

각 데이터 별 결과 시각화

```
In [6]: import pandas as pd
import os
import matplotlib.pyplot as plt
import seaborn as sns

# 파일이 저장된 디렉토리 경로 (예시)
directory_path = './combined_data'

# 디렉토리에서 모든 CSV 파일 리스트 가져오기
csv_files = [f for f in os.listdir(directory_path) if f.endswith('.csv')]
```

```
In [7]: csv_files
```

```
Out[7]: ['dohoon_I_1_combined.csv',
'dohoon_I_2_combined.csv',
'dohoon_I_3_combined.csv',
'dohoon_I_4_combined.csv',
'dohoon_S_1_combined.csv',
'dohoon_S_2_combined.csv',
'dohoon_S_3_combined.csv',
'dohoon_S_4_combined.csv',
'jaeho_I_1_combined.csv',
'jaeho_I_2_combined.csv',
'jaeho_I_3_combined.csv',
'jaeho_I_4_combined.csv',
'jaeho_S_1_combined.csv',
'jaeho_S_2_combined.csv',
'jaeho_S_3_combined.csv',
'jaeho_S_4_combined.csv',
'jaewan_I_1_combined.csv',
'jaewan_I_2_combined.csv',
'jaewan_I_3_combined.csv',
'jaewan_I_4_combined.csv',
'jaewan_S_1_combined.csv',
'jaewan_S_2_combined.csv',
'jaewan_S_3_combined.csv',
'jaewan_S_4_combined.csv',
'jiyoung_I_1_combined.csv',
'jiyoung_I_2_combined.csv',
'jiyoung_I_3_combined.csv',
'jiyoung_I_4_combined.csv',
'jiyoung_S_1_combined.csv',
'jiyoung_S_2_combined.csv',
'jiyoung_S_3_combined.csv',
'jiyoung_S_4_combined.csv',
'minho_I_1_combined.csv',
'minho_I_2_combined.csv',
'minho_I_3_combined.csv',
'minho_I_4_combined.csv',
'minho_S_1_combined.csv',
'minho_S_2_combined.csv',
'minho_S_3_combined.csv',
'minho_S_4_combined.csv',
'nayoung_I_1_combined.csv',
'nayoung_I_2_combined.csv',
'nayoung_I_3_combined.csv',
'nayoung_I_4_combined.csv',
'nayoung_S_1_combined.csv',
'nayoung_S_2_combined.csv',
'nayoung_S_3_combined.csv',
'nayoung_S_4_combined.csv',
'seongjun_I_1_combined.csv',
'seongjun_I_2_combined.csv',
'seongjun_I_3_combined.csv',
'seongjun_I_4_combined.csv',
'seongjun_S_1_combined.csv',
'seongjun_S_2_combined.csv',
'seongjun_S_3_combined.csv',
'seongjun_S_4_combined.csv',
'seoyeong_I_1_combined.csv',
'seoyeong_I_2_combined.csv',
'seoyeong_I_3_combined.csv',
'seoyeong_I_4_combined.csv',
```

```

'seoyeong_S_1_combined.csv',
'seoyeong_S_2_combined.csv',
'seoyeong_S_3_combined.csv',
'seoyeong_S_4_combined.csv',
'taeyoon_I_1_combined.csv',
'taeyoon_I_2_combined.csv',
'taeyoon_I_3_combined.csv',
'taeyoon_I_4_combined.csv',
'taeyoon_S_1_combined.csv',
'taeyoon_S_2_combined.csv',
'taeyoon_S_3_combined.csv',
'taeyoon_S_4_combined.csv']

```

In [9]: # 각 파일의 *Sequence* 값을 설정하는 조건

```

sequence_map = {
    "dohoon_I": "2", "dohoon_S": "1",
    "jiyoung_I": "2", "jiyoung_S": "1",
    "jaeho_I": "2", "jaeho_S": "1",
    "jaewan_I": "1", "jaewan_S": "2",
    "minho_I": "1", "minho_S": "2",
    "nayoung_I": "1", "nayoung_S": "2",
    "seongjun_I": "1", "seongjun_S": "2",
    "taeyoon_I": "2", "taeyoon_S": "1"
}

# 결과를 저장할 리스트 초기화
comparison_results = []

# 각 CSV 파일 경로를 불러와 처리하는 코드
def process_and_add_sequence_with_path(csv_files, sequence_map, directory_path):
    for csv_file in csv_files:
        # 파일 경로 만들기
        file_path = os.path.join(directory_path, csv_file)

        # 파일 이름에서 접두사 추출 (예: dohoon_I, jiyoung_S 등)
        prefix = '_' .join(csv_file.split('_')[:2])

        # CSV 파일 불러오기
        try:
            df = pd.read_csv(file_path)
        except FileNotFoundError:
            print(f"File not found: {file_path}")
            continue

        # Sequence 값 설정
        sequence_value = sequence_map.get(prefix)

        if sequence_value:
            # Sequence 컬럼 추가
            df['Sequence'] = sequence_value

            # maxforce와 result 열이 있는지 확인
            if 'maxforce' in df.columns and 'result' in df.columns:
                # 상관계수 계산
                correlation = df['maxforce'].corr(df['result'])

                # 차이의 절대값 평균 계산
                mean_absolute_difference = (df['maxforce'] - df['result']).abs()

                # maxforce와 result가 동일한 값의 비율 계산

```

```

        identical_percentage = (df['maxforce'] == df['result']).mean() *

# 결과 저장
comparison_results.append({
    'filename': csv_file,
    'correlation': correlation,
    'mean_absolute_difference': mean_absolute_difference,
    'identical_percentage': identical_percentage,
    'Sequence': sequence_value
})

# 수정된 파일을 같은 경로에 저장
df.to_csv(file_path, index=False)
print(f"Processed {file_path} and added Sequence {sequence_value}")

process_and_add_sequence_with_path(csv_files, sequence_map, directory_path)

# 결과를 데이터프레임으로 변환
results_df = pd.DataFrame(comparison_results)

# Sequence별 identical_percentage 평균 계산
sequence_grouped = results_df.groupby('Sequence')['identical_percentage'].mean()

# 그래프 그리기
plt.figure(figsize=(8, 6))
bars = sequence_grouped.plot(kind='bar', color=['blue', 'orange'])

# 각 막대에 퍼센티지 값 표시
for bar in bars.patches:
    plt.text(bar.get_x() + bar.get_width() / 2,
             bar.get_height(),
             f'{bar.get_height():.2f}%', # 퍼센티지 값 표시
             ha='center',
             va='bottom')

# 그래프 제목 및 라벨 설정
plt.title('Average Identical Percentage by Sequence')
plt.xlabel('Sequence')
plt.ylabel('Average Identical Percentage (%)')

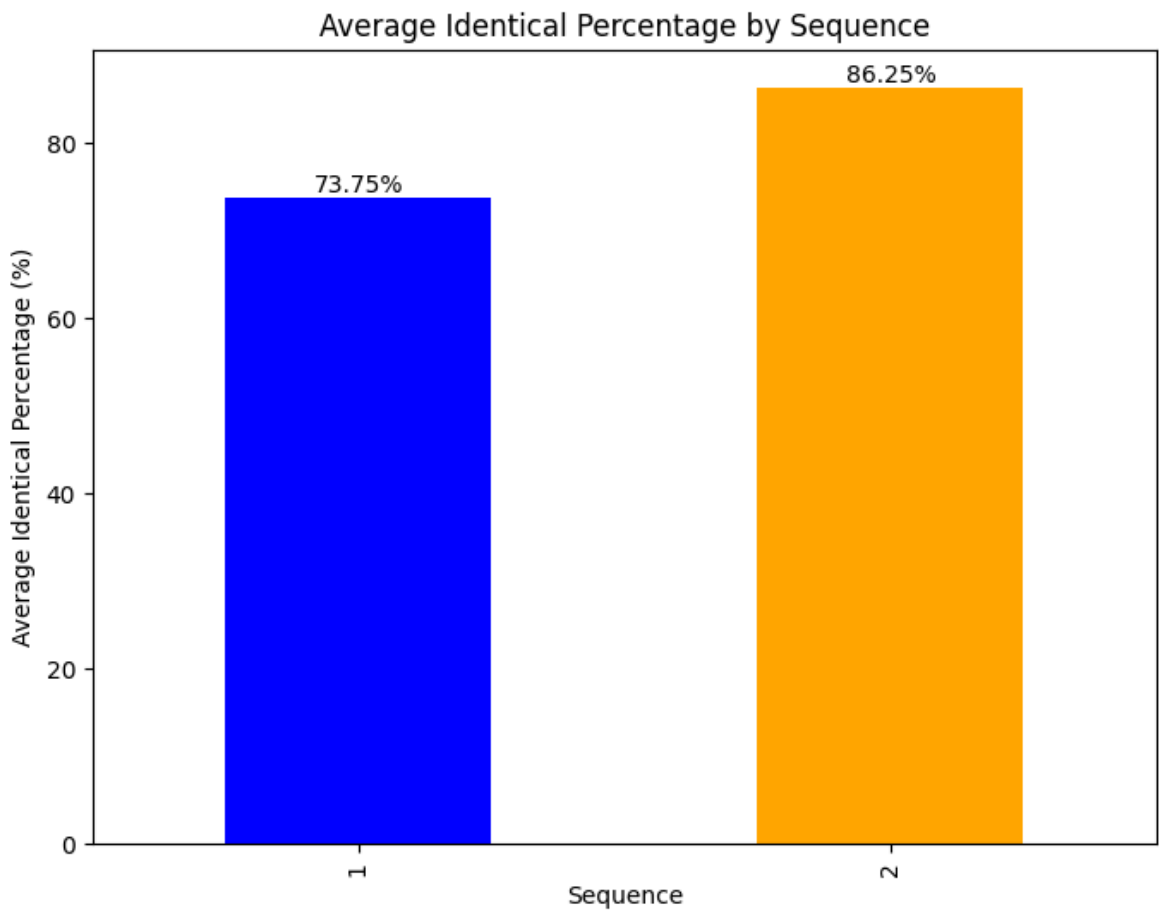
# 그래프 출력
plt.show()

# 결과 데이터프레임 출력
print(results_df)

