# **Tracking NDC mitigation commitments** under the ETF Webinar Series

-22-23-24 July 2024-

## Workshop 3: Reporting on NDC projections under ETF and messing data

July 24, 2024

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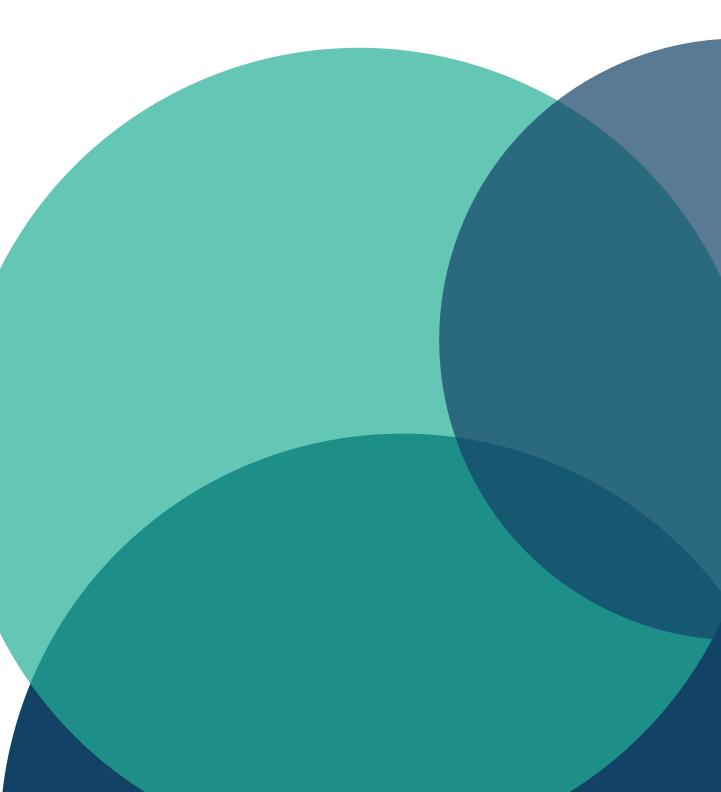
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# Agenda

Time	Session
10 min	Opening and Welcoming Remarks & Group Photo Speaker: Khetsiwe Khumalo, Advisor on Climate Tr UNEP CCC
20 min	Elements on Projections of GHG emissions and re
05 min	Interaction and Q&A
15 min	Inroduction and explanation of Tables 6 to 9: GHC and removals, Projections with WM, WEM, WAM S
15 min	Introduction and explanation of Tables 10, 11, 12
10 min	Interaction and Q&A
20 min	Missing data   What to do in case of missing data
10 min	Interaction and Q&A
10 min	Feedback Mentimeter Survey
ed py: 05 min	Wrap-up and Closing Remarks
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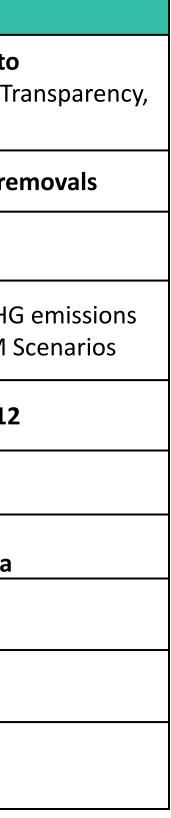
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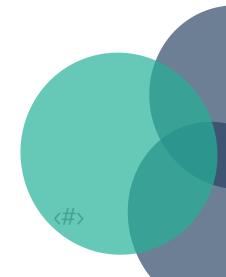












#### Tracking NDC mitigation commitments under the ETF Webinar Series

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Workshop 3: Reporting on NDC projections under ETF and messing data

