, fates of unmanaged plastic (e.g., tons/year)
	Temporal Scale	Global, split into eight geographic archetypes	Global, with sub-national granularity	District to global	Municipal/City Level
	Accuracy, Validation	Uncertainty analysis via Monte Carlo simulations; sensitivity analysis included	Validation through comparison with existing estimates and extensive sensitivity analysis	Higher accuracy with additional data collection; scenarios modeled to reflect interventions	reliability assessed using a traffic light system; higher accuracy with primary data
Analysis of Practicality and Versatility	Granularity	Differentiates macroplastic types and microplastic sources, adapted for regional income levels	Differentiates rigid and flexible plastics, adaptable to various socioeconomic contexts	Allows item-based detail, adaptable to district or city scales for local specificity	Simplified yet effective for rapid assessment; adaptable to data availability and local contexts
	Data Collection	Relies on global datasets, subject to data availability and regional variability	Incorporates multiple global and national datasets, but faces challenges in rural and informal data accuracy	Can use default values or require on-ground data collection for higher reliability	Supports using existing data but emphasizes primary data collection for higher accuracy
	Impact Verification	Models source, pathway, and hotspot impacts; scenarios tested for intervention effectiveness	Effective in identifying emission hotspots and assessing intervention impacts at municipal levels	Identifies sources, pathways, and hotspots for evidence-based solutions	Identifies hotspots and potential interventions; useful for scenario analysis and SDG alignment
Additional Notes	Strengths	Comprehensive global model, accounts for economic and social variability	High-resolution data at municipal scale, robust handling of uncertainties	Comprehensive toolkit, scenario modeling for effective planning	User-friendly, aligns with SDG indicators, rapid assessment with low data requirements
	Weaknesses	High data uncertainty, limited granularity in informal sector dynamics	Limited applicability for regions with insufficient data, exclusion of microplastics and plastic exports	Requires investment in data collection for higher accuracy	Limited to macroplastic, data-dependent, does not track microplastic or inter-stage plastic movement
	Tool kit	No	No	No	Yes

2-3. Elaboration of the Classification Table for Plastic Leakage Estimation Method (F: Coastal and other Land Field Measurements)

Name		CSIRO and COBSEA, 2024, Regional Assessment on Marine Litter in the East Asian Seas
Category		F1. Field Survey (Land), F2. Field Survey (Waterways)
Scale of Estimation	Spatial	Local (Cambodia coast, inland, and river sites)
	Temporal	Short-term (based on data from two collection periods in 2022 and 2023)
Input Data (Variables)	Data Types	Site accessibility, cleanliness, population density, infrastructure, distance to roads, rivers, and coastlines
	Sources	On-site observations, GIS data, socioeconomic datasets
	Plastic Granularity	Macro-based (whole and fragmented items)
	Sector Details	Consumer products, specifically single-use items (water bottles, polystyrene foam)
Estimation Approach	Type of Model	General Additive Model (GAM)
	Description	GAMs assess debris density relationships with socioeconomic and geographic factors across site types
	Leakage Pathways	Land to coast, inland to river, river to ocean
	Assumptions	Consistent distribution patterns and influence of local socioeconomic variables
Output (Estimation Result)	Results:	Estimated density of 70,000 items per km of coastline, equating to over 80 million items on Cambodian coastlines
	Temporal Scale	Spatial Scale: Local (Cambodia)
	Accuracy, Validation	Site data compared across socioeconomic and geographic parameters, noting influence by nearby infrastructure and river distance
Details of applied model	Name of the model	General Additive Model (GAM)
	Description of the model, equation,	Utilizes socioeconomic predictors (population, lighting, distance to coast/river) to determine debris density
	Calibration	Based on socioeconomic variables
	Verification	AIC criterion used to verify model accuracy
Analysis of Practicality and Versatility	Granularity	Model tracks macroplastics and differentiates whole vs. fragmented items
	Data Collection	Requires in-person site surveys with GIS support, socioeconomic databases
	Impact Verification	Model provides data on debris distribution and source pathways, aiding hotspot identification
Additional Notes	Strengths	Robust method providing high-resolution insights into plastic pollution hotspots
	Weaknesses	Limited applicability beyond local scale; may not account for seasonal variations due to short-term data collection periods
	Tool kit	Monitoring guides and apps, etc. (Estimation method tools are not explicitly specified.)



