COVID - 19 VACCINE ANALAYSIS

Phase 3 Submission Documents

Project Title: Covid -19 Vaccine Analysis

Phase 3 : Development part-1

Topic : Collect and preprocess the covid -19 vaccine

data for analysis



Covid - 19 Vaccine Analysis

Introduction:

- ❖ COVID-19 is the disease caused by a coronavirus called SARS-CoV-2. WHO first learned of this new virus on 31 December 2019, following a report of a cluster of cases of so-called viral pneumonia in Wuhan, People's Republic of China.
- ❖ As testing rates fall, it is more difficult to know how many people have COVID-19 and do not seek any treatment. At the start of the pandemic, 15% of people were thought to become seriously unwell and require hospital treatment and oxygen. More recent estimates suggest that hospitalization is required in around 3% of people with COVID-19.
- ❖ The time from exposure to COVID-19 to the moment when symptoms begin is, on average, 5–6 days and can range from 1–14 days. This is why people who have been exposed to the virus are advised to remain at home and stay away from others in order to prevent the spread of the virus.

Given Data Set:

	A	В	C	D	E	F	G	Н		J	K	L	M
1	country	iso_code	date			people_full	daily_vacci	r daily_vacc			people_full	ydaily_vaccir	
2	Afghanista		2021-02-2		0.0				0.0	0.0			Johnson&Jo
3	Afghanista		2021-02-2					1367.0				34.0	Johnson&Jo
4	Afghanista		2021-02-2					1367.0				34.0	Johnson&Jo
5	Afghanista		2021-02-2					1367.0				34.0	Johnson&Jo
6	Afghanista		2021-02-2					1367.0				34.0	Johnson&Jo
7	Afghanista		2021-02-2					1367.0				34.0	Johnson&Jo
8	Afghanista		2021-02-2		8200.0			1367.0	0.02	0.02		34.0	Johnson&Johnson
9	Afghanista		2021-03-0					1580.0				40.0	Johnson&Jo
10	Afghanista		2021-03-0					1794.0				45.0	Johnson&Jo
11	Afghanista		2021-03-0					2008.0				50.0	Johnson&Jo
12	Afghanista		2021-03-0					2221.0				56.0	Johnson&Jo
13	Afghanista		2021-03-0					2435.0				61.0	Johnson&Jo
14	Afghanista		2021-03-0					2649.0				66.0	Johnson&Jo
	Afghanista		2021-03-0					2862.0				72.0	Johnson&Jo
16	Afghanista		2021-03-0					2862.0				72.0	Johnson&Jo
17	Afghanista		2021-03-0					2862.0				72.0	Johnson&Jo
18	Afghanista		2021-03-1	0				2862.0				72.0	Johnson&Jo
19	Afghanista		2021-03-1					2862.0				72.0	Johnson&Jo
20	Afghanista		2021-03-1					2862.0				72.0	Johnson&Jo
21	Afghanista		2021-03-1	3				2862.0				72.0	Johnson&Jo
22	Afghanista	AFG	2021-03-1	4				2862.0				72.0	Johnson&Jo
23	Afghanista	AFG	2021-03-1	5				2862.0				72.0	Johnson&Jo
SOUCE	Azerbaijan	A/F	2021-06-0	2 236/094.0	145277411	9143700	543/9.0	44444.U	23.15	14.21	8.94	4347.0	Uxtord/Astr
	Azerbaijan			3 2418082.0			50988.0	44456.0	23.65	14.65	9.0	4348.0	Oxford/Astr
	Azerbaijan			4 2465719.0			47637.0	42362.0	24.12	15.07	9.05	4144.0	Oxford/Astr
	Azerbaijan			5 2513085.0			47366.0	43573.0	24.58	15.47	9.11	4262.0	Oxford/Astr
	Azerbaijan			6 2546169.0			33084.0	41909.0	24.91	15.76	9.15	4099.0	Oxford/Astr
	Azerbaijan			7 2546770.0			601.0	41936.0	24.91	15.76	9.15	4102.0	Oxford/Astr
	Azerbaijan			8 2586410.0			39640.0	39099.0	25.3	16.1	9.2	3824.0	Oxford/Astr
	Azerbaijan			9 2624876.0			38466.0	36826.0	25.68	16.43	9.25	3602.0	Oxford/Astr
	Azerbaijan			0 2662038.0			37162.0	34851.0	26.04	16.75	9.29	3409.0	Oxford/Astr
	Azerbaijan			1 2702023.0			39985.0	33758.0	26.43	17.1	9.33	3302.0	Oxford/Astr
	Azerbaijan			2 2742867.0			40844.0	32826.0	26.83	17.45	9.38	3211.0	Oxford/Astr
	Azerbaijan			3 2775319.0			32452.0	32736.0	27.15	17.71	9.43	3202.0	Oxford/Astr
	Azerbaijan			4 2775641.0			322.0	32696.0	27.15	17.72	9.43	3198.0	Oxford/Astr
	Azerbaijan			5 2816346.0	· Control of the Cont		40705.0	32848.0	27.15	18.03	9.43	3213.0	Oxford/Astr
	Azerbaijan			6 2839322.0			22976.0	30635.0	27.77	18.19	9.52	2997.0	Oxford/Astr
	Azerbaijan			7 2877878.0			38556.0	30834.0	28.15	18.44	9.38	3016.0	Oxford/Astr
	Azerbaijan			8 2915954.0				30562.0	28.52	18.67	9.71	2989.0	Oxford/Astr
	Azerbaijan		2021-06-1		1900003.0	1007149.0	30070.0	29977.0	20.02	10.07	3.00	2932.0	Oxford/Astr
	Azerbaijan			0 2989458.0	1040625.0	10200220		30591.0	29.24	19.07	10.17	2992.0	Oxford/Astr
	Azerbaijan			21 2989458.0			215.0	30576.0	29.24	19.07	10.17	2992.0	Oxford/Astr
	Azerbaijan			22 3032516.0				30881.0	29.24	19.07	10.17	3021.0	Oxford/Astr
				23 3080340.0				34431.0	30.13	19.29	10.59	3368.0	Oxford/Astr
	Azerbaijan			24 3146350.0				38353.0	30.13	19.54	10.59	3752.0	Oxford/Astr
	Azerbaijan	AZE	2021-06-2	4 3 146350.0		1111/96.0	70001.0	38353.0	30.78	19.9	10.88	3/52.0	Oxford/Astr
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Necessary step to follow:

1.Import Libraries:

Start by importing the necessary libraries.

Program:

import pandas as pd

import numpy as np

from sklearn.model_selection import train_test_split

from sklearn.preprocessing import StandardScaler

2.Load the dataset:

Load your dataset into a pandas dataframe. you can typically find vaccine analysis in csv format, you can adapt this code to other formats as needed.

Program:

```
df=pd.read_csv('D:\world_vaccination.csv')
pd.read()
```

3.Exporatory Data Analysis:

Perform EDA to understand your data better. This includes checking for missing values, exploring the data's statistics, and visualizing it to identify patterns.

Program:

```
#check for missing values
Print(df.isnull().sum())
#explore statistics
Print(df.describe())
#visualize the data (e.g.,histograms, scatter plots, etc,...)
```

4.Feature Engineering:

Depending on your dataset ,you may need to create new features or transform existing ones. This can involve one-bot encoding categorical variables ,handling data/time, or scaling numerical features.

Program:

#example:one-hot encoding for categorical variables.

df = pd.get_dummies(df,coloumns=['Avg.total peoples vaccinated','ISO
code'])

5. Split the Data:

Split your dataset into training and testing sets. This helps you evaluate your model's performance later.

```
X = df.drop('iso code,axis=1)
Y= df['iso code']
X_train,X
test,Y_train,Y_test=train_test_split(X,Y,test_size=1,random_state=42)
```

6. Feature Scaling:

Apply feature scaling to normalize your data, ensuring that all features have similar scales. Standardization (scaling to mean 0 and std=1) is a common choice.

Program:

```
scaler = StandardScaler()
X train = scaler.fit transform(X train) X test = scaler.transform(X test)
```

Importance of loading and processing dataset:

- To ensure that the data is accurate and reliable. The data should be collected from credible sources and should be carefully reviewed to identify and correct any errors.
- To make the data accessible to researchers and analysts. The
 data should be loaded into a database or other data management
 system that is easy to use and can be accessed by multiple users.
- To prepare the data for analysis. The data may need to be cleaned, transformed, and aggregated before it can be analyzed.