```

[illegible]

Processed ./combined_data\taeyoon_S_1_combined.csv and added Sequence 1
 Processed ./combined_data\taeyoon_S_2_combined.csv and added Sequence 1
 Processed ./combined_data\taeyoon_S_3_combined.csv and added Sequence 1
 Processed ./combined_data\taeyoon_S_4_combined.csv and added Sequence 1



	filename	correlation	mean_absolute_difference \
0	dohoon_I_1_combined.csv	0.993892	0.008
1	dohoon_I_2_combined.csv	1.000000	0.000
2	dohoon_I_3_combined.csv	0.993877	0.008
3	dohoon_I_4_combined.csv	1.000000	0.000
4	dohoon_S_1_combined.csv	0.994014	0.008
..
59	taeyoon_I_4_combined.csv	0.805299	0.116
60	taeyoon_S_1_combined.csv	0.427669	0.276
61	taeyoon_S_2_combined.csv	0.515988	0.236
62	taeyoon_S_3_combined.csv	0.767243	0.136
63	taeyoon_S_4_combined.csv	0.619546	0.184

	identical_percentage	Sequence
0	96.0	2
1	100.0	2
2	96.0	2
3	100.0	2
4	96.0	1
..
59	72.0	2
60	32.0	1
61	48.0	1
62	60.0	1
63	56.0	1