2-4. Elaboration of the Classification Table for Plastic Leakage Estimation Method (F: Aquatic (mainly Riverine) Field Measurements (1/2))				
Name		Schmidt et al., 2017 Export of plastic debris by rivers into the sea	Lebreton et al., 2017 River plastic emissions to the world's oceans	Mellink et al., 2022, The plastic pathfinder
Category		M2. Waste Management, F2. Field Survey (Waterways):	M2. Waste Management, F2. Field Survey (Waterways):	M2. Waste Management, F1. Field Survey (Land)
Scale of Estimation	Spatial	Global, Regional	Global, with focus on large river basins	Regional to Local (terrestrial environments and river basins)
	Temporal	Continuous	Continuous, with emphasis on seasonal variability	Daily with potential for annual projections
Input Data (Variables)	Data Types	Population density, waste generation rates, mismanaged plastic waste, river discharge	Population density, waste generation rates, mismanaged plastic waste, river discharge, catchment characteristics	Wind speed, surface runoff, land use, terrain slope, MPW generation
	Sources	Statistics, remote sensing data, river network, HydroSHEDS, GPWv3	National statistics, remote sensing data, river network databases	Global databases (e.g., HydroSHEDS, ESA CCI)
	Plastic Granularity	Microplastic (<5mm) and Macroplastic (>5mm)	Macro- and microplastic	Macroplastic (>0.5 cm)
	Sector Details	N.A.	N.A.	N.A.
Estimation Approach	Type of Model	Power-law regression model, Linear regression	Empirical model with MPW data, hydrology, and population density	Numerical, spatiotemporal model
	Description	Used MPW data as a predictor for estimating plastic loads transported by rivers to the sea	Estimates plastic emissions by correlating MPW and runoff data to observed riverine plastic concentrations	Calculates plastic transport based on wind and runoff overcoming terrain friction thresholds
	Leakage Pathways	MPW transported from land to rivers, then to oceans	Plastic moves from inland areas to rivers and then to oceans	Plastic moves overland toward river basins based on topography
	Assumptions	High uncertainties due to limited and heterogeneous observational data	Artificial barriers (e.g., dams) act as sinks for plastic, influencing transport	Uniform thresholds for macroplastic mobilization; assumptions about runoff based on rainfall coefficients
Output (Estimation Result)	Results:	Estimated global plastic debris inputs to the sea range between 0.41 to 4 million tons per year	Global riverine plastic input to oceans estimated at 1.15 to 2.41 million tonnes per year, with over 80% from Asia	Generates potential plastic routing maps and accumulation hotspots; shows MPW distribution over time
	Temporal Scale	Global	Global, with local and regional inputs	Local river basins, adaptable to regional
	Accuracy, Validation	High uncertainties acknowledged, with validation through comparison to other studies such as Lebreton et al. ,2017	Model calibrated with field data from selected global rivers, but high uncertainties remain	Requires empirical validation; calibration based on topography and expert inputs
Details of applied model	Name of the model	Power-law model	Empirical regression model	Plastic Pathfinder
	Description of the model, equation,	Log-log linear regression between mismanaged plastic waste (MPW) and plastic load	Regression model based on MPW and runoff data for prediction	Simulates plastic movement based on wind, runoff, and terrain friction thresholds
	Calibration	Performed using 5000 bootstrap samples to account for uncertainties	Calibrated with 30 records from 13 rivers globally	Designed to calibrate based on land cover and terrain slope data
	Verification	Through comparisons with independent estimates from other research (e.g., Lebreton et al.)	Validated against observed data from selected river systems	Requires empirical studies; threshold values based on expert assumptions
Analysis of Practicality and Versatility	Granularity	Micro/macro-plastic; affect by transport efficiency, with microplastics showing higher transport efficiency in rivers with high MPW	Addresses both micro- and macroplastic transport in rivers	Macroplastic, accounting for physical transport influenced by terrain
	Data Collection	Limited availability and accessibility of required data for plastic concentrations in rivers	Limited data for plastic concentrations in specific rivers; uses available field data for calibration	Requires accessible land use and topographic data; some limitations in high-resolution terrain mapping
	Impact Verification	Ability to verify high-load rivers' contribution to ocean plastic pollution enables targeted mitigation efforts	Model effectively identifies high-leakage rivers, aiding targeted intervention	Aims to identify hotspots for plastic accumulation and entry points to rivers
Additional Notes	Strengths	Major rivers contributing significantly to ocean, targeted interventions; combines multiple data sources for comprehensive analysis	Major rivers contributing significantly to marine plastic pollution; seasonal variability provides insights for intervention timing	First terrestrial plastic transport model to include detailed threshold-based mobilization approach
	Weaknesses	High uncertainties due to limited temporal and spatial data, limited applicability in regions with sparse data coverage, reliance on MPW data without considering other potential sources	High uncertainties in estimations due to limited data, especially in under-monitored regions	Lacks empirical calibration for all terrain types and specific plastic types; currently generalized thresholds for various terrains

2-4. Elaboration of the Classification Table for Plastic Leakage Estimation Method (F: Aquatic (mainly Riverine) Field Measurements (2/2))					
Name		Van Emmerik et al., 2018, A Methodology to Characterize Riverine Macroplastic Emission Into the Ocean	Van Emmerik et al., 2019, Seasonality of Riverine Macroplastic Transport	Strokal et al., 2023, River Export of Macro- and Microplastics to Seas by Sources Worldwide	Schreyers et al., 2024, River Plastic Transport and Storage Budget
Category		M2. WM, F2. Field Survey (Waterways)	M2. WM, F2. Field Survey (Waterways)	M2. WM, F2. Field Survey (Waterways)	M2. WM, F1/2. Field Survey (Land/Waterways)
Scale of Estimation	Spatial	Local (Saigon River, Vietnam)	Spatial: Local (Saigon River, Vietnam)	Global	Regional (focused on the Rhine River sections)
	Temporal	Short-term with extrapolation to daily, monthly, annually	Continuous monthly, seasonally, and annually	Annual	Long-term
Input Data (Variables)	Data Types	Plastic piece count, mass, size, type; hydrological data	Plastic item count, plastic mass, hydrological data	Population density, MPW, sewage system data	Plastic transport, storage/river compartment, flow velocity
	Sources	Field observations, visual counting, static bridge-mounted trawls	Visual counts, trawling for samples, rainfall and discharge records	National statistics, geospatial data, modeled parameters for retention rates in rivers	Field measurements, literature (e.g., van Emmerik et al., 2022a)
	Plastic Granularity	Macroplastic (>5 cm)	Primarily macroplastic (>5 cm) with polymer-specific observations (e.g., PS-E, POsoft)	Macro/Micro	Macroplastics (>2.5 cm)
	Sector Details	N.A.	N.A.	Waste management, sewage treatment	N.A.
	Type of Model	Empirical field-based model combined with hydrology	Empirical field observation model	Statistical Model, Process-based Modeling	Conceptual model based on transport and storage budget
Estimation Approach	Description	Count plastic debris, determines cross-sectional flux profile, and calculates plastic mass flux	Combines visual and trawl data to estimate seasonal flux and spatial distribution of plastic	Socio-economic data & river retention rate for plastic export, Integrate macro/micro pathways/pollution source	Quantifies plastic storage and transport in river compartments including surface, suspended, floodplain
	Leakage Pathways	Plastic emissions estimated from rivers to ocean	Plastic emissions from river transport to coastal waters	Diffuse (mismanaged waste), point (sewage systems)	From land to river, floodplain to river
	Assumptions	Rely on tidal patterns and discharge data for extrapolation	Consistent tidal impacts on plastic distribution, seasonal variance and organic content effects	Assumptions: Differing regional treatment rates and river retention parameters	Steady-state, simplified storage across compartments
Output (Estimation Result)	Results:	0.18-0.33 tons/day and 7.5-13.7 kilotons annually for the Saigon River	Estimated plastic outflow of $1.1 \times 10^3$ to $1.6 \times 10^3$ tons annually from the Saigon River	Approx. 0.5 million tons of plastic exported to seas annually	Quantifies plastic mass storage in floodplains (98%) and transport across river sections
	Temporal	Local (Saigon River)	Local (Saigon River)	Global with regional details	Regional, annual average
	Accuracy, Validation	High variability due to tidal influences and limited temporal coverage	Limitation by specific sampling location and conditions; Variability of tidal patterns & organic accumulation	Compared with other models (Lebreton, Jambeck) and validated against individual river measurements	Comparison with observed plastic transport in field studies
Details of applied model	Name of the model	Empirical macroplastic emission model	Empirical macroplastic transport model	MARINA-Plastics Model (Over 10,000 sub-basins worldwide including SE Asia)	Conceptual River Plastic Budget Model
	Description of the model	Estimates plastic flux by observed cross-sectional plastic profiles, mass per piece, and river discharge	Derives plastic flux using item counts and mass, accounting for cross-sectional and vertical distribution	Accounts for both diffuse and point sources of plastic pollution.	Assesses plastic flux and retention using input parameters from field measurements
	Calibration	Calibrated with daily plastic counts and discharge estimates	Calibrated based on visual counts and trawl samples	Based on field data from specific rivers	Based on observational data for the Rhine
	Verification	Comparison with previous estimates and observations in similar settings	Comparisons with existing estimates for similar river systems	Comparison with existing models for macro- and microplastic pollution	Compared with literature values for riverine plastic dynamics
Analysis of Practicality and Versatility	Granularity	Focus on macro emissions, with an emphasis on spatial variability across river width	Includes size and polymer-specific granularity; variability in plastic transport by MSW patterns	Differentiates between macro- and microplastics	Focused on macroplastics, adaptable for other size classes
	Data Collection	Field measurements adaptable to available resources (e.g., visual counts, trawling)	Visual counts and trawling are practical but sensitive to river dynamics	Extensive use of available global data, assumptions for regions lacking direct data	Based on field observations, challenges include lack of consistent floodplain measurements
	Impact Verification	Effective in estimating emissions in rivers influenced by urban waste	Seasonal insights aid in identifying peak transport periods for targeted interventions	Enables source-specific policy recommendations for diffuse and point sources	Limited by assumptions on retention and exchange rates
Additional Notes	Strengths	Adaptable for various river systems; includes short-term, high-frequency sampling	Seasonal variability; provides detailed composition and distribution of riverine plastic	Sub-basin-level analysis, differentiates between point and diffuse sources	Retention estimates for floodplain plastics, importance of floodplains in plastic storage
	Weaknesses	Limited local hydrodynamic influences and seasonal variations; required for year-round estimation	Limited temporal and spatial coverage; dependent on specific sampling points & influenced by high river traffic	Model uncertainties due to variability in data sources and assumptions across regions	Limited short-term temporal data, assumption on floodplain exchange & transport to downstream boundary
	Tool kit	No	No	No	No