Missing data | What to do in case of missing data

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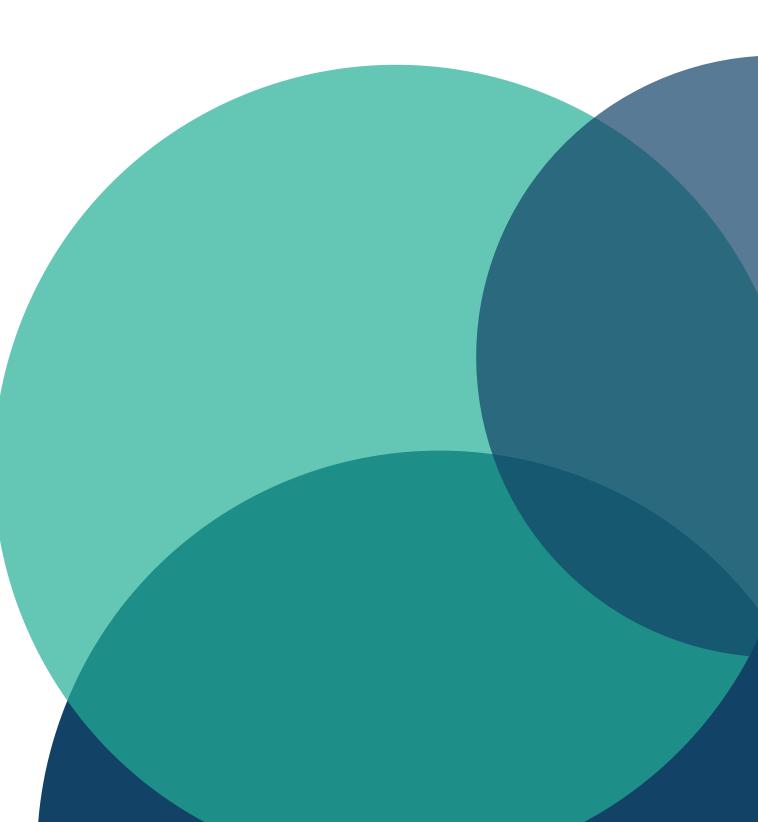
gef

environment programme

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# Situation resulting in data gaps:

**Inconsistency of data** 

**Incomplete Data** 

Changes in activity data, in emission rates

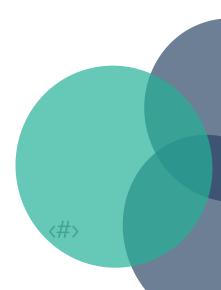
The changes and gaps in data availability

Lack of institutional framework

Lack of ownership by stakeholders (lack of sharing or calculation of data)







## Types of data and gaps

	Input Data Type	Examples of Input Data	Typical Sources of Data	Typical Gaps
Historical Data	- Economic indicators	GDP, unemployment rates, inflation rates	National statistical offices, World Bank, IMF	Inconsistent time series, Missing data for certain years
	- Demographic data	Population size, age distribution, urbanization rates	National census bureaus, UN Department of Economic & Social Affairs	Incomplete datasets, Lack of granular data (regional, age-group)
	- Energy usage trends	Energy consumption by sector, renewable energy usage	National energy agencies, IEA	Insufficient granularity, Under-reported sectors
	- Changes in land use	Land use changes, deforestation rates, urban expansion	National environmental agencies, remote sensing databases	Time-lags in reporting, Uncertainty in measurement techniques
Historical Emissions	- Activity data	Energy production and consumption, industrial activities	National environmental agencies, industry reports	Lack of source-specific data, Inconsistent methodologies
	- Emission factors	GHG emission factors for various sectors and activities	IPCC guidelines, national research institutions	Use of default factors, Lack of country-specific data
Non- emissions Data	- Environmental data	Deforestation rates, air and water quality data	National environmental agencies, satellite imagery	Sporadic data collection, Methodological inconsistencies
	- Socio-economic data	Income levels, educational attainment, health indicators	National statistical offices, World Bank	Insufficient data on vulnerable groups, Lag in data reporting
Projected Data	- Drivers	Predicted economic growth, demographic changes, energy price fluctuations	National economic agencies, international financial institutions	Uncertainties in projections, Over-reliance on historical trends
	- Policies	Upcoming government policies, industry standards for emission reductions	Government policy documents, industry reports	Uncertainty in policy implementation, Lack of detailed policies



# Data Types

Туре	Description	Example of GF
Measured Data	Data obtained through direct measurement, such as emissions directly measured from a smokestack.	Using sensors <sup>-</sup> power plant sr
Modelled Data	Data derived from quantitative models, representing processes like emissions from landfills or livestock.	Using simulation emissions from
Calculated Data	Data obtained by multiplying activity data by an emission factor.	Calculating fut projected num vehicle.
Estimated Data	Proxy or alternative data sources used to fill data gaps when more accurate or representative data is lacking.	Estimating agri use patterns an regions.