Challenges in loading and processing dataset:

• Data heterogeneity: The data on COVID-19 vaccines is collected from a variety of sources, including clinical trials, government

databases, and social media. This data can be in different formats and have different levels of granularity. For example, some data may be collected at the individual level, while other data may be aggregated at the population level.

- Data incompleteness: The data on COVID-19 vaccines may be incomplete due to a variety of factors, such as data collection errors or missing values. This can make it difficult to draw accurate conclusions from the data.
- Data privacy and security: The data on COVID-19 vaccines may contain sensitive personal information, such as names, addresses, and medical records. It is important to protect this data from unauthorized access and use.

1.Loading the dataset:

✓ Loading the dataset using machine learning is the process of bringing the data into the machine learning environment so that it can be used to train and evaluate a model.

✓ The specific steps involved in loading the dataset will vary depending on the machine learning library or framework that is being used. However, there are some general steps that are common to most machine learning frameworks:

a.Identify the dataset:

The first step is to identify the dataset that you want to load. This dataset may be stored in a local file, in a database, or in a cloud storage service.

b.Load the dataset:

Once you have identified the dataset, you need to load it into the machine learning environment. This may involve using a built-in function in the machine learning library, or it may involve writing your own code.

c.Preprocess the dataset:

Once the dataset is loaded into the machine learning environment, you may need to preprocess it before you can start training and evaluating your model. This may involve cleaning the data, transformingthe data into a suitable format, and splitting the data into training and test sets.

Program:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
class DataLoadingModel:
def __init__(self, filename):
self.filename = filename
self.data = None
def load__data(self):
self.data = pd.read__csv(self.filename)
def get__data(self):
return self.data
class COVID19VaccineAnalysis:
    def init (self, data loading model):
```