### 3. Elaborated Classification of Existing Estimation Methods

Main Input Data				Existing method		Life cycle*					Data Source, Model, National Statistic, Leakage rate	Stakeholder / Measure	
						P	C	W	E	O		Stakeholder	Indicator
1	M1	Production & Consumption (total plastic, sector, polymer)		<ul style="list-style-type: none"><li>Nakatani et al., 2020</li><li>UNITAR, 2023 (Toolkit for the Product-Lifespan Method)</li><li>OECD, 2020 (ENV-Linkages Modelling Framework)</li></ul>							ENVI/CGE, Input-Output Table, Physical Supply and Use Tables	Ministry of Industry, etc.	Reduction, Substitution, SUP reduction/phase-out, green procurement, etc.
		Production & Consumption (item)	MacP	<ul style="list-style-type: none"><li>MOEJ, 2024 (Lost fishing gear estimation)</li></ul>							National Statistic, Discharge rate, etc.		
				<ul style="list-style-type: none"><li>MOEJ, 2024 (PET bottle estimation)</li></ul>							National Statistic, Discharge rate, etc.		
			MicP	<ul style="list-style-type: none"><li>ECHA, 2020</li><li>ICF&amp;Eunomia, 2018</li><li>UNEP, 2018</li><li>MOEJ, 2024 (Buildup estimation by sources and items)</li></ul>							National Statistic, Discharge rate, etc.		
2	M2	Waste Management	Waste Flow	<ul style="list-style-type: none"><li>GIZ, 2020 (Waste Flow Diagram (WFD))</li></ul>							National Statistic, etc.	Ministry of Environment (MOE), Ministry of Interior (MOI), Local Government (LGs), etc.	3Rs, improved waste collection, EPR, various recycling schemes, introduction of WtE, improved landfill sites, etc.
				<ul style="list-style-type: none"><li>Basel Convention, 2022 (Toolkit for Material Flow Analysis Method)</li><li>University of Leeds, 2019 (ISWA Plastic Pollution Calculator (PPP))</li></ul>							National Statistic, Discharge rate, etc.		
				<ul style="list-style-type: none"><li>GPAP, 2022 (National Analysis and Modelling (NAM) Tool)</li><li>Cottom et al., 2024 (SPOT Model)</li></ul>							Global, National Statistic, Statistic model, etc.		
3	F	Field survey and Monitoring	Coastal and other Land Field Measurements	<ul style="list-style-type: none"><li>CSIRO and COBSEA, 2024 (Regional Assessment on Marine Litter in the East Asian Seas)</li></ul>							Survey, Statistic model	MOE, LGs	Mitigation measures (Collection, Awareness-raising, etc.), Impact assessment of the mitigation measures
				<ul style="list-style-type: none"><li>MOEJ, 2024 (Beach litter survey)</li><li>S-19 (Environment Research and Technology Development Fund)</li></ul>							Survey, extrapolation		
			Aquatic (mainly Riverine) Field Measurements	<ul style="list-style-type: none"><li>Nihei et al., 2020    Schmidt et al., 2017    S-19</li><li>van Emmerik et al., 2019    Lebreton et al., 2017</li><li>Meijer et al., 2021    World Bank et al., 2021</li><li>Mellink et al., 2022</li><li>MOEJ,2024 (Pumping station survey)</li></ul>							Survey, Statistic model	MOE, Water & Marine related departments, LGs	
											Survey, Discharge rate, Statistic model		

(Reference: The estimation methods used by OECD (ENV-Linkages model + UOL + DTU) and UNEP (National Guidance on Plastic Pollution Hotspotting and Shaping Action) integrate data from both M1 and M2.)

\* Blue: Data Input, Yellow: Output, P: Production、C: Consumption、W: Waste Generation/Management、E: Leakage to the environment、O: Leakage to the ocean

## 4. Consideration of the Algorithm for Selecting Practical Estimation Method

**Three steps** for selecting practical estimation method based on the elaborated classification table (previous page)

- Identify the applicable estimation method and corresponding existing methods (yellow & red boxes in the figure) based on the intended purpose
- Classify the selected estimation method according to the number of steps involved in the leakage pathway to calculate the target output.
- The concept of this document (illustrated on the next page) is to classify the level (① – ③ ) based on the number of steps required to calculate the output data. (e.g. IPCC GHG inventory guidelines adapt the “Tier” approach to classify methods based on the level of accuracy, reflecting the extent to which country-specific data and national circumstances are incorporated.