[64 rows x 5 columns]

```

In [35]: # 각 CSV 파일에 대해 작업 수행
for csv_file in csv_files:
    file_path = os.path.join(directory_path, csv_file)

    # CSV 파일 읽기
    df = pd.read_csv(file_path)

    # maxforce와 result 열이 있는지 확인
    if 'maxforce' in df.columns and 'result' in df.columns:
        maxforce_values = [0.0, 0.1, 0.3, 0.5, 1.0]
        result_values = [0.0, 0.1, 0.3, 0.5, 1.0]

        distribution = []

        for maxforce_value in maxforce_values:
            for result_value in result_values:
                count = len(df[(df['maxforce'] == maxforce_value) & (df['result'
                distribution.append({
                    'maxforce_value': maxforce_value,
                    'result_value': result_value,
                    'count': count
                })

        distribution_df = pd.DataFrame(distribution)

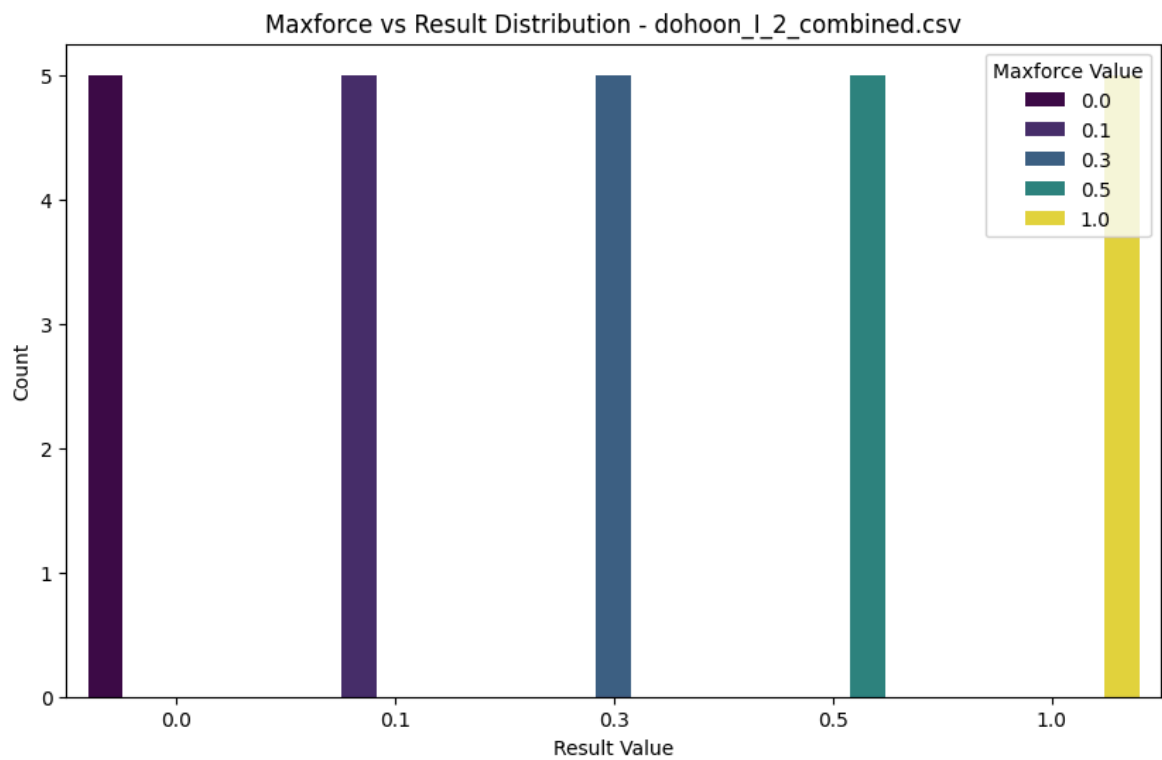
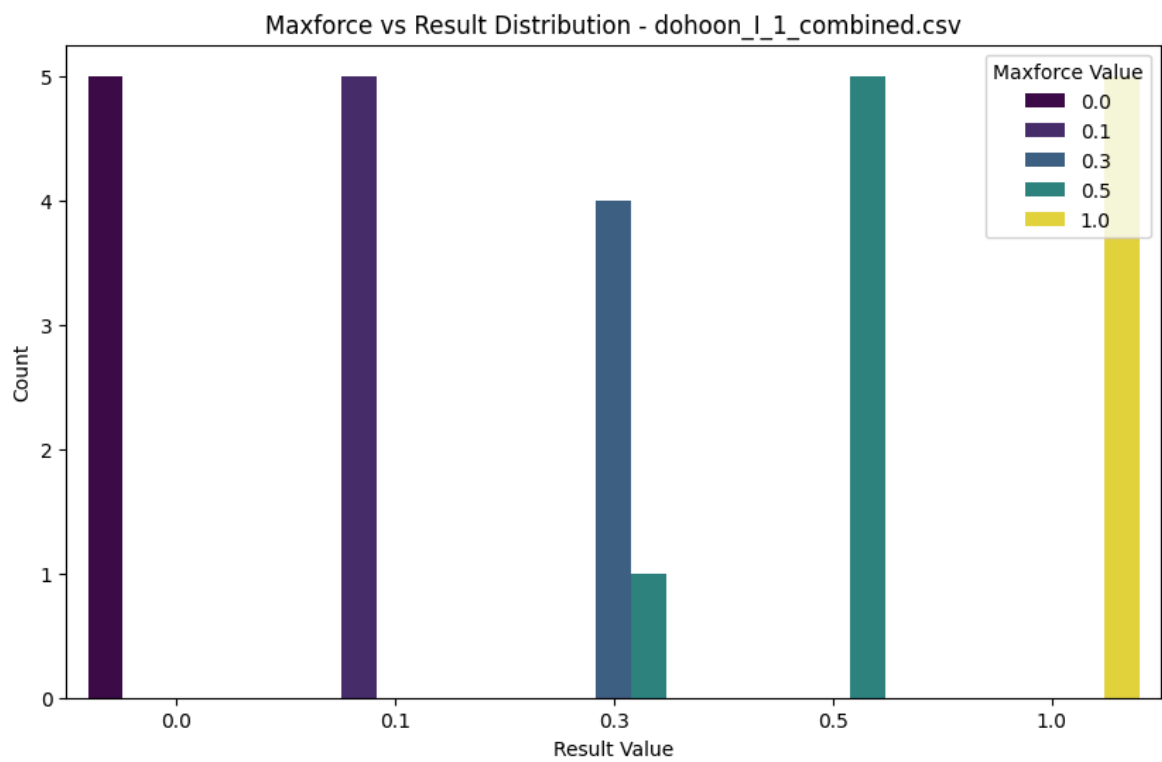
        # 막대그래프 생성
        plt.figure(figsize=(10, 6))
        sns.barplot(x='result_value', y='count', hue='maxforce_value', data=distribution_df)

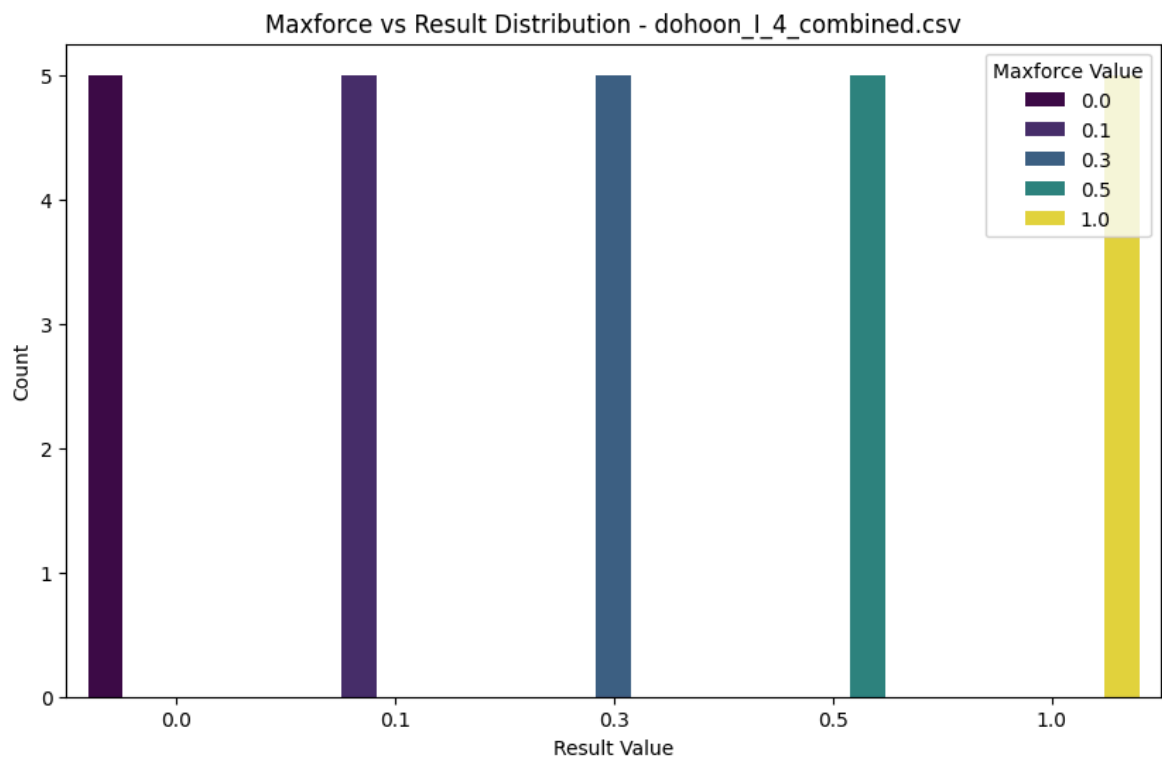
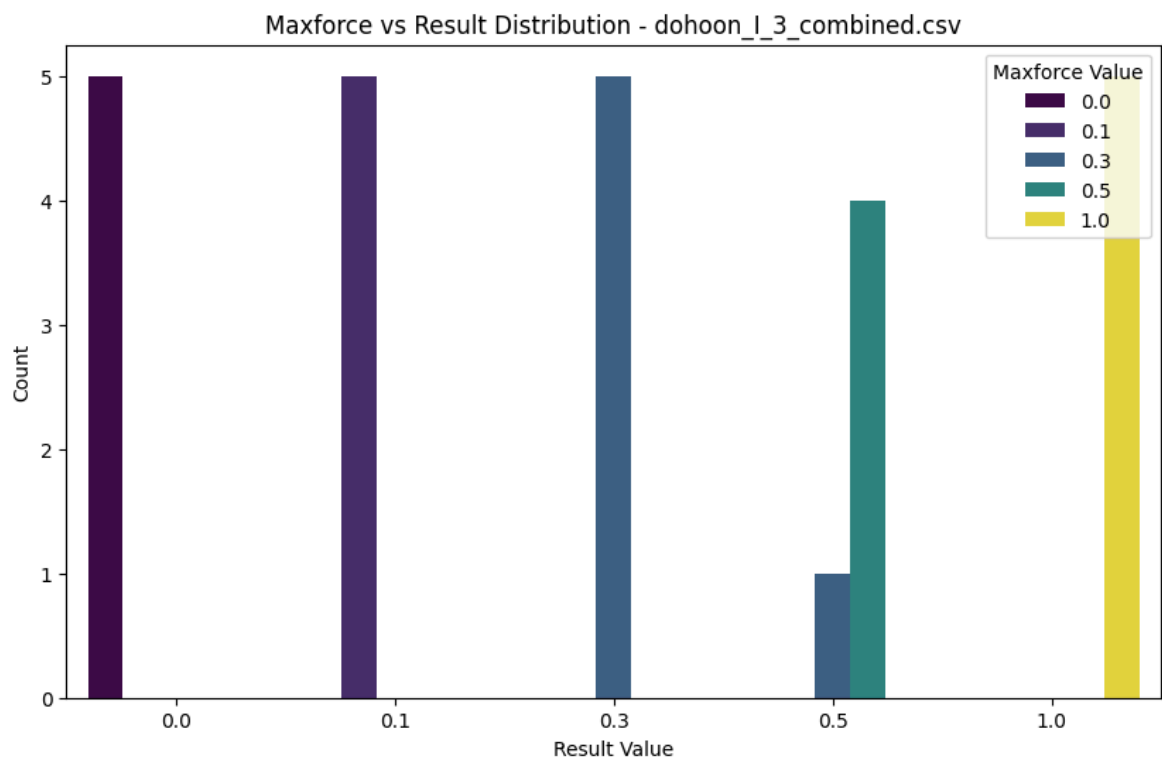
        # 그래프 세부 설정
        plt.title(f'Maxforce vs Result Distribution - {csv_file}')
        plt.xlabel('Result Value')
        plt.ylabel('Count')
        plt.legend(title='Maxforce Value')

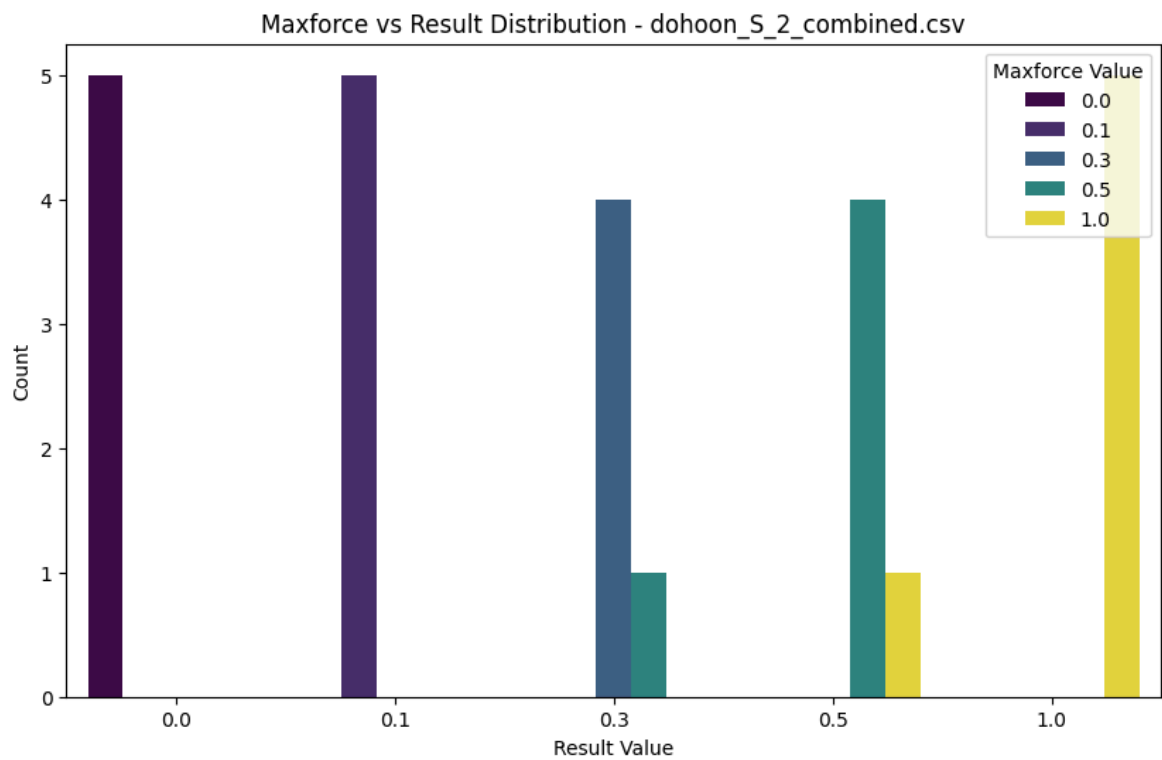
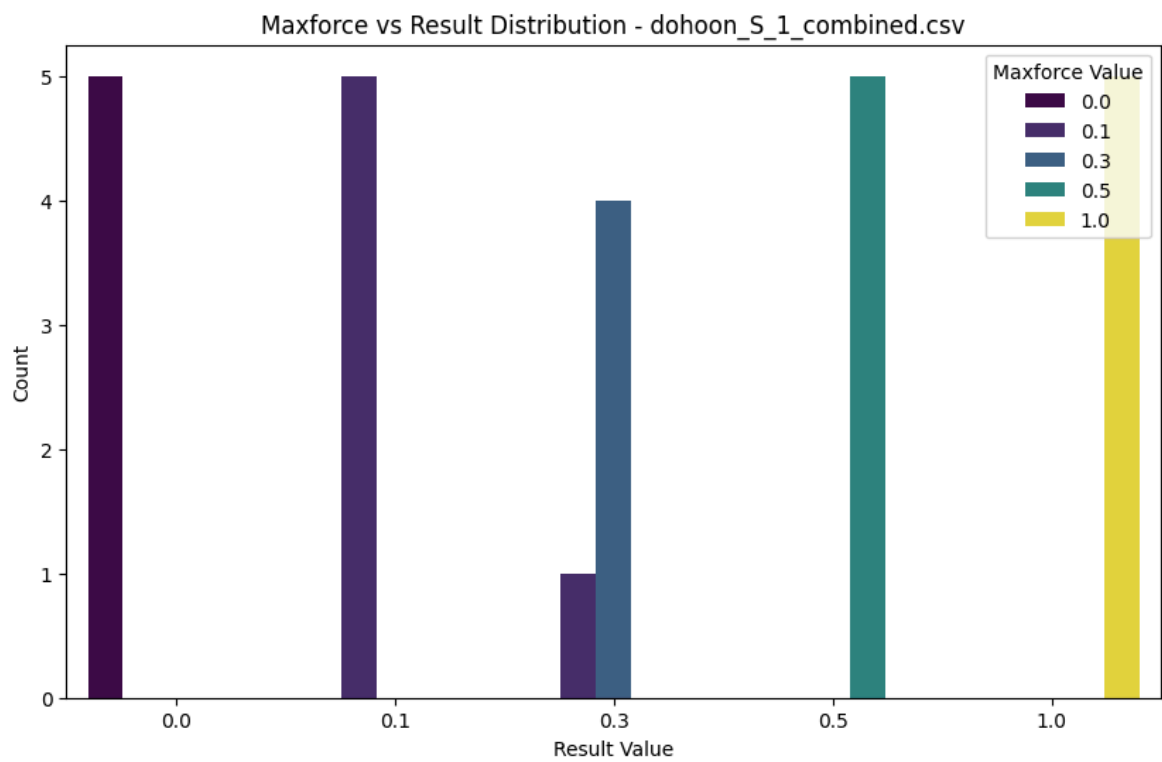
        # 그래프 출력
        plt.show()

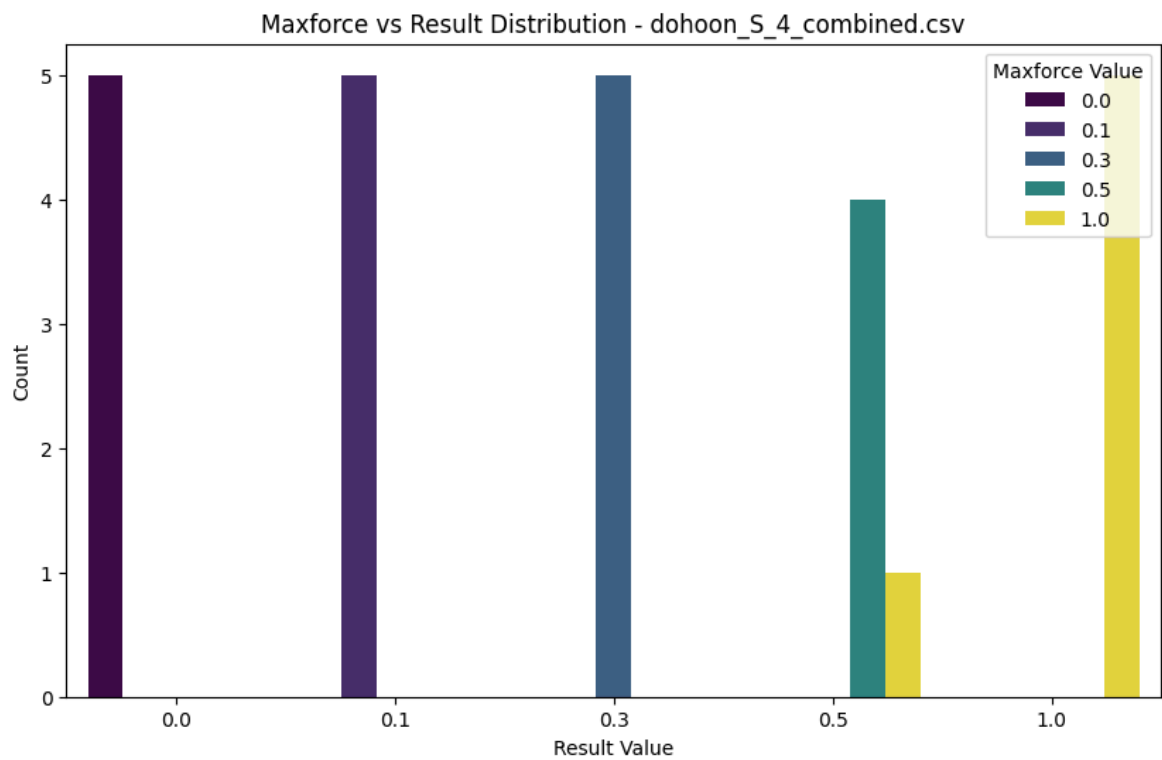
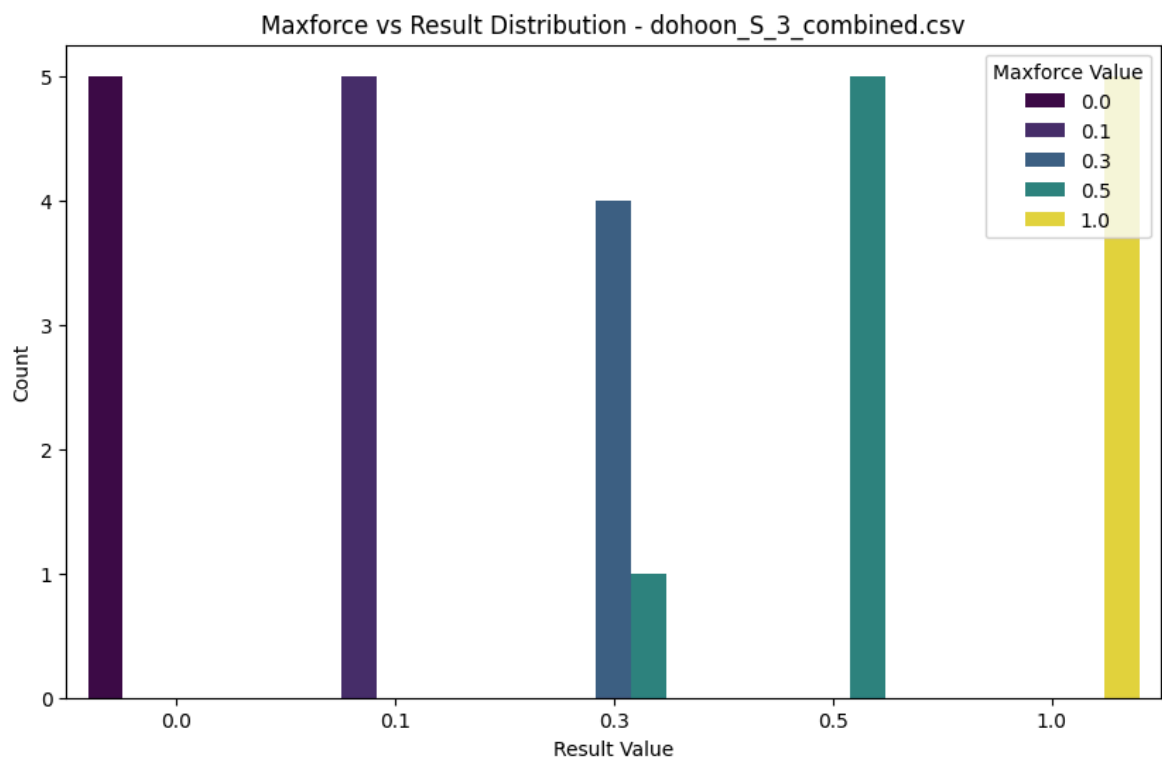
    else:
        print(f"'maxforce' or 'result' column missing in {csv_file}")

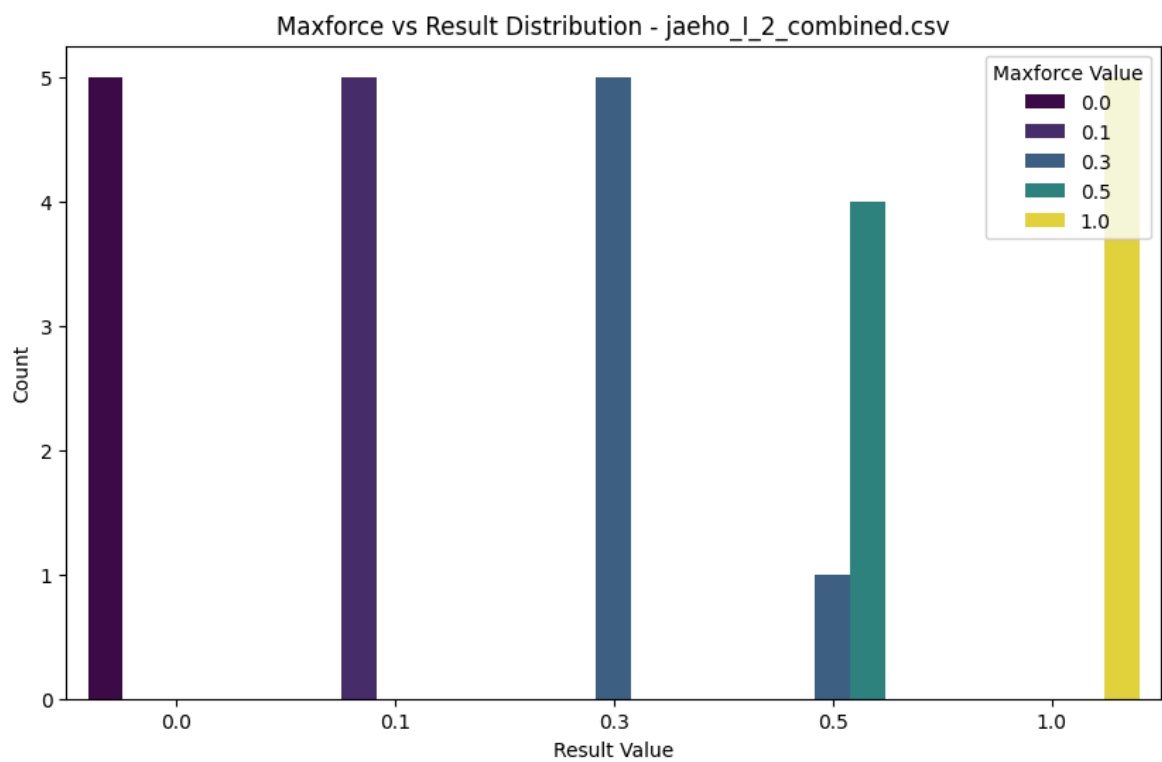
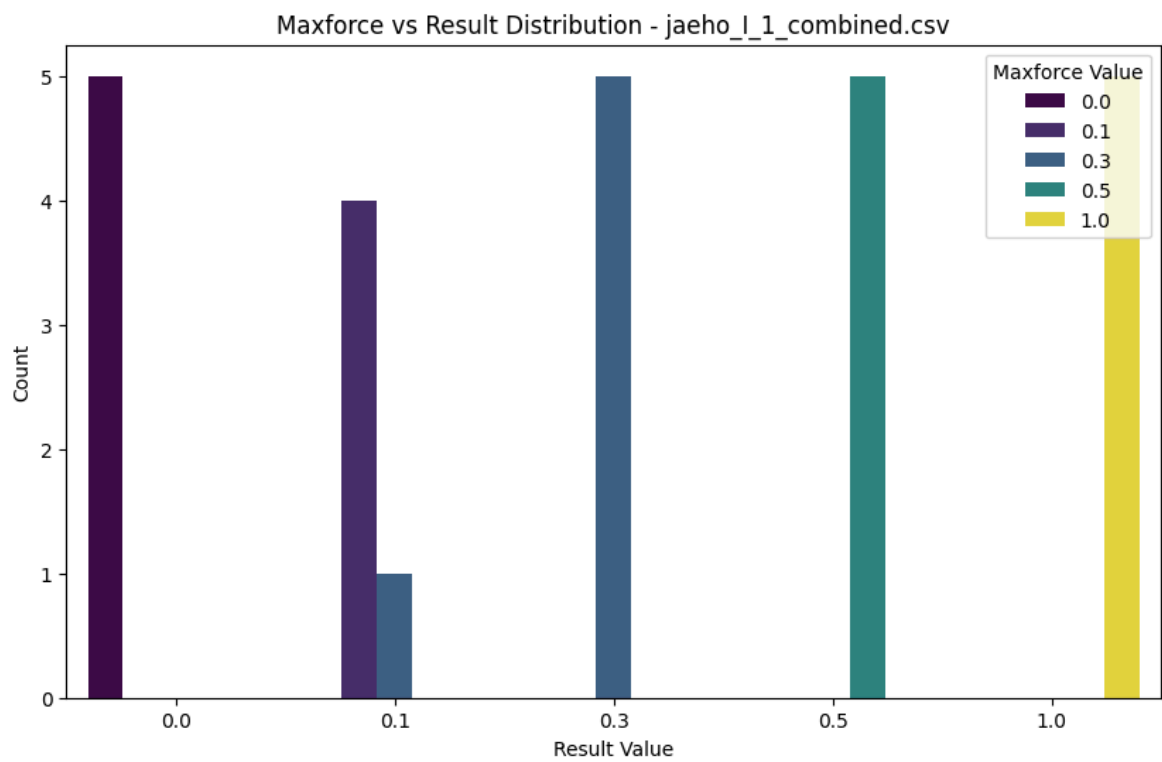
```