#### HG Emission Projections

s to directly measure CO2 emissions from a smokestack.

ion models to project methane m a landfill site in the next 10 years.

iture vehicle emissions by multiplying the mber of vehicles by the emission factor per

ricultural GHG emissions based on land and applying emission factors from similar



## Relevance/Quality of data

Regardless of the approach adopted, the need for good quality data is paramount for transparent and valuable mitigation/adaptation assessments.

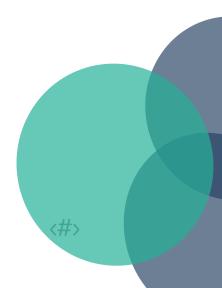
Data management cycle to perform mitigation assessments:



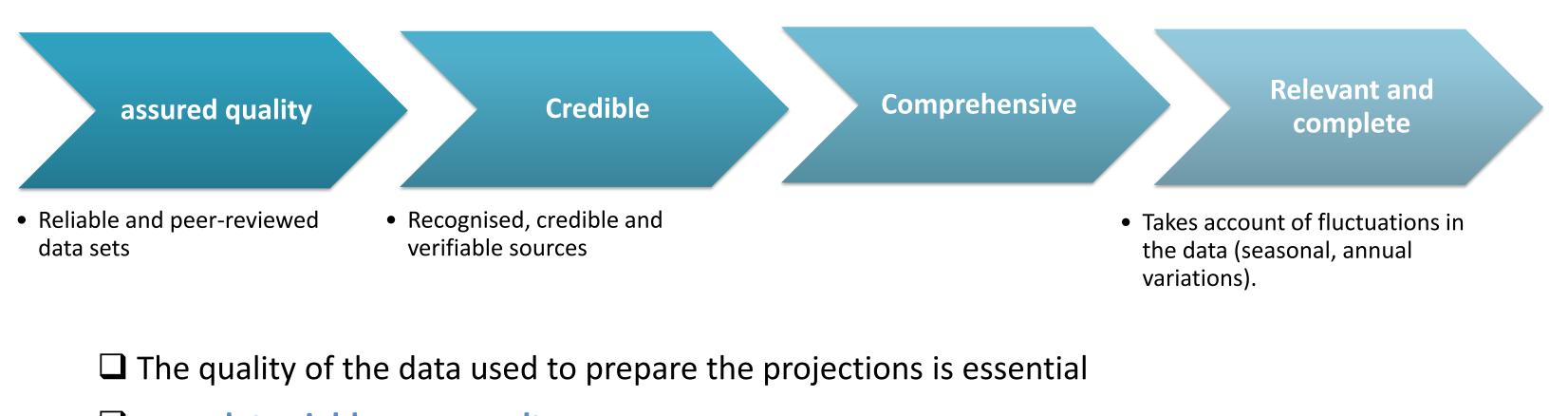
Adapted from WRI. Policy and Action Standard (2014).

Source: adapted from WRI. Policy and Action Standard (2014).





# Quality of data



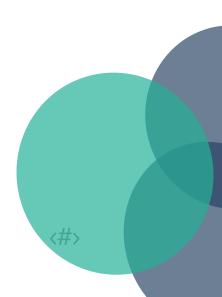
**poor data yields poor results.** 

□ Invest in the implementation of online transparency systems (MRV).

□ Invest in Quality Assurance and Quality Control (QA/QC).

**□**Establish the necessary institutional arrangements (SNI-GES).



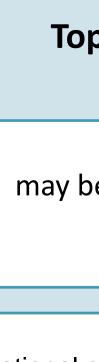


### Data Sources



measured, monitored, or collected at the source, facility, and entity or project level.

Energy used at a facility (by fuel type) and the output of the facilities production.