### Examples of Policy & Interventions

- **Production and Consumption Measure**  
Reduction in plastic use, Alternatives, Reduction or phase-out of SUPs, Green procurement policies , Measures for lost and abandoned fishing gear, etc.
- **Improvement of Waste Management**  
3Rs (Reduce, Reuse, Recycle) Enhance waste collection system, EPR, Various recycling schemes, Waste-to-Energy, Landfill O&M, etc.
- **Impact on Terrestrial, Riverine, and Marine Areas**  
Cleanup, Awareness-raising, Countermeasure development for plastic leakage, Impact assessment after the implementation of policies/interventions, etc.

### Examples of Targeted Indicator

- **Waste Management**
  - ✓ Waste Generation Amount
  - ✓ Waste Flow
- **Leakage to Environment**
  - ✓ Mismanaged Plastic Waste Amount/Density
- **Leakage to Ocean**
  - ✓ Plastic leakage into the ocean via rivers and coastal areas

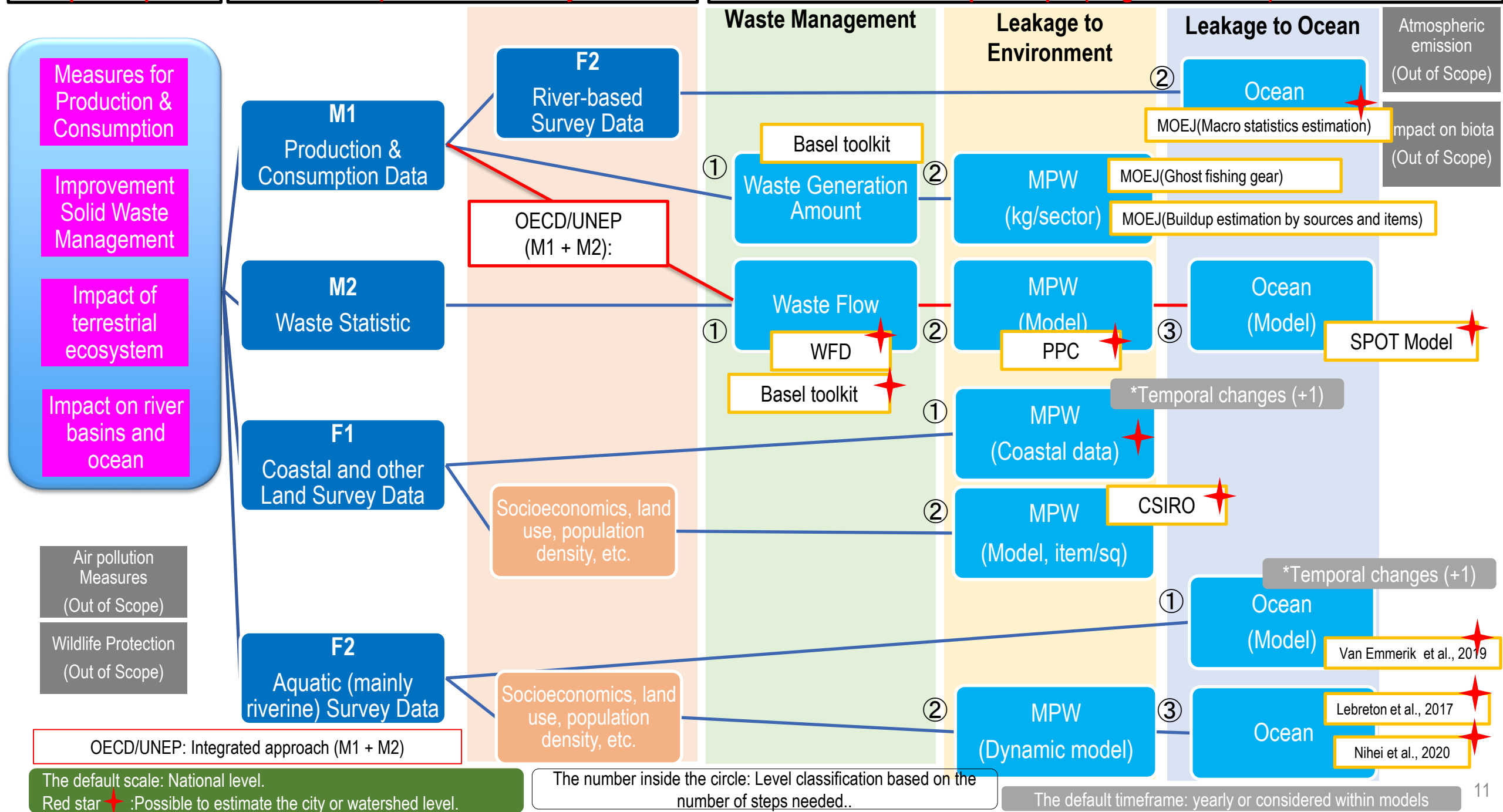
### Examples of Data Types

- **M1: Production & Consumption Data**  
Production, import, and consumption amounts, Trade statistics, etc.
- **M2: Waste Statistic**  
Waste amount, Self-managed waste amount, Waste collection rate, Recycling rate Landfill conditions Government statistics, Global data, etc.
- **F: Coastal and other Land Data**  
Illegal dump, Waste found in riverbanks and coastal areas, Population density, Land use, etc.
- **F: Aquatic (mainly riverine) Data**  
Measured data from river basins, Model analysis, etc.