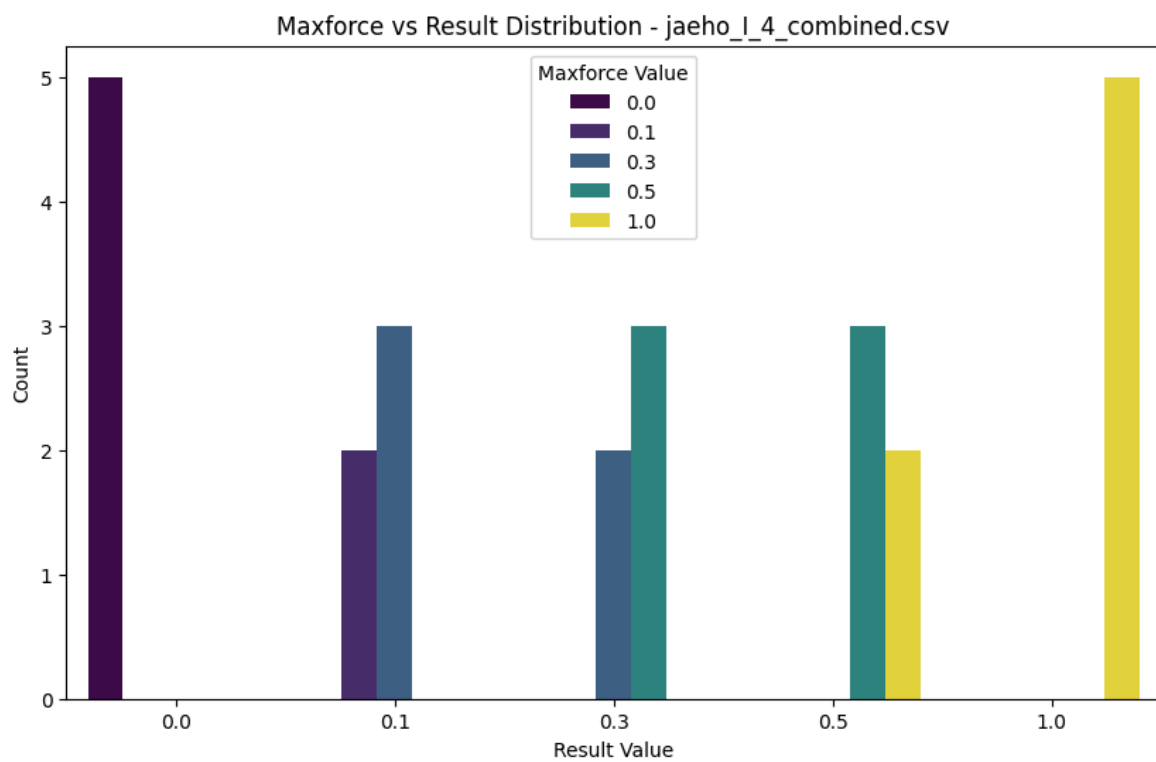
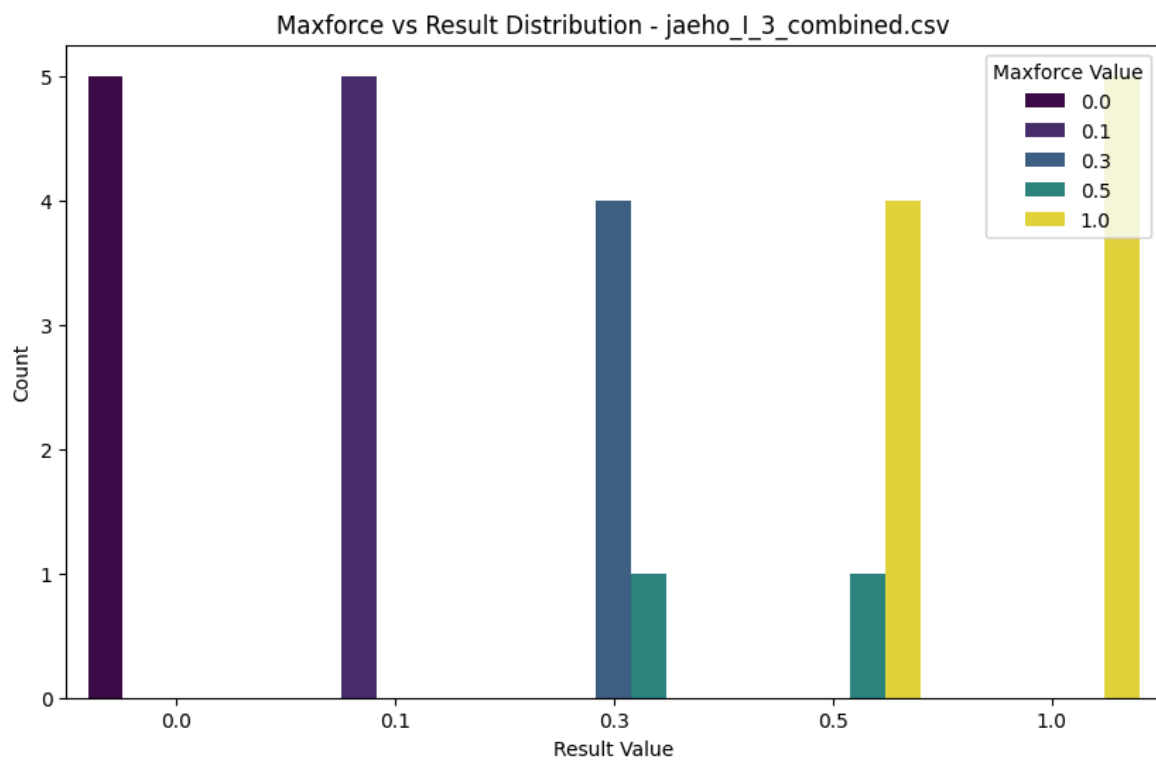