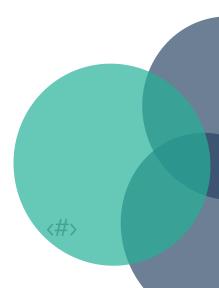
CBIT GSP

#### **Top-Down Data**

may be macroeconomic statistics

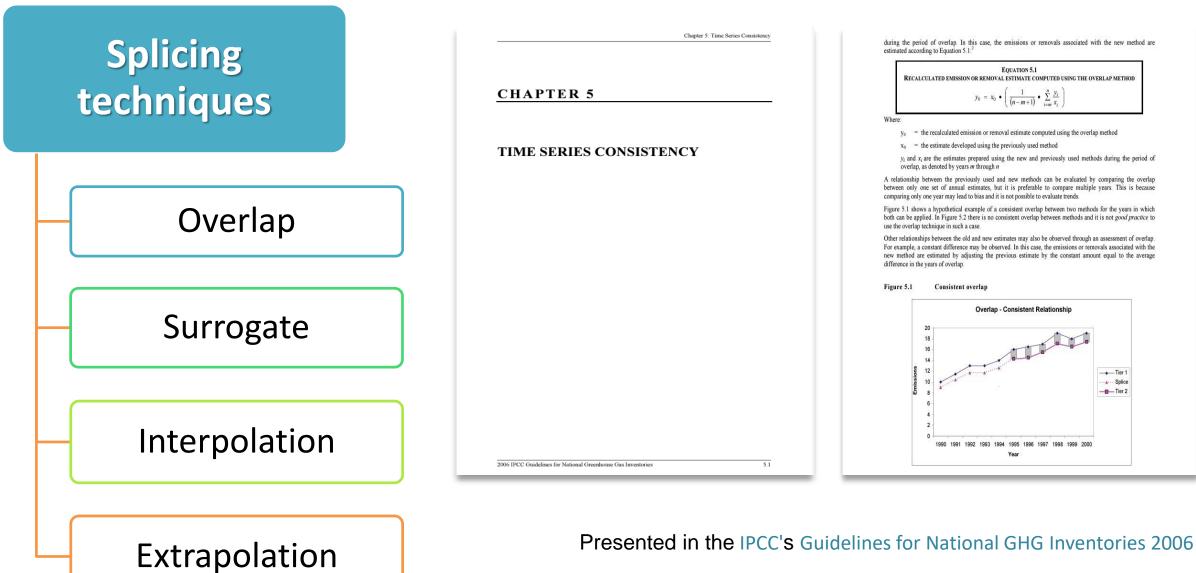
national energy use, population data, GDP, sectoral production and fuel prices.

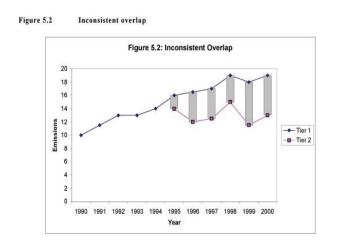




#### Overcoming non-availability of data in a time series:

Techniques to fill data gaps: Splicing Techniques

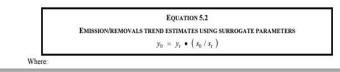




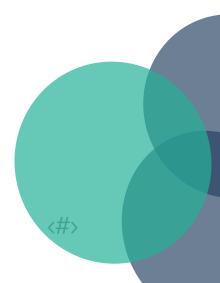
#### 5.3.3.2 SURROGATE DATA

The surrogate method relates emissions or removals to underlying activity or other indicative data. Changes in these data are used to simulate the trend in emissions or removals. The estimate should be related to the statistical data source that best explains the time variations of the category. For example, mobile source emissions may be related to trends in vehicle distances travelled, emissions from domestic wastewater may be related to population, and industrial emissions may be related to production levels in the relevant industry. See Chapter 2, Approaches to Data Collection.