## Step 1. Purpose

## Step 3. Data Availability

## Step 2. Output (Targeted Indicator)



## 5-1. In-Depth Hearings in the in the Southeast Asian Region (Indonesia)

Criteria	Summary																													
Estimation of Leakage Amount and Monitoring-Related Activities	<ul style="list-style-type: none"><li><b>The National Coordinating Team for Marine Debris Management (TKN PSL)</b> estimates “Land-based” and “Sea-based” sources by reviewing the National Action Plan (NAP), defining the total of these estimates as “Marine Debris Leakage”.</li><li><b>Estimation method for land-based leakage:</b> MPW: Derived from SIPSN and Ministry of Health survey data, Impact of plastic regulations (such as EPR and bans on single-use plastics)</li><li><b>Estimation method for sea-based leakage:</b> Fisheries waste (vessel numbers, operating day), Passenger vessel waste (passenger capacity, operational hours, etc.), Illegal marine dump (Self-reported, monitoring data)</li><li><b>Upstream data (production and consumption)</b> is not included in the estimation.</li></ul>	<table border="1"><caption>Estimated Indonesian Marine Debris Leakage 2018-2023</caption><thead><tr><th>Year</th><th>Land-based Plastic Waste (ton/year)</th><th>Sea-based Plastic Waste (ton/year)</th><th>Total (ton/year)</th></tr></thead><tbody><tr><td>2018</td><td>508,183</td><td>77,492</td><td>615,675</td></tr><tr><td>2019</td><td>516,327</td><td>47,718</td><td>566,075</td></tr><tr><td>2020</td><td>508,716</td><td>12,785</td><td>521,541</td></tr><tr><td>2021</td><td>407,118</td><td>32,889</td><td>440,107</td></tr><tr><td>2022</td><td>309,616</td><td>88,379</td><td>398,001</td></tr><tr><td>2023</td><td>329,123</td><td>19,938</td><td>359,061</td></tr></tbody></table>	Year	Land-based Plastic Waste (ton/year)	Sea-based Plastic Waste (ton/year)	Total (ton/year)	2018	508,183	77,492	615,675	2019	516,327	47,718	566,075	2020	508,716	12,785	521,541	2021	407,118	32,889	440,107	2022	309,616	88,379	398,001	2023	329,123	19,938	359,061
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Methodologies and Tools Used																														
Challenges in Data Collection, Analysis, and Management	<ul style="list-style-type: none"><li><b>TKN PSL:</b> collecting data from relevant ministries and conducts estimations.</li><li><b>Progress toward the 70% reduction target (by 2025):</b> 42% achieved as of 2023.</li><li><b>Data Gaps:</b> Lack of digital integration (No real-time plastic leakage monitoring system), Difficulty in short-term variation analysis (Seasonal fluctuations and short-term trend), Insufficient data (Offshore industries, ports, and aquaculture sectors), Limited onsite measurements (Heavy reliance on statistical data, Less measurement).</li><li><b>Policy Sustainability Challenges:</b> Uncertainty (Under the new administration), Post-2025 revision of the NAP</li><li><b>TKN PSL’s Role:</b> Stipulated by Presidential Regulation No. 83/2018, Funds from UNDP</li></ul>	<table border="1"><thead><tr><th>Source</th><th>Year</th><th>Leakage (million tones)</th></tr></thead><tbody><tr><td>BRIN/LIPI</td><td>2018</td><td>0.270 - 0.590</td></tr><tr><td>World Bank</td><td>2018</td><td>0.201 - 0.553</td></tr><tr><td>TKN PSL</td><td>2020</td><td>0.615</td></tr><tr><td>NPAP</td><td>2020</td><td>0.620</td></tr></tbody></table>	Source	Year	Leakage (million tones)	BRIN/LIPI	2018	0.270 - 0.590	World Bank	2018	0.201 - 0.553	TKN PSL	2020	0.615	NPAP	2020	0.620													
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Required Support and Resources	<ul style="list-style-type: none"><li><b>Data Management &amp; Integration:</b> Enhance real-time monitoring systems, Improve the national data platform (Integrating data from LGs and the private sector).</li><li><b>Technology Development:</b> Establish AI-based data analysis infrastructure, Introduce new measurement technologies (Drones, sensors, etc.).</li><li><b>Policy &amp; Funding:</b> Maintain a long-term policy framework, Promote private sector participation, Expand international funding and technical cooperation.</li></ul>																													
Existing Policies, Programs, and Guidelines	<ul style="list-style-type: none"><li><b>National Action Plan (NAP):</b> Presidential Regulation No. 83/2018, Target: 70% reduction by 2025, 5 strategies and 59 activities by 22 ministries and agencies.</li><li><b>Key Measure:</b> Extended Producer Responsibility, Ban on Single-Use Plastics, (Regulatory strengthening contributes to leakage reduction).</li><li><b>Infrastructure Development:</b> 850 waste management facilities in 200 cities by Ministry of Public Works</li></ul>																													
Key Outcomes	<ul style="list-style-type: none"><li><b>Reduction in Plastic Leakage:</b> Reduced from 615,000 tons in 2018 to 359,000 tons in 2023 (a 41% decrease)</li><li><b>Improvement in Data Collection &amp; Reporting Systems:</b> Establish a national reporting platform enables real-time progress tracking across ministries</li><li><b>Advancements in Policy &amp; Legal Framework:</b> Introduce plastic reduction policies, including EPR and bans on SUPs</li></ul>																													
Stakeholders	<ul style="list-style-type: none"><li><b>Ministry of Environment (MOE):</b> SIPSN (266 verified &lt; 441 registered &lt; among 514 cities), SIMBA</li><li><b>Ministry of Industry (MOI):</b> National database on plastic production (Based on voluntary reporting by businesses and surveys)</li><li><b>Ministry of Public Works (MPW):</b> SI-INSAN (Infrastructure database)</li><li><b>Ministry of Health (MOH):</b> Household sanitation survey (Conducted every 5 years nationwide in Indonesia)</li><li><b>Ministry of Home Affairs (MOHA):</b> SIPD (Statistical data from local governments)</li></ul>																													



## 5-2. In-Depth Hearings in the Southeast Asian Region (Vietnam)

- **Challenges in Addressing Plastic Pollution and Marine Debris**
  - ✓ Lack of dedicated policies on plastic pollution and marine debris
  - ✓ Poor coordination among relevant agencies (e.g. PCD: land, rivers, and waterways, VASI/VEMSI: estuaries, coastal areas, and the ocean, and no harmonized approach)
  - ✓ Disproportionate funding allocation (Majority to VASI, rather than PCD)
- **Key Reports & Data Collection Efforts**
  - ✓ UNDP/GEF's initial plastic pollution data collection (including production and consumption)
  - ✓ 2022 Status Report on Plastic Waste
- **Monitoring & Budgetary Constraints**
  - ✓ No official monitoring activities in place
  - ✓ Rely heavily on donor funding (No budget)
  - ✓ No national standards or methodologies
- **Planned Actions & Data Limitations:**
  - ✓ VEA Plan: Plan to formulate a plastic leakage inventory and estimation method
  - ✓ Data limitations: Primarily SWM data, limited production and consumption data collection (VPA is the main source of information)