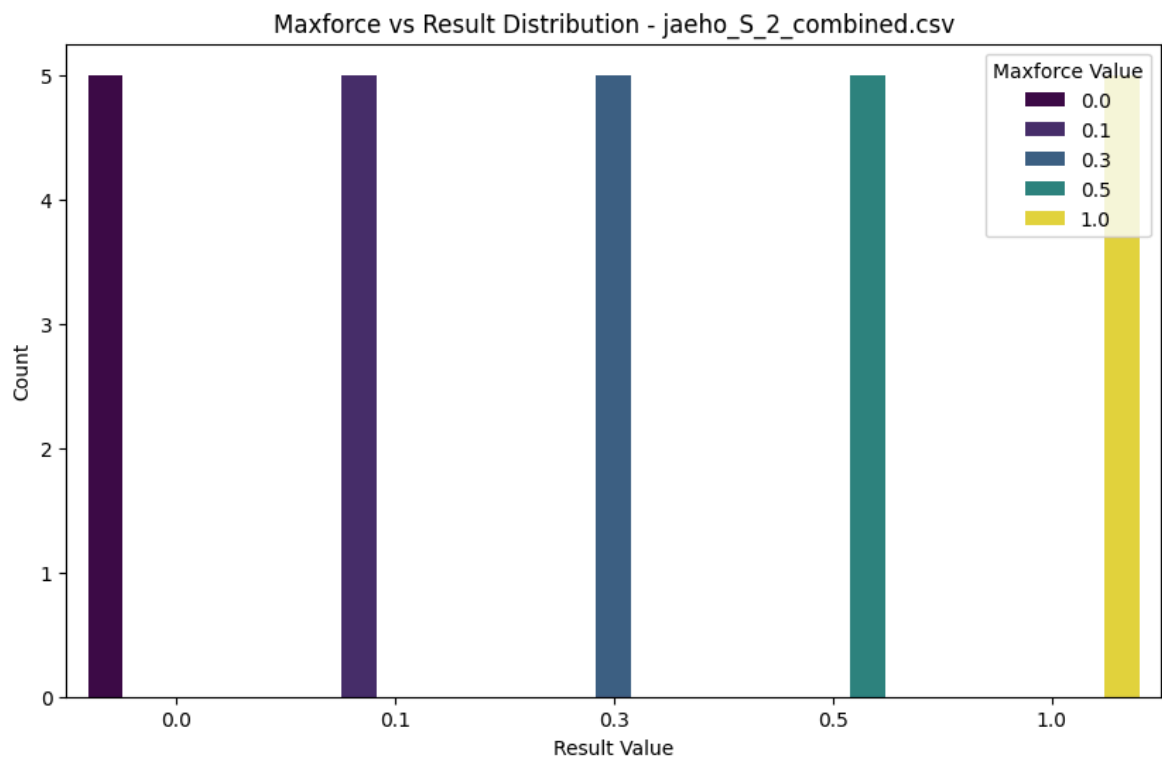
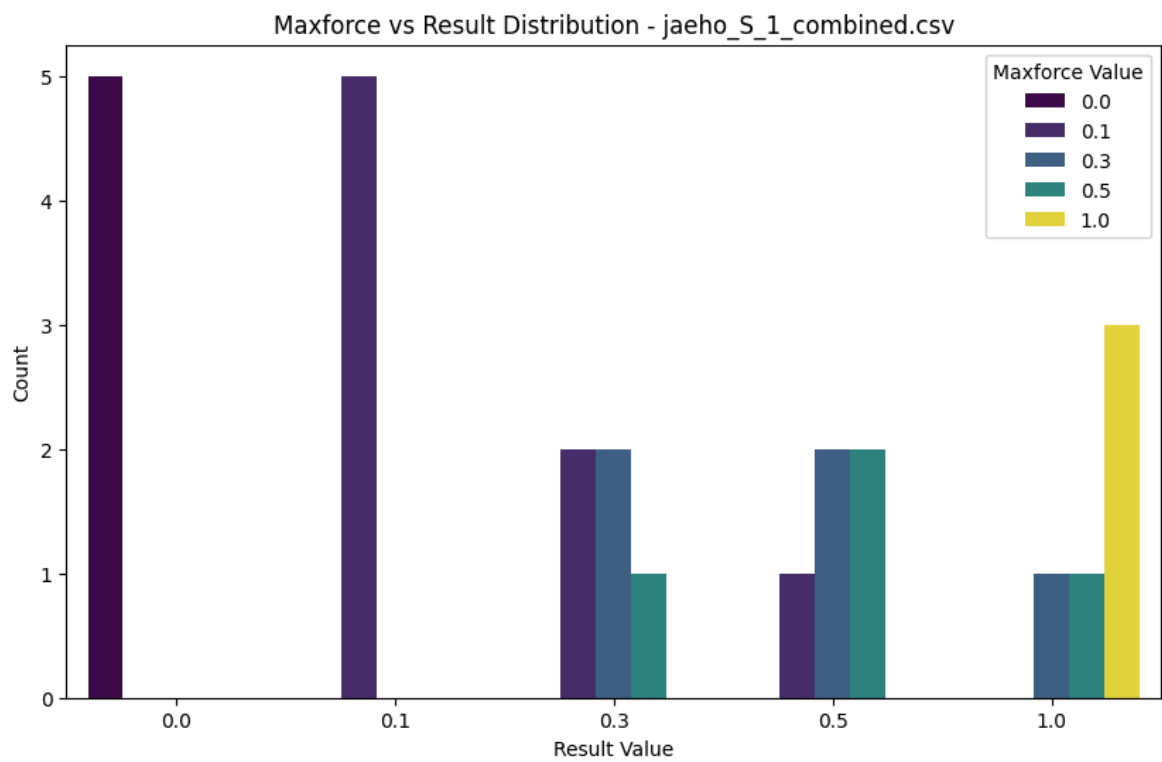