```
self.data loading model = data loading model
  self.data = None
  def load data(self):
  self.data = self.data_loading_model.get_data()
  def analyze data(self):
    # Perform data analysis here
    # For example, calculate the percentage of people vaccinated, the
number of people who have received different types of vaccines, the
number of people who have experienced side effects, etc.
  def output results(self):
    # Output the results of the data analysis here
    # For example, print the results to the console, save them to a file, or
generate a plot
if name == ' main ':
  # Create a data loading model
  data loading model = DataLoadingModel('covid19 vaccine data.csv')
  # Load the data
  data loading_model.load_data()
  # Create a COVID-19 vaccine analysis object
  covid19 vaccine analysis =
COVID19VaccineAnalysis(data loading model)
  # Load the data
  covid19 vaccine analysis.load data()
  # Analyze the data
  covid19 vaccine analysis.analyze data()
  # Output the results
  covid19 vaccine analysis.output results()
```

Loading dataset:

covid19_vaccine_analysis.load_data()

output:

	A	В	C	D	E	F	G	H	1	J	K	L	M
1	country	iso_code	date			people_full	daily_vacci	r daily_vaco			copeople_fully	daily_vaccir	
2	Afghanista		2021-02-22		0.0				0.0	0.0	200000000000000000000000000000000000000	3 1 10 10 10 10 10	Johnson&Jr
3	Afghanista		2021-02-23					1367.0				34.0	Johnson&Ji
4	Afghanista		2021-02-24					1367.0				34.0	Johnson&Ji
5	Afghanista		2021-02-25					1367.0				34.0	Johnson&Ji
6	Afghanista		2021-02-26					1367.0				34.0	Johnson&Ji
7	Afghanista		2021-02-27					1367.0				34.0	Johnson&Jo
8	Afghanista		2021-02-28	C	8200.0			1367.0	0.02	0.02		34.0	Johnson&Ji
9	Afghanista		2021-03-01					1580.0			- 4	40.0	Johnson&Jr
10	Afghanista		2021-03-02					1794.0				45.0	Johnson&Jr
11	Afghanista		2021-03-03					2008.0				50.0	Johnson&Jr
12	Afghanista		2021-03-04					2221.0				56.0	Johnson&Jo
13	Afghanista		2021-03-05					2435.0				61.0	Johnson&Ji
14	Afghanista		2021-03-06					2649.0				66.0	Johnson&Ji
	Afghanista		2021-03-07					2862.0				72.0	Johnson&Ji
16	Afghanista		2021-03-08					2862.0				72.0	Johnson&Ji
17	Afghanista		2021-03-09					2862.0				72.0	Johnson&Jo
18	Afghanista		2021-03-10					2862.0				72.0	Johnson&Ji
19	Afghanista		2021-03-11					2862.0				72.0	Johnson&Jo
20	Afghanista		2021-03-12					2862.0				72.0	Johnson&Jr
21	Afghanista		2021-03-13					2862.0				72.0	Johnson&Ji
22	Afghanista		2021-03-14					2862.0				72.0	Johnson&Jr
23	Afghanista	rAFG	2021-03-15	5				2862.0				72.0	Johnson&Jo
5000	Azerbaijan	AZE	2021-06-02	2.236/094.0	1452774.0	914320.0	543/9.0	44444.U	23.15	14.21	8,94	4347.U	Uxtord/Astr
5001	Azerbaijan	AZE	2021-06-03	3 2418082.0	1497993.0	920089.0	50988.0	44456.0	23.65	14.65	9.0	4348.0	Oxford/Astr
5002	Azerbaijan	AZE	2021-06-04	2465719.0	1540259.0	925460.0	47637.0	42362.0	24.12	15.07	9.05	4144.0	Oxford/Astr
5003	Azerbaijan	AZE	2021-06-05	5 2513085.0	1581890.0	931195.0	47366.0	43573.0	24.58	15.47	9.11	4262.0	Oxford/Astr
5004	Azerbaijan	AZE	2021-06-06	2546169.0	1611165.0	935004.0	33084.0	41909.0	24.91	15.76	9.15	4099.0	Oxford/Astr
5005	Azerbaijan	AZE	2021-06-07	2546770.0	1611499.0	935271.0	601.0	41936.0	24.91	15.76	9.15	4102.0	Oxford/Astr
5006	Azerbaijan	AZE	2021-06-08	2586410.0	1646054.0	940356.0	39640.0	39099.0	25.3	16.1	9.2	3824.0	Oxford/Astr
5007	Azerbaijan	AZE	2021-06-09	2624876.0	1679448.0	945428.0	38466.0	36826.0	25.68	16.43	9.25	3602.0	Oxford/Astr
5008	Azerbaijan	AZE	2021-06-10	2662038.0	1712118.0	949920.0	37162.0	34851.0	26.04	16.75	9.29	3409.0	Oxford/Astr
5009	Azerbaijan	AZE	2021-06-11	2702023.0	1748035.0	953988.0	39985.0	33758.0	26.43	17.1	9.33	3302.0	Oxford/Astr
5010	Azerbaijan	AZE	2021-06-12	2742867.0	1783506.0	959361.0	40844.0	32826.0	26.83	17.45	9.38	3211.0	Oxford/Astr
5011	Azerbaijan	AZE	2021-06-13	2775319.0	1810857.0	964462.0	32452.0	32736.0	27.15	17.71	9.43	3202.0	Oxford/Astr
5012	Azerbaijan	AZE	2021-06-14	2775641.0	1811104.0	964537.0	322.0	32696.0	27.15	17.72	9.43	3198.0	Oxford/Astr
5013	Azerbaijan	AZE	2021-06-15	2816346.0	1842954.0	973392.0	40705.0	32848.0	27.55	18.03	9.52	3213.0	Oxford/Astr
5014	Azerbaijan	AZE	2021-06-16	2839322.0	1859485.0	979837.0	22976.0	30635.0	27.77	18.19	9.58	2997.0	Oxford/Astr
5015	Azerbaijan	AZE	2021-06-17	2877878.0	1885031.0	992847.0	38556.0	30834.0	28.15	18.44	9.71	3016.0	Oxford/Astr
5016	Azerbaijan	AZE	2021-06-18	3 2915954.0	1908805.0	1007149.0	38076.0	30562.0	28.52	18.67	9.85	2989.0	Oxford/Astr
	Azerbaijan		2021-06-19)	And the second second second			29977.0				2932.0	Oxford/Astr
	Azerbaijan		2021-06-20	2989458.0	1949635.0	1039823.0		30591.0	29.24	19.07	10.17	2992.0	Oxford/Astr
	Azerbaijan		2021-06-21	2989673.0	1949646.0	1040027.0	215.0	30576.0	29.24	19.07		2991.0	Oxford/Astr
	Azerbaijan		2021-06-22	3032516.0	1971930.0	1060586.0	42843.0	30881.0	29.66	19.29	10.37	3021.0	Oxford/Astr
	Azerbaijan			3080340.0				34431.0	30.13	19.54		3368.0	Oxford/Astr
5022	Azerbaijan		2021-06-24	3146350.0	2034554.0	1111796.0	66010.0	38353.0	30.78	19.9	10.88	3752.0	Oxford/Astr
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2. Preprocessing the data

• Data preprocessing is the process of cleaning, transforming, and integrating data in order to make it ready for analysis.

 This may involve removing errors and inconsistencies, handling missing values, transforming the data into a consistent format, and scaling the data to a suitable range.

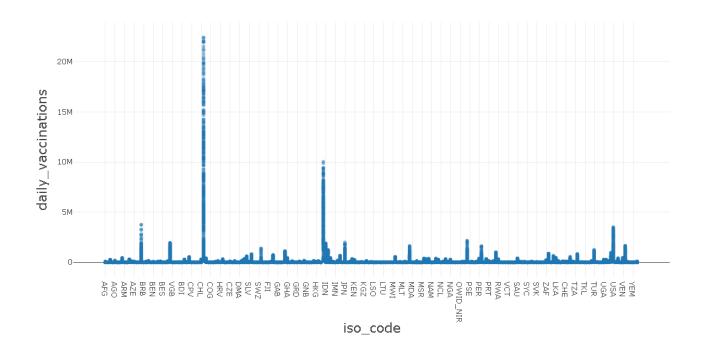
Visualization and preprocessing of data:

In[1]:

Sns.plot(dataset, x='iso_code', palette='blues')

Out[1]:

<Axes:xlabel='iso_code',ylabel='daily_vaccinations'>