Criteria	Summary
Estimation of Leakage Amount and Monitoring-Related Activities	• VASI's Marine Pollution Report (WWF State Report, 2023, MCD's Waste Composition Survey by communities and interceptors (The Ocean Cleanup classification).VNU's State Report (Macroplastic, waste flow analysis) and microplastic research, UNDP/GEF's baseline survey (initiated from 2025)
Methodologies and Tools Used	• Macroplastic Monitoring (Field surveys, interceptor systems, GIS-based hotspot mapping, Waste management data), Microplastic Analysis (using FTIR microscopy, USEPA sampling protocols), Data Sources (Local government reports, Customs data, Private sector contributions)
Challenges in Data Collection, Analysis, and Management	• Data Gaps (Lack of baseline data and polymer-specific data), Coordination Challenges (Insufficient collaboration among relevant agencies), Infrastructure Barriers (Limited availability of advanced tools, Inconsistent monitoring protocols)
Required Support and Resources	• Capacity Development (Training programs for government officials and local communities), Policy & Guidelines (National standards for plastic and microplastic monitoring), Financial & Technical Support (Funding for large-scale surveys and advanced monitoring tools), International Cooperation (Sharing best practices and collaborative initiatives)
Existing Policies, Programs, and Guidelines	• NAP for Marine Plastic Waste Management 2019–2025 (Focuses on integration across sectors), Environmental Protection Law 2022 (Includes waste management regulations but does not specifically address plastic leakage/pollution)
Key Outcomes	• Research on Baseline Data for Plastic Waste Sources (Deficiencies in waste collection systems, Identification of data gaps and the need for standardization), Scalable Model (MCD's initiative in Ha Long Bay as a potential expansion model)
Stakeholders	• Government/NGO Collaboration (Joint initiative by VASI, PCD, and WWF like State Report), Academic Institutions (research collaboration with int. organizations), Community Engagement (Waste monitoring activities by local communities)

# Classification of methods based on the level of accuracy in the input data (harmonization between macro and field data )

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Input		Output		
Tier 1 (Basic Form)		Waste Generation Amount	Leakage (Terrestrial)	Leakage (Aquatic)
M1	Production & Consumption	<ul style="list-style-type: none"> <li>• Nakatani et al., 2020</li> <li>• UNITAR Toolkit</li> </ul>	(M1 x Leakage rate)	M1 x Leakage rate (MOEJ Macro Statistics estimation)
M2	Waste Management	<ul style="list-style-type: none"> <li>• Basel Toolkit</li> <li>• WFD</li> <li>• SPOT Model</li> </ul>		<ul style="list-style-type: none"> <li>• Lebreton et al., 2017</li> <li>• Meijer et al., 2021</li> <li>• World Bank et al., 2021</li> <li>• Mellink et al., 2022 (based MPW)</li> </ul>
F1	Field Data (land)		Accumulation of Field Data, Extrapolation	Leakage rate, Model, Simulation
F2	Field survey (waterways)	—	—	Accumulation of Field Data, Extrapolation (MOEJ Pumping station survey)
Input		Output		
Tier 2 (Applied Form)		Waste Generation Amount	Leakage (Terrestrial)	Leakage (Aquatic)
M1 -> M2 (Harmonized)		• UNEP National Guidance	• OECD (ENV-Linkages Modelling Framework + UOL + DTU)	
F1 -> F2 (Harmonized)		—	x	x
M2 <-> F1 (Verification)		x	x	x
M2 <-> F2 (Verification)		x	x	x
Input		Output		
Tier 3 (Advanced Form)		Production, Consumption & WM	Leakage (Terrestrial)	Leakage (Aquatic)
M1 -> M2 -> F1 -> F2		x	x	x



# **Results of a Survey on the Feasibility of Implementing Estimation Methods Using Macro Statistical Data Overseas (FY 2024)**

# Estimation by using macro-statistical data

In theory, there are two methods using macro statistical data:

1. Estimating the outflow volume by subtracting macro statistical data such as disposal amounts from macro statistical data such as usage amounts.

2. Estimating the outflow volume by multiplying macro statistical data such as usage amounts by parameter/coefficients (e.g., outflow rates).

Method (1) is calculating by using macro statistical data obtained for different purposes (which may contain errors), raising concerns about high uncertainty. Therefore, Method (2) was chosen for consideration.

- The calculation of the discharge amount using the macro-statistical data **is done by multiplication**, not subtraction.

$$A \text{ (Macro-statistical data)} \times B \text{ (Discharge rate)} = C \text{ (Discharge amount to the env./ocean)}$$

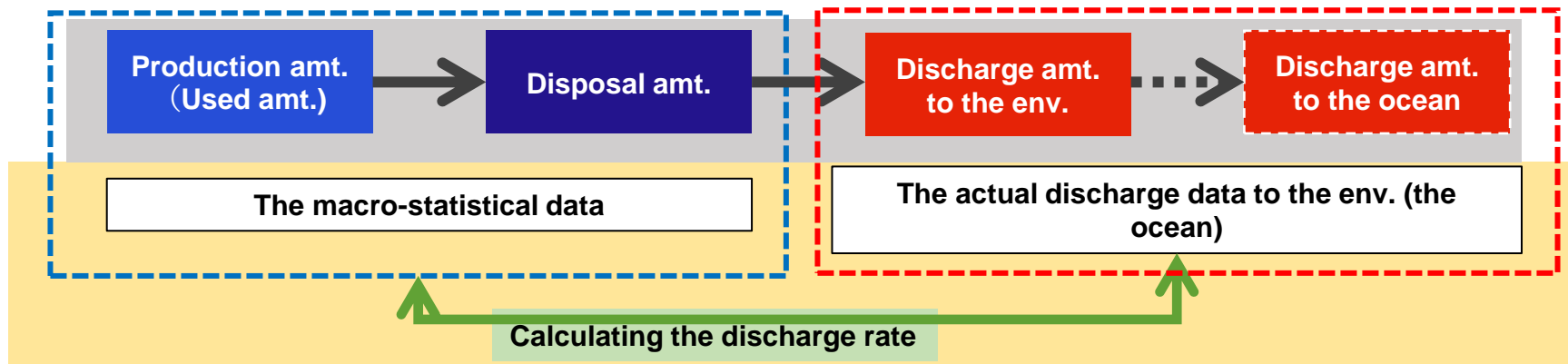
- For “Calculation of **discharge rate**” and “Consideration of estimation methods by (the macro-statistical data  $\times$  **discharge rate**)”, item (product) whose “macro-statistical data” and “data of the actual discharge amount to the environment” are considered relatively easy to obtain should be selected
- Trial using Japanese data in FY2023, and other developing countries data in F’2024

(Example of Japan’s case in FY2023)

【Item (product) example】 PET bottles

【Candidates for obtaining the macro-statistical data】 Data from the Council for PET Bottle Recycling, results of input-output analysis






【Actual data on the amount discharged to the environment】 Results of a survey of the amount of waste collected at a pump station (using results from a separate operation).