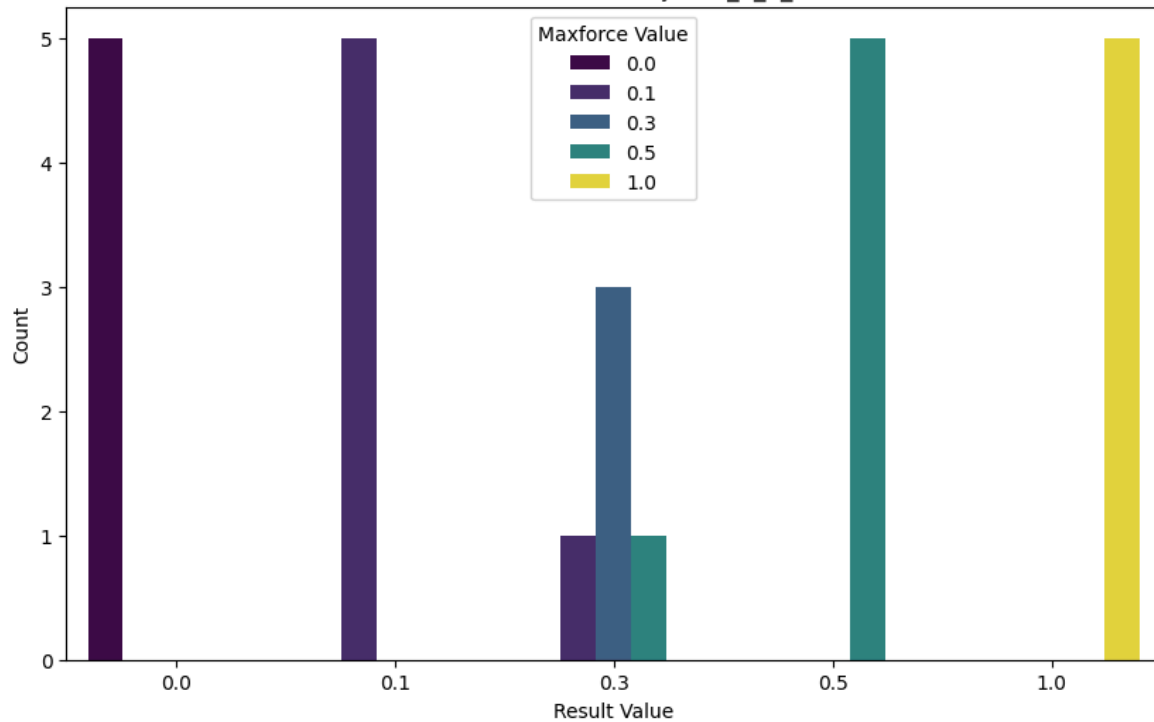




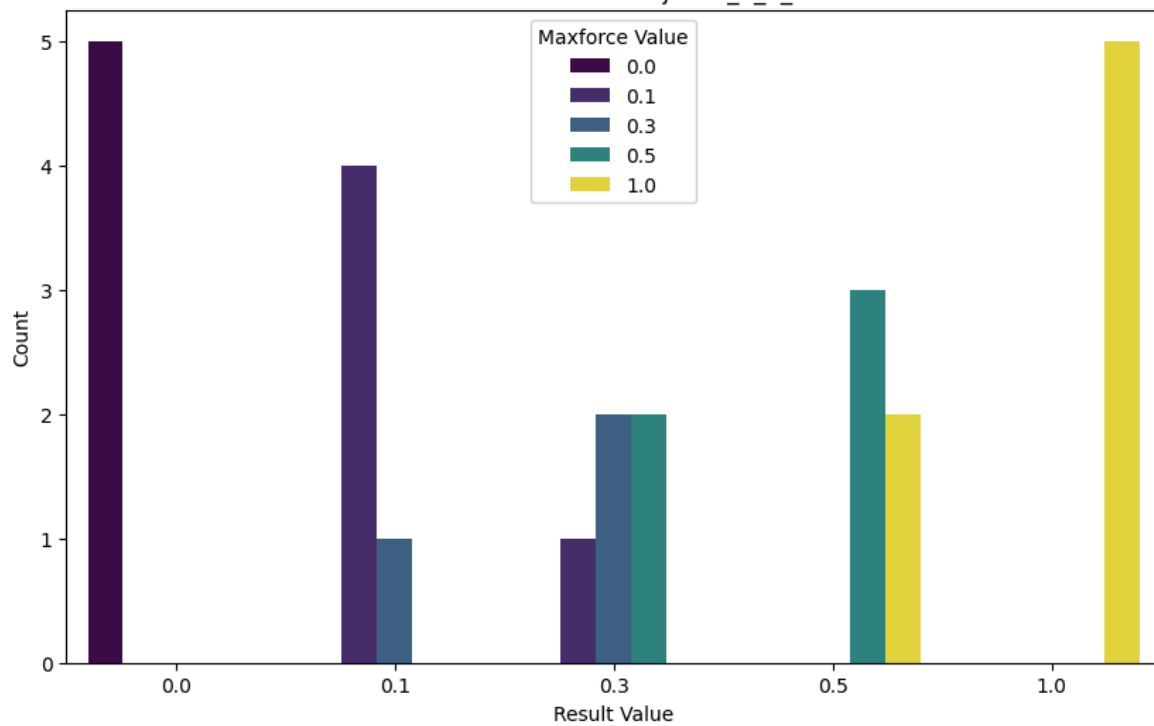




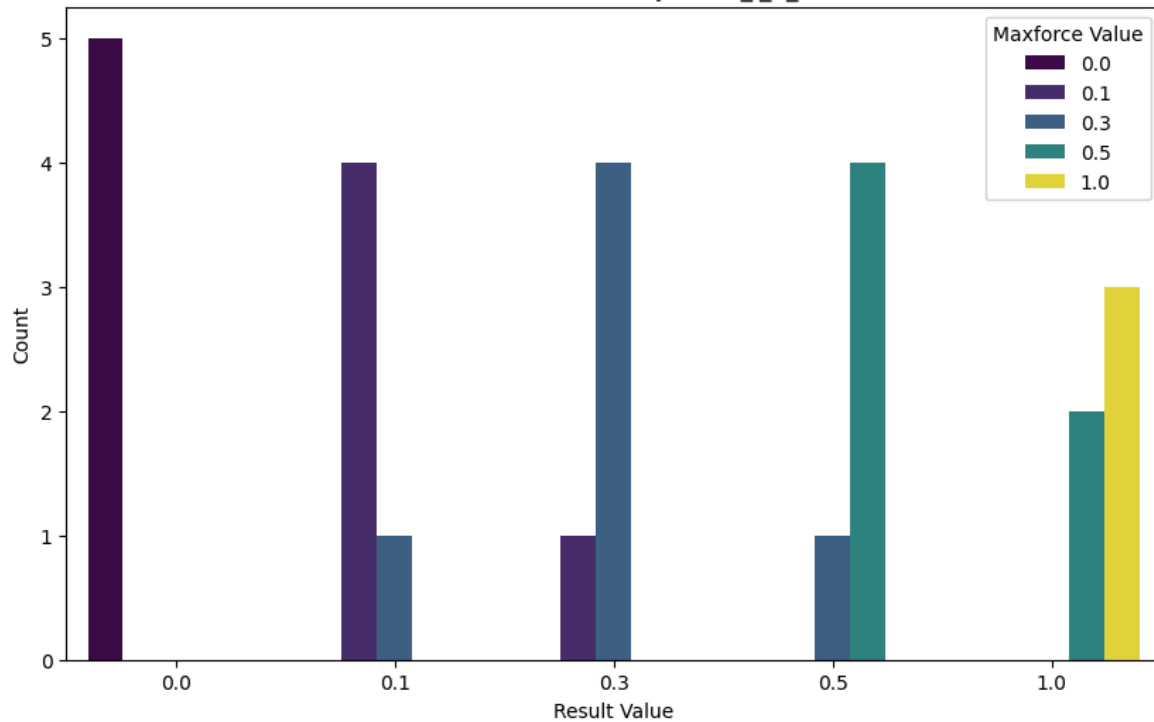
Maxforce vs Result Distribution - jaeho_S_3_combined.csv