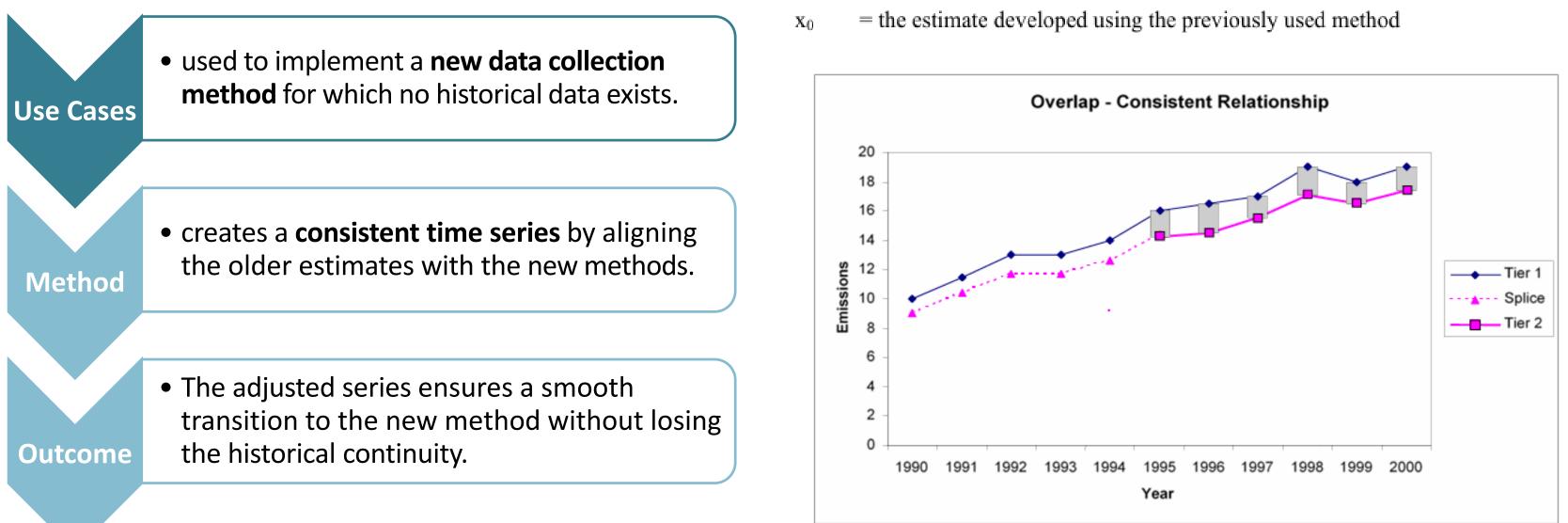
In its simplest form, the estimate will be related to a single type of data as shown in Equation 5.2:







## 1. Splicing technique: Overlap



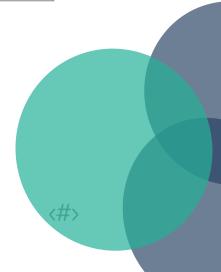
 $y_0$ 

The formula adjusts the original emission estimates to align them with a new method by averaging the ratios of the new to the old estimates over the years when the two data sets overlap.

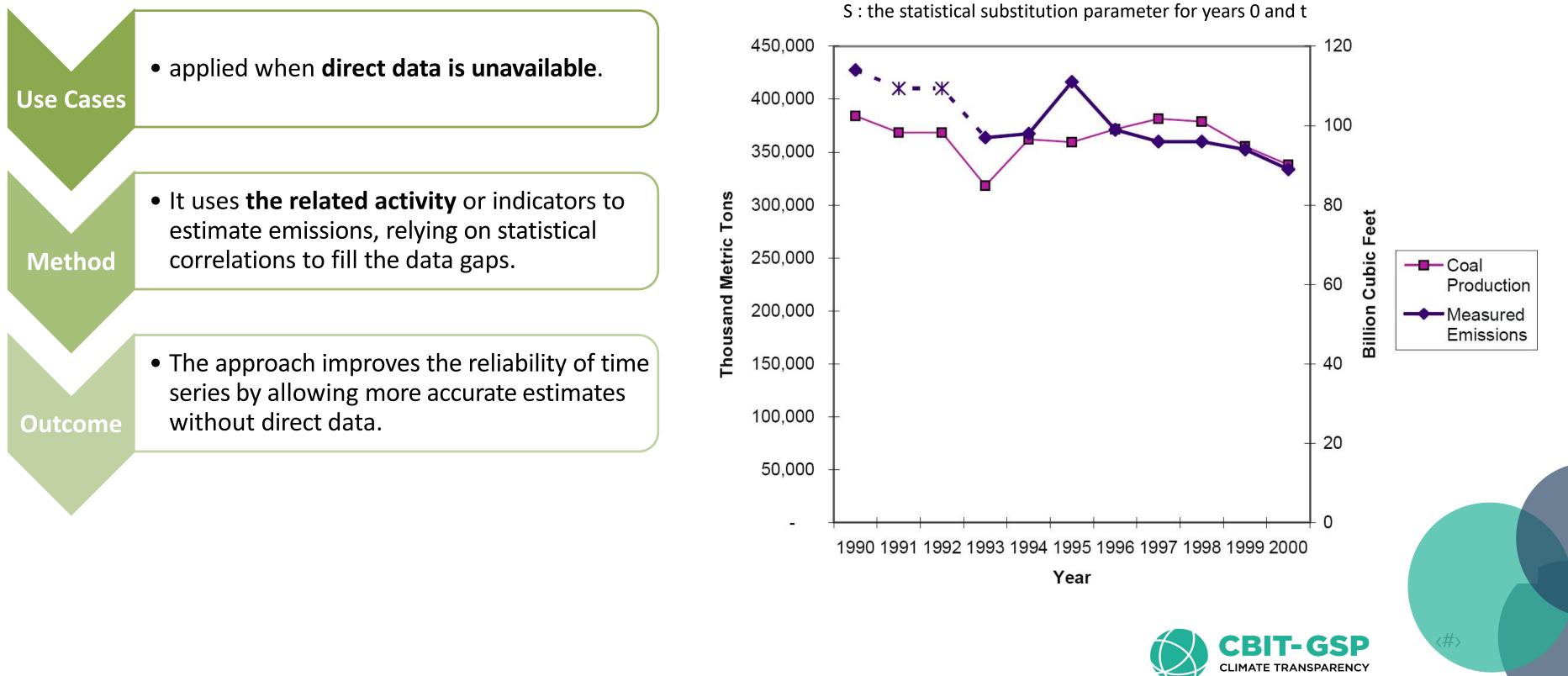
$$y_0 = x_0 \bullet \left(\frac{1}{n-m+1} \bullet \sum_{i=m}^n \frac{y_i}{x_i}\right)$$