# List of results of estimation methods using macro statistical data in foreign countries.

(3) is calculated by dividing (2), which was calculated using research papers from various countries, by statistical data (1).  
(5) is calculated by multiplying (3) by statistical data (4), which was obtained separately.

The following are the results of applying the estimation method using macro-statistical data to other countries, and are “Marine leakage rate of PET bottles [%]”, “Marine leakage volume of PET bottles [t/y]”, and “Estimated marine leakage volume of plastic products [t/y]”.

country	(1) PET Bottles Sales volume [t/y]	(2)Marine leakage volume of PET bottles [t/y]	(3)=(2)/(1) Marine leakage rate of PET bottles [%]	(4) Plastic Products Amount used [t/y]	(5)=(3) × (4) Estimated marine leakage volume of plastic products[t/y]	Amount of plastic leaked into the ocean as reported by countries[t/y]	(Reference) Jambeck (2015). Marine Leakage [t/y]	(Reference) Meijer (2021). Marine Leakage [t/y]
 Indonesia	298,113	2,004~4,008	0.67~1.34	4,206,000	28,000~56,000 (28,180~56,360)	359,061	480,000~1,290,000	56,000
 Vietnam	203,774	3,755~6,259	1.84~3.07	2,930,000	54,000~90,000 (53,912~89,951)	-	280,000~730,000	28,000
 Thailand	184,906	5,560	3.01	6,070,000	183,000 (182,707)	-	150,000~410,000	23,000
 Philippines	147,170	2,414	1.64	2,150,000	35,000 (35,260)	-	280,000~750,000	360,000
 Japan	583,000	629	0.108	4,558,433~8,950,000	4,900~9,700 (4,923~9,666)	11,000~27,000	21,000~57,000	1,835

	Bibliography related to (1)	Bibliography related to (2)	Bibliography related to (4)	Reference material on the amount of plastic waste leaking into the sea reported by each country.	Differences from estimates in Japan	Points to note
Indonesia	GA Circular, 2019, Full Circle: Accelerating the Circular Economy for Post- Consumer PET Bottles in Southeast Asia	Hariyadi et al. 2022 Plastic debris in citarum river	Direktorat Industri Kimia Hilir dan Farmasi (2025)	TKN PSL (2025).The Indonesia National Action Plan for Marine Debris Management “Secretariat of The National Coordinating Team for Marine Debris Management	<ul style="list-style-type: none"><li>• Normal data.</li><li>• Targeting of PET bottles flowing on the surface.</li><li>• Measured by wet weight (converted to dry weight by a factor).</li></ul>	<ul style="list-style-type: none"><li>• Using 2021 data.</li><li>• When a river branches, it is necessary to pay attention to the population of the catchment area.</li><li>• The boundaries of the river basins may not correspond to the actual situation.</li></ul>
Vietnam		Nguyen et al., 2024 Assessment and sustainable management strategies for plastic waste in Can Tho City, Vietnam: A circular economy approach	WWF, 2023 Report on plastic waste generation in 2022	-	<ul style="list-style-type: none"><li>• PET bottles stagnating under the bridge are targeted.</li></ul>	<ul style="list-style-type: none"><li>• Using data from 2021 data.</li><li>• The collection period is not specified, so we assume it is one day.</li><li>• Difficult to identify detailed catchment area due to Mekong Delta region.</li></ul>
Thailand		UNEP, IUCN and Life Cycle Initiative, 2020 National guidance for plastic pollution hotspotting and shaping action	Ministry of Natural Resources and Environment, 2021 Action plan on plastic waste management phase I (2020 – 2022)	-	- (No observation information)	<ul style="list-style-type: none"><li>• Calculated based on 2019 data.</li><li>• Amount of leakage into the environment, not the ocean.</li></ul>
Philippine s		Requiron and Bacosa, 2022 Macroplastic transport and deposition in the environs of pulauan river, Dapitan City, Philippines	WWF-Philippines, 2020 EPR scheme assessment for plastic packaging waste in the Philippines	-	<ul style="list-style-type: none"><li>• Normal data.</li><li>• Targeting of PET bottles flowing on the surface.</li><li>• Measured by wet weight (converted to dry weight by a factor).</li></ul>	<ul style="list-style-type: none"><li>• Using 2021 data.</li><li>• When a river branches, it is necessary to pay attention to the population of the catchment area.</li><li>• The boundaries of the river basins may not correspond to the actual situation.</li></ul>
Japan	Ministry of the Environment “FY2023 Study Results: Estimation of the Amount of Marine Plastic Litter in Japan”					

# Evaluate availability of data in other countries and the direction of future surveys.

- As an evaluation of the data utilized in this study, in terms of statistical data, in the **case of PET bottles, there is no established system or mechanism to collect individual information as in Japan, and the availability is generally low.** In the case of plastic products, information is often published by government agencies (especially for Indonesia, there is detailed information by product and polymer in cooperation with industry associations). Surveyed data is literature-based information only, and leakage rates may be underestimated/ overestimated depending on the setting of watershed boundaries, consideration of high water, and the scope of data acquisition.
- In the future, with **Indonesia and Vietnam** in mind, **where detailed statistical data may be available**, we will consider conducting field surveys at **locations where watershed boundaries can be established and where surveys can be conducted under normal conditions and during high water**, and obtaining our own measured data (leakage rates) (we aim to calculate leakage rates for plastic products rather than PET bottles in combination with statistical data).