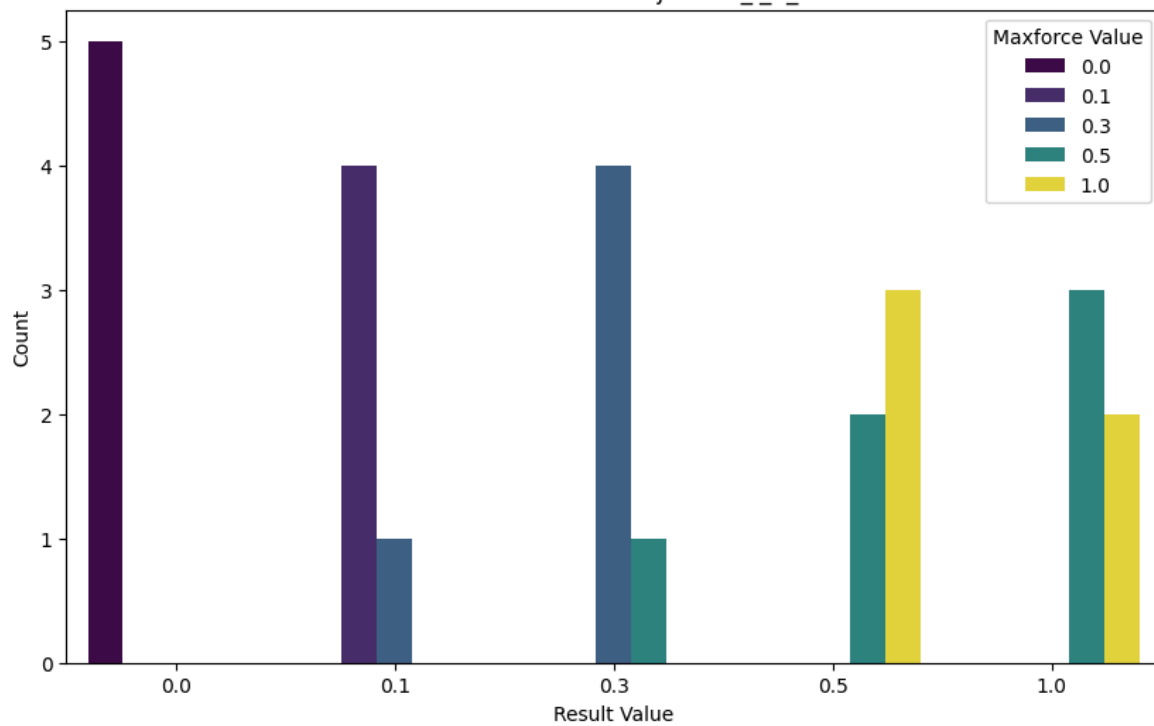
Maxforce vs Result Distribution - jaeho_S_4_combined.csv



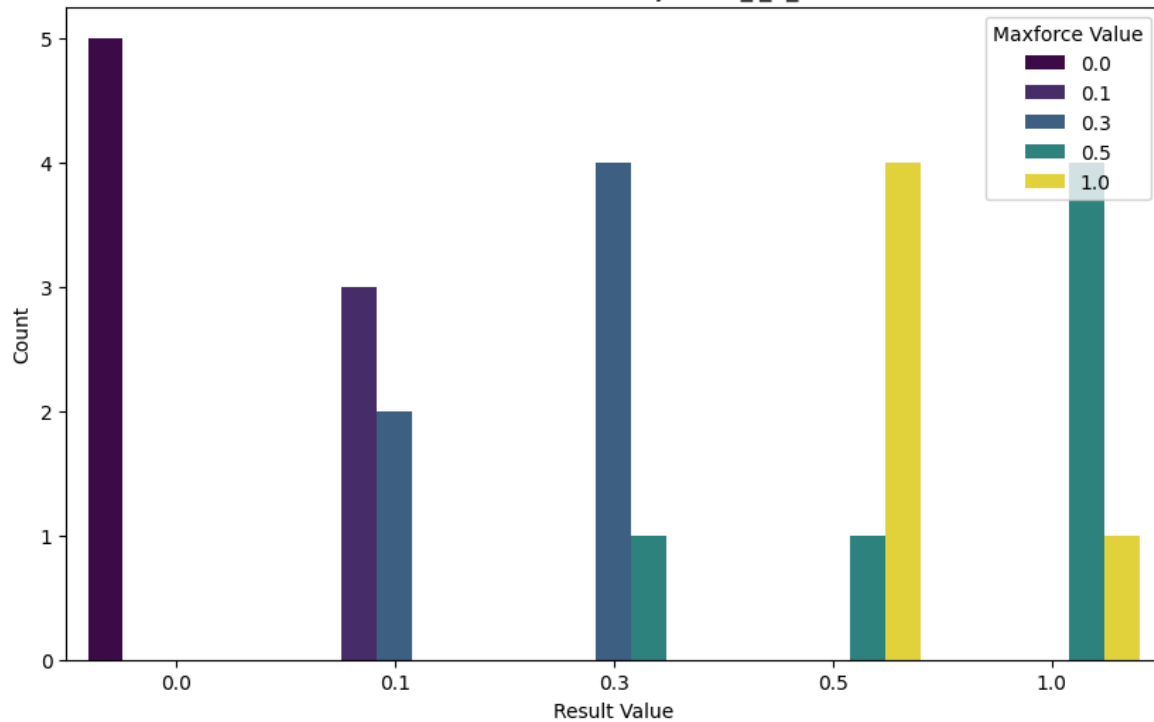
Maxforce vs Result Distribution - jaewan_l_1_combined.csv



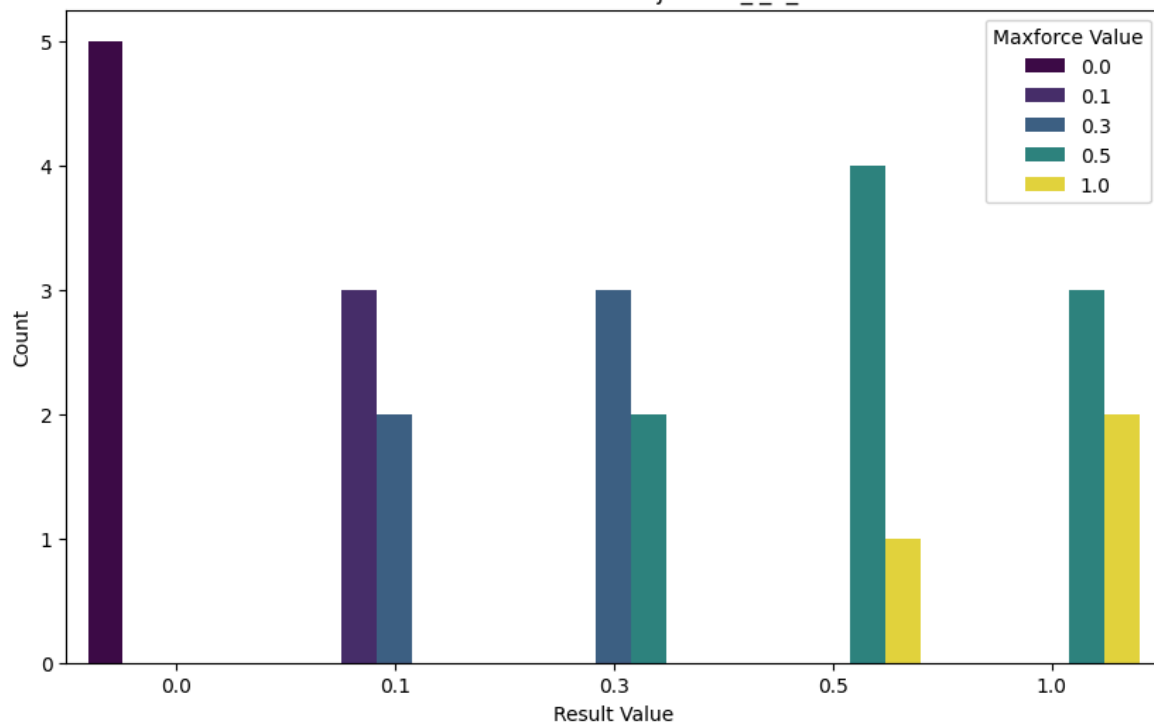
Maxforce vs Result Distribution - jaewan_l_2_combined.csv