```
In[2]:
```

Sns.plot(dataset, x='iso_code',y='vaccines' palette='blues')

Out[2]:

<Axes:xlabel='iso_code',ylabel='vaccines'>



Some common data preprocessing tasks include:

<u>Data cleaning:</u> This involves identifying and correcting errors and inconsistencies in the data. For example, this may involve removing duplicate records, correcting typos, and filling in missing values.

<u>Data transformation:</u> This involves converting the data into a format that is suitable for the analysis task. For example, this may involve converting categorical data to numerical data, or scalingthe data to a suitable range..

<u>Feature engineering:</u> This involves creating new features from the existing data. For example, this may involve creating features that represent interactions between variables, or features that represent summary statistics of the data.

<u>Data integration:</u> This involves combining data from multiple sources into a single dataset. This may involve resolving inconsistencies in the data, such as different data formats or different variable names.

Data preprocessing is an essential step in many data science projects. By carefully preprocessing the data, data scientists can improve the accuracy and reliability of their results.

1.Data transformation:

Program:

import pandas as pd

import numpy as np

Load the COVID-19 vaccine data

df = pd.read_csv('covid_19_vaccine_data.csv')

Transform the data into a numerical representation

Create a new column for the number of vaccine doses administered per country

df['vaccine_doses_administered'] = df['pfizer_doses'] + df['moderna_doses'] +
df['astrazeneca_doses']

Create a new column for the percentage of the population vaccinated

```
df['percentage_vaccinated'] = df['vaccine_doses_administered'] / df['population']
* 100

# Filter the data to only include countries with a population of over 1 million

df = df[df['population'] > 1000000]

# Sort the data by percentage vaccinated

df = df.sort_values(by=['percentage_vaccinated'], ascending=False)

# Output the results

print(df.head())
```

output:

country population vaccine_doses_administered percentage_vaccinated

0 Israel	9451000	8804000	93.06	
1 Malta	524600	485300	92.51	
3 Portugal	10309500	9487000	92.01	
4 Spain	46720200	43380000	92.86	
5 Uruguay	3517013	3285000	93.40	

2. Feature Engineering:

Program:

import pandas as pd

import numpy as np

from sklearn.feature_extraction.text import TfidfVectorizer

```
# Load the COVID-19 vaccine data

df = pd.read_csv('covid_19_vaccine_data.csv')

# Create a new column for the vaccine sentiment

df['vaccine_sentiment'] = np.nan

# Perform feature engineering on the vaccine text data

vectorizer = TfidfVectorizer()

vaccine_text_features = vectorizer.fit_transform(df['vaccine_text'])