Availability [Low]				
statistics data	PET bottle	• The only information we were able to confirm was the information on sales volumes indicated in the literature. • There is information that it is unlikely that industry groups in Vietnam and Indonesia are compiling and updating data on the production, sales and consumption of PET bottles through local surveys (there is no established system or mechanism for collecting information on PET bottles).		
	Plastic Products	<b>Availability [High]</b> • Government agencies work with industry groups to release information. • Detailed information available by product and polymer.	<b>Availability [Medium]</b> • Information is available in a report from a government-affiliated research institute. • There is a possibility that the industry group may have information, and we are currently investigating.	<b>Availability [High]</b> • Government agencies release information.
actual measurements data	PET bottle Leakage rate	<b>Availability [Medium]</b> • Problems related to watersheds are small. • Adopted survey results for normal conditions only. • Sampling at river surface (leakage rates may be underestimated). • The number of samplers used is unknown (one or two).	<b>Availability [Medium]</b> • It is difficult to set the boundaries of the watershed around the Mekong Delta (there is a possibility of underestimation). • Data on the garbage collected under the bridge (there is a possibility that it is overestimated because it may include garbage from outside the survey period).	<b>Availability [Medium]</b> • The source of information on which it is based is not the amount of leakage into the river, but the amount of leakage into the environment (the leakage rate may be overestimated). • No observation information.
				<b>Availability [Medium]</b> • It may be difficult to set up a watershed. • Adopted survey results from normal and high water conditions. • Visual observation at river surface (leakage rate may be underestimated).

# Conditions and methods for obtaining the data used in the calculation.

- The amount of plastic bottles littered in each country is an estimate based primarily on the results of academic studies.
- ✓ The sampling methods and conditions for each country's data are as follows. **Terms vary by country.**
  - ✓ It is important to note that, unlike in Japan, data is obtained from the surface.
  - ✓ For Indonesia and the Philippines, where wet weight is used, we selected appropriate wet/dry conversion factors to convert to dry weight, considering whether it would be possible to estimate the weight of PET bottles. The wet and dry conversion factors are taken from the "FY2023 Survey on the Current State of Plastic Waste in Rivers, Lakes and Marshes Ending up in the Ocean" by MOEJ in Japan.

country	Normal/ High water	Sampling method	Data acquisition range (Transverse direction)	Data acquisition range (Vertical direction)	Wet/ Dry weight <sup>※2</sup>	Estimation Method
Indonesia	Normal	Catching in a net	2m or 4m	Surface	Wet <sup>※3</sup>	Days converted into
Vietnam	Normal	Collection of retained garbage under the bridge	3m or 5m	Surface	Dry	Days converted into
Thailand <sup>※1</sup>	-					
Philippines	Normal and High water	Visual observation	Whole area	Surface	Wet <sup>※4</sup>	Days converted into
Japan	Normal and High water	Collected at the river mouth weir and drainage pump station	Whole area	Whole layer	Dry	Days, rainfall conversion

※1 : Organized as the quantity for the entire country of Thailand, with no observational information to serve as a basis.

※2 : The wet weight is converted to the dry weight by multiplying by a coefficient. The coefficient is based on the results of the “FY2023 Survey on the Current State of Plastic Waste in Rivers, Lakes, and Marshes Endangering the Ocean”(The weight of the samples before drying was measured when they were brought back, and then the weight after drying was measured after drying in a dryer (50°C to 100°C).)

※3 : It was not possible to estimate the weight of PET bottles in relation to the total weight of plastic. Therefore, the total wet weight of the plastic was multiplied by the plastic conversion factor of 0.67, and the resulting figure was multiplied by the proportion of PET bottles to estimate the PET bottle leakage rate.

※4 : The amount of PET bottles leaked was estimated by estimating the wet weight of PET bottles from the total weight of the plastic surveyed, and then multiplying by the wet/dry conversion factor of 0.86.

country	elevation	Reference Watershed Boundaries	Rivers used as reference	catchment population
Indonesia Philippines Vietnam	Approx. 30m DEM NASA Earthdata	HydroBASINS HydroSHEDS	HydroRIVERS HydroSHEDS Google map Google	Approx. 5 km × 5km grid SEDAC
Japan	5~10m DEM basic mapping information	Watershed Mesh National land data, river development plans, etc.	river line National Land Numerical Data	3rd to 5th order (250m to 1km) grid e-Stat

(1) Consideration of monitoring methods that can be compared.

In the survey, the amount of plastic product leakage was estimated by calculating the amount of PET bottle leakage in each country from existing literature. As mentioned above, data can vary between countries due to data collection methods, such as whether it's collected during normal or high water, and whether it's visually observed or sampled with a net.

**Visual sampling is considered to be the most suitable** method for conducting simple cross-country comparisons during next year's field survey. On the other hand, in the case of visual sampling, there is a possibility of errors being made by the observer, so if possible, it is preferable to conduct monitoring surveys using equipment such as cameras. In addition, since the visual sampling only targets the surface layer, there is a possibility that the results of the estimated leakage volume will be underestimated. In order to accurately determine the leakage volume of each country, it is **desirable to conduct parallel surveys using "trash removal devices" and "interceptors" that have functions similar to those of Japanese drainage pumping stations.**

(2) Investigation of estimation methods.

In this trial calculation, the amount of PET bottle leakage per person in the target catchment area was calculated and multiplied by the total population to estimate the amount of leakage nationwide. This method may lead to a large error when estimating the entire country from the point of view of representativeness of the target sites and other land uses such as agricultural land.

In addition to this calculation method, it is **desirable to compare and consider other estimation methods**, such as estimating from the area using water balance analysis and estimating by taking into account land use, with reference to Japanese calculation methods. It is also important to check the data that has been organized in each country.

(3) Survey of typical plastic waste that is desirable as a target for tracking in Southeast Asia.

In calculating the leakage rate by product in this year's survey, PET bottles were selected as the tracking target for the following reasons.

- As in Japan, statistical data is considered to be well developed.
- Even if it leaks into the environment, there is little chance that it will be mistaken for another substance, and it is a typical classification item.

On the other hand, this year's field survey confirmed that it is not appropriate to calculate the leakage coefficient using PET bottles in Southeast Asian countries due to the following problems.

- Statistical data on production and sales volumes, etc., are not available.
- PET bottles have a market value after use and are considered to be collected by the informal sector, so they are not representative as an item for estimating leakage rates due to their high recycling rate compared to other plastic products.

In preparation for next year's survey, it is necessary to **select polymers and products that are suitable for verification**, such as polyethylene. Factors to be considered include the **availability of statistical data** (production volume, sales volume, etc.), the ease of **identifying the target substance in the event of a leak**, and the representativeness of the product (for example, the fact that the recovery rate is not as high as for other plastic products such as PET bottles).