= the recalculated emission or removal estimate computed using the overlap method
 = the estimate developed using the previously used method





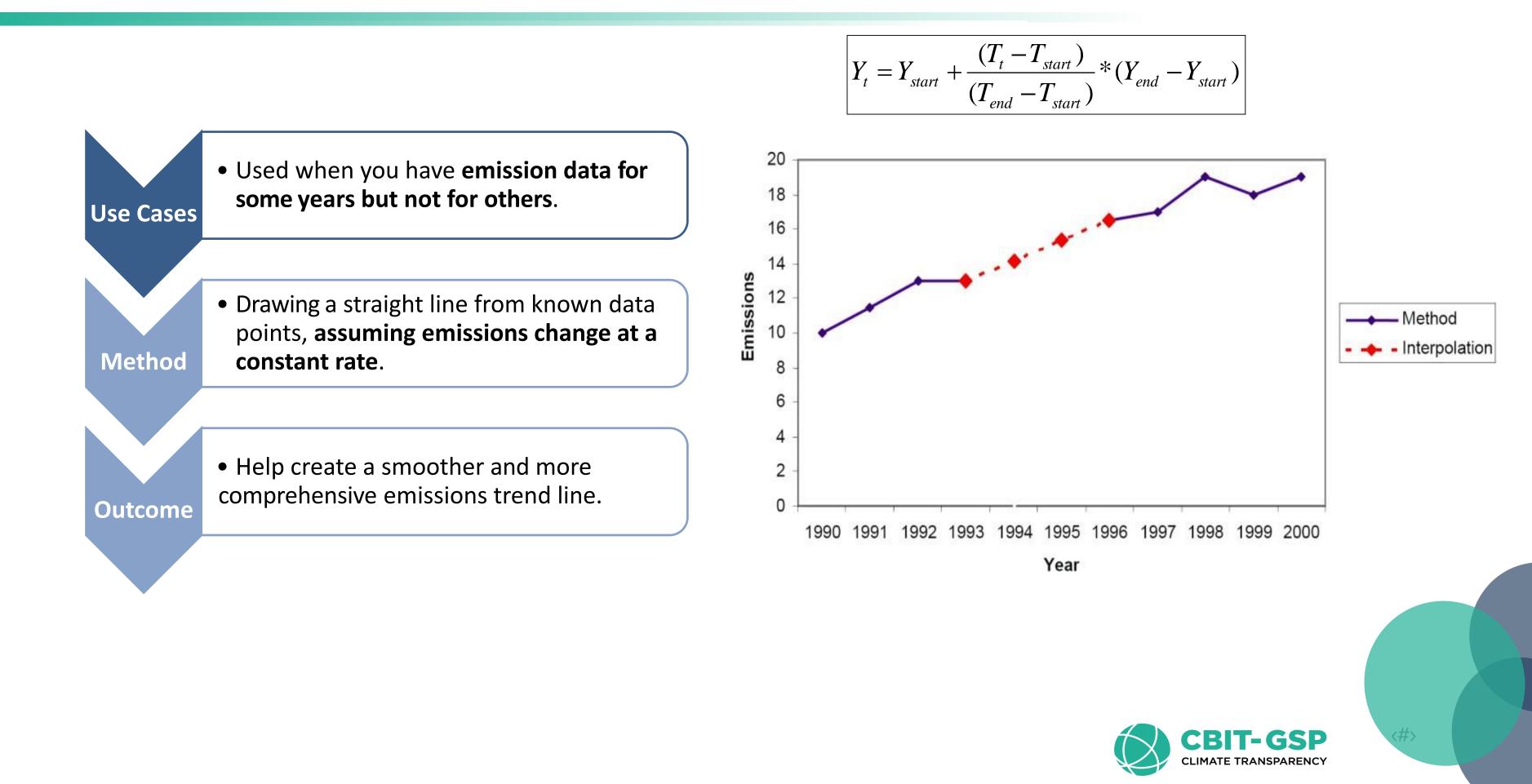
### 2. Splicing technique: Surrogate



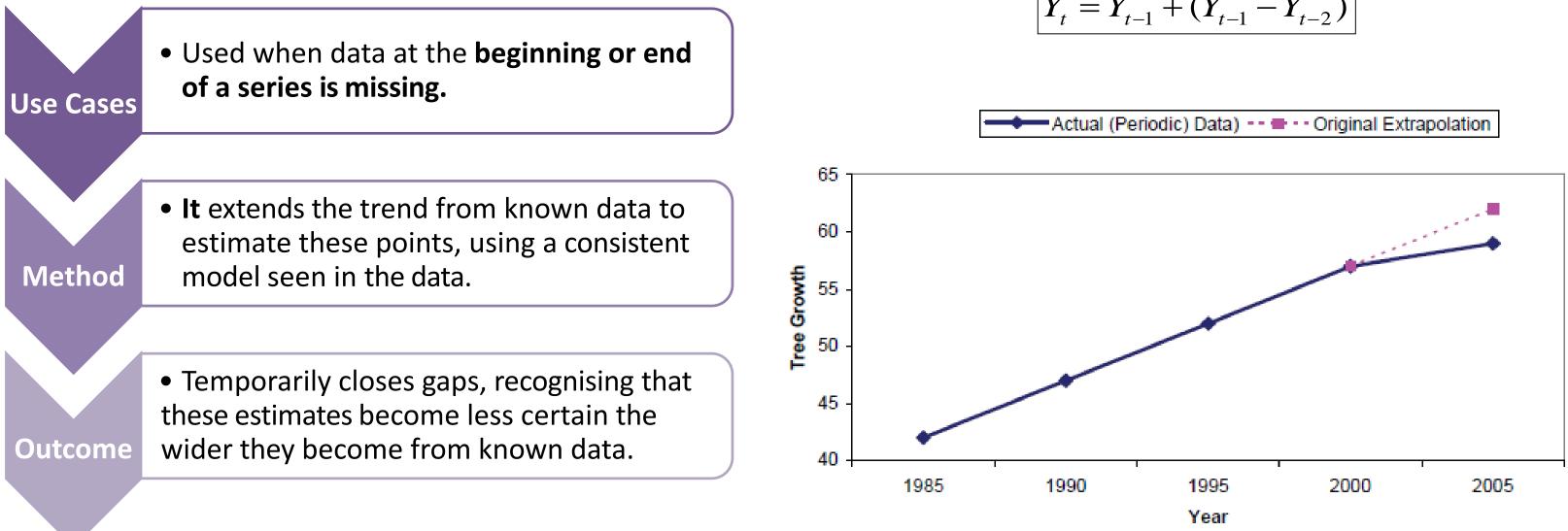
## $y_0 = y_t \bullet (s_0 / s_t)$

y : estimated emissions/removals for years 0 and t

#### **3. Splicing technique: Interpolation**



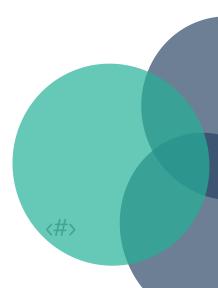
### 4. Splicing technique: Extrapolation