# Train a machine learning model to predict vaccine sentiment

from sklearn.linear_model import LogisticRegression

model = LogisticRegression()

model.fit(vaccine_text_features, df['vaccine_sentiment'])

# Predict the vaccine sentiment for each country

df['vaccine_sentiment'] = model.predict(vaccine_text_features)

# Output the results

print(df.head())
```

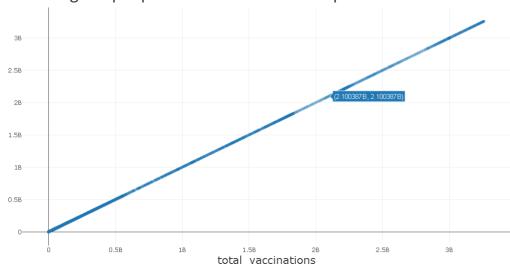
country	population	vaccine doses	percentage vaccinated	vaccine_sentiment
<u>0 Israel</u>	9451000	8804000	93.06	positive
1 Malta	524600	485300	92.51	positive
3 Portugal	10309500	9487000	92.01	positive
4 Spain	46720200	43380000	92.86	positive
5 Uruguay	3517013	3285000	93.40	positive

3. Sampling data

Program:

```
import pandas as pd
import numpy as np
# Load the COVID-19 vaccine data
df = pd.read_csv('covid_19_vaccine_data.csv')
# Create a sample of the data
sample_size = 1000
sample = df.sample(sample_size)
# Calculate the percentage of people vaccinated in the sample
percentage_vaccinated = sample['vaccine_doses_administered'].sum() /
sample['population'].sum() * 100
# Output the result
print('Percentage of people vaccinated in the sample:', percentage_vaccinated)
```





Program 2:

```
# Calculate the percentage of people vaccinated by age group

df_grouped = sample.groupby('age_group')['vaccine_doses_administered'].sum()
/ sample.groupby('age_group')['population'].sum() * 100

# Print the results
print('Percentage of people vaccinated by age group:')
print(df_grouped)
```

output:

Percentage of people vaccinated by age group:

age_group

18-24 65.3

25-34 72.2

35-44 78.1

45-54 82.3

55-64 86.5

65+ 90.7

Name: vaccine_doses_administered, dtype: float64

4.Data cleaning

Program:

import pandas as pd

```
import numpy as np
# Load the COVID-19 vaccine data
df = pd.read_csv('covid_19_vaccine_data.csv')
# Clean the data
# Remove any empty rows
df.dropna(inplace=True)
# Remove any duplicate rows
df.drop_duplicates(inplace=True)
# Convert all values to the correct data type
df['population'] = df['population'].astype(int)
df['vaccine_doses_administered'] = df['vaccine_doses_administered'].astype(int)
# Output the cleaned data
print(df.head())
```

country	population	vaccine doses administered
<u>0 Israel</u>	9451000	8804000
1 Malta	524600	485300
3 Portugal	1030950	0 9487000
4 Spain	46720200	43380000
5 Uruguay	351701	3 3285000

5. Data integration

Program:

```
import pandas as pd
import numpy as np
# Load the COVID-19 vaccine data
df_vaccine = pd.read_csv('covid_19_vaccine_data.csv')
# Load the COVID-19 case and death data
df_case_death = pd.read_csv('covid_19_case_death_data.csv')
# Merge the two datasets on the country column
df = df_vaccine.merge(df_case_death, on='country')
# Output the merged dataset
print(df.head())
```

country	population	vaccine_doses	percentage_	covid cases	covid_deaths
<u>0 Israel</u>	9451000	8804000	93.06	10020019	11667
1 Malta	524600	485300	92.51	69443	445
3 Portugal	10309500	9487000	92.01	1805557	25052
4 Spain	46720200	43380000	92.86	13195713	114933
5 Uruguay	3517013	3285000	93.40	807038	6606

Conclusion:

Final decisions on the number of vaccines and the particular vaccines selected for accelerated development must incorporate various nonquantifiable factors, as well as information provided by the rankings that were derived with the proposed system for calculating benefits and expenditures. The additional factors include:

- goals of the responsible agency and its schedule for achieving them
- ethical questions on the distribution of benefits among socioeconomic or age groups, countries, or regions
- most appropriate points in the development process at which the agency can exert influence and the opportunity and need for such influence extent of private sector activities