It is best to update them when the information becomes available.

$$Y_{t} = Y_{t-1} + (Y_{t-1} - Y_{t-2})$$





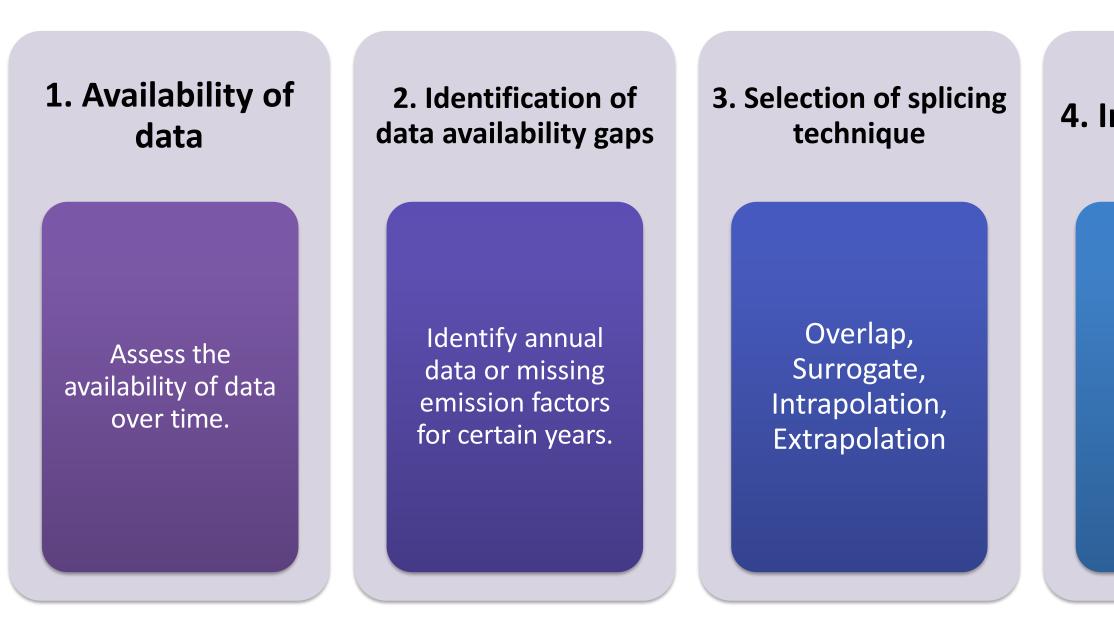
### **Summary of Splicing Techniques**

Approach	Applicability	Comments
Overlap	Data necessary to apply both the previously used and the new method must be available for at least one year, preferably more.	<ul> <li>Most reliable when the original estimates can be assessed.</li> <li>If the trends observed us inconsistent, this approar</li> </ul>
Surrogate Data	Emission factors, activity data or other estimation parameters used in the new method are strongly correlated with other well-known and more readily available indicative data.	<ul> <li>Multiple indicative data</li> <li>Should not be done for l</li> </ul>
Interpolation	Data needed for recalculation using the new method are available for intermittent years during the time series.	<ul> <li>Estimates can be linearly cannot be applied.</li> <li>The method is not applied</li> </ul>
Trend Extrapolation	Data for the new method are not collected annually and are not available at the beginning or the end of the time series.	<ul> <li>Most reliable if the trend</li> <li>Should not be used if the may be more appropriat</li> <li>Should not be done for l</li> </ul>
Other Techniques	The standard alternatives are not valid when technical conditions are changing throughout the time series (e.g., due to the introduction of mitigation technology).	<ul><li>Document customised a</li><li>Compare results with state</li></ul>

- e overlap between two or more sets of annual sed.
- using the previously used and new methods are bach is not good practice.
- a sets (singly or in combination) should be tested in order long periods.
- ly interpolated for the periods when the new method
- licable in the case of large annual fluctuations.
- nd over time is constant. he trend is changing (in this case, the surrogate method ate). <sup>•</sup> long periods.
- approaches thoroughly. tandard techniques.



#### Data collection and filling steps



#### 4. Implementation

Apply the technical selection

# 5. Validation and adjustment

Validate estimates with known data or indicators, adjust methods as needed



# Thank you for your attention!

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