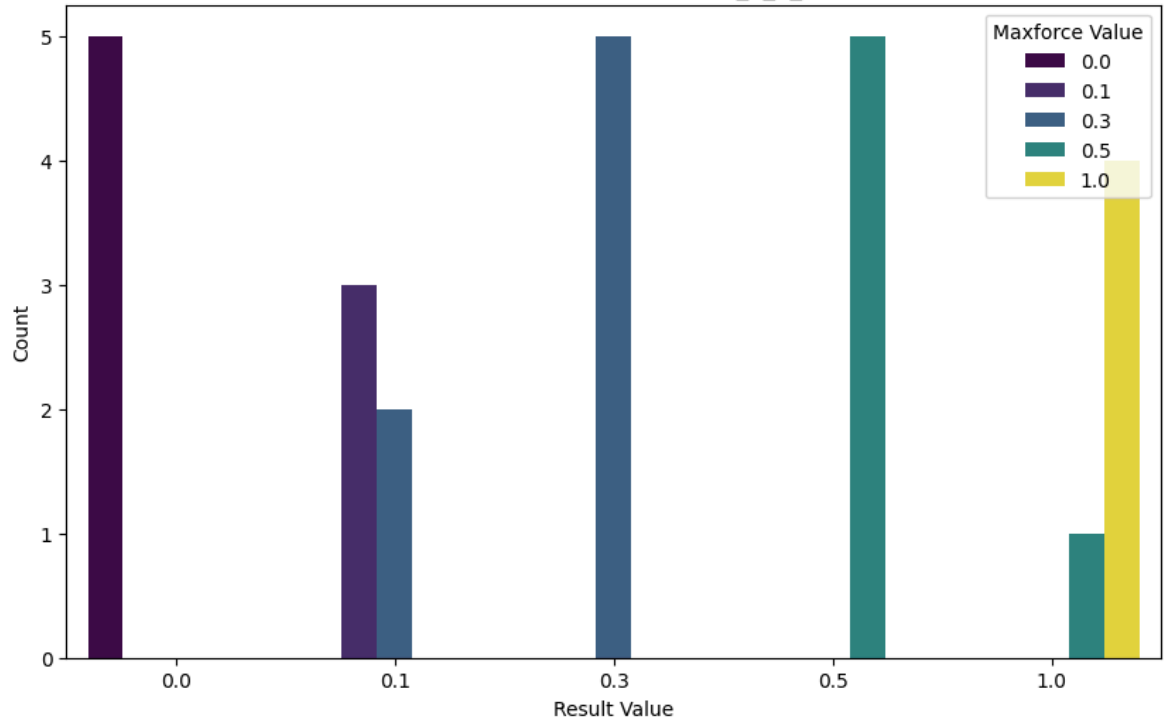
Maxforce vs Result Distribution - jaewan_l_3_combined.csv



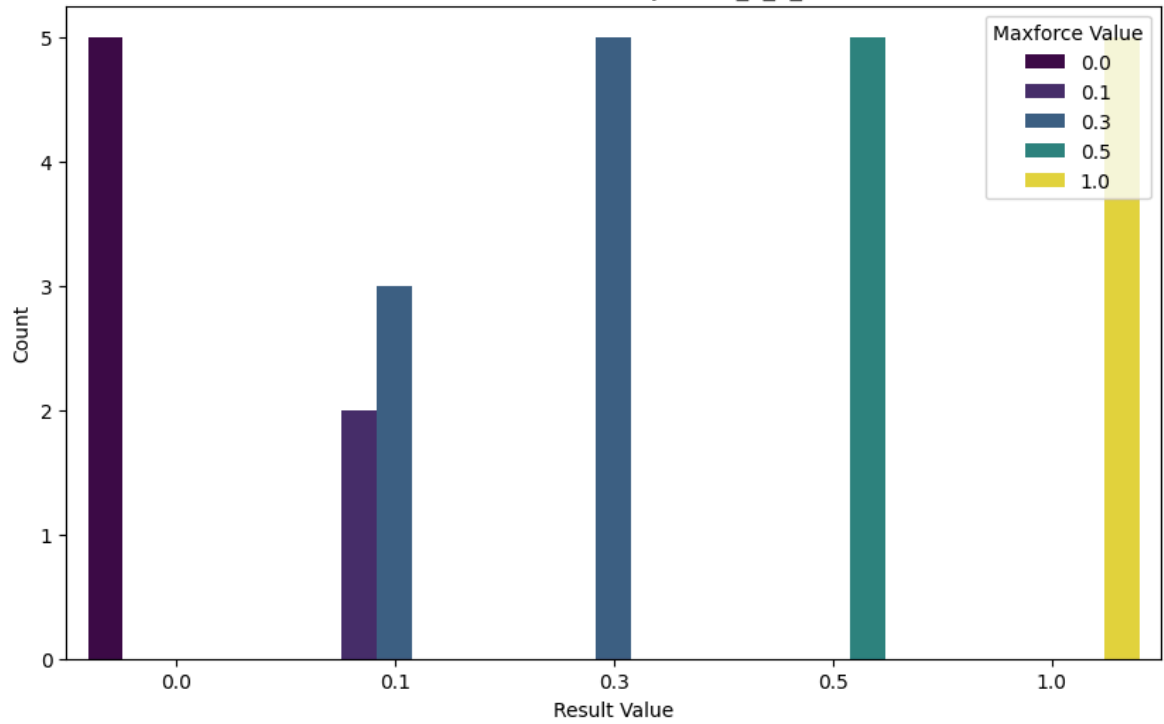
Maxforce vs Result Distribution - jaewan_l_4_combined.csv



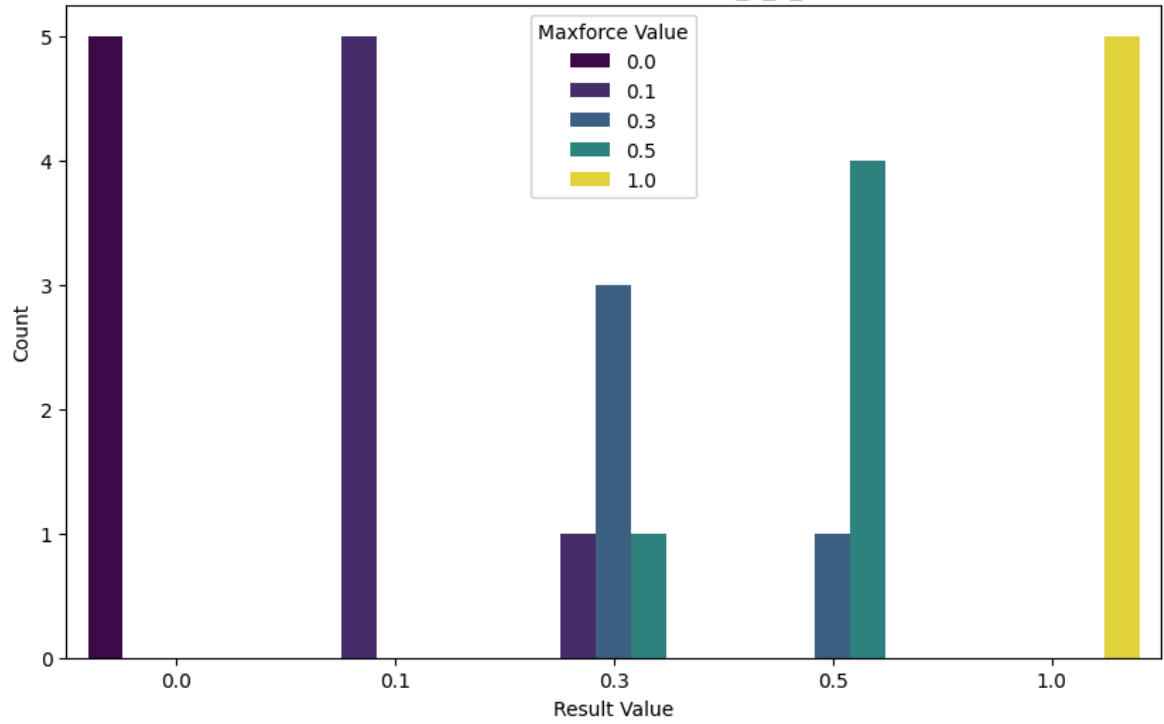
Maxforce vs Result Distribution - jaewan_S_1_combined.csv



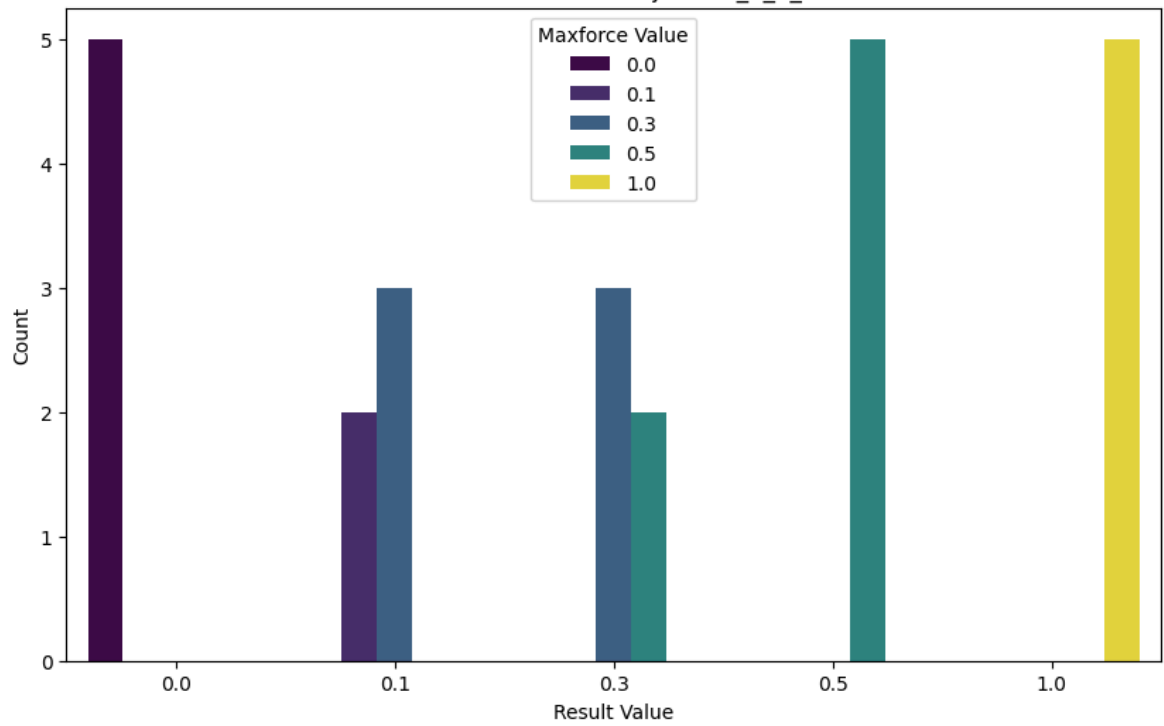
Maxforce vs Result Distribution - jaewan_S_2_combined.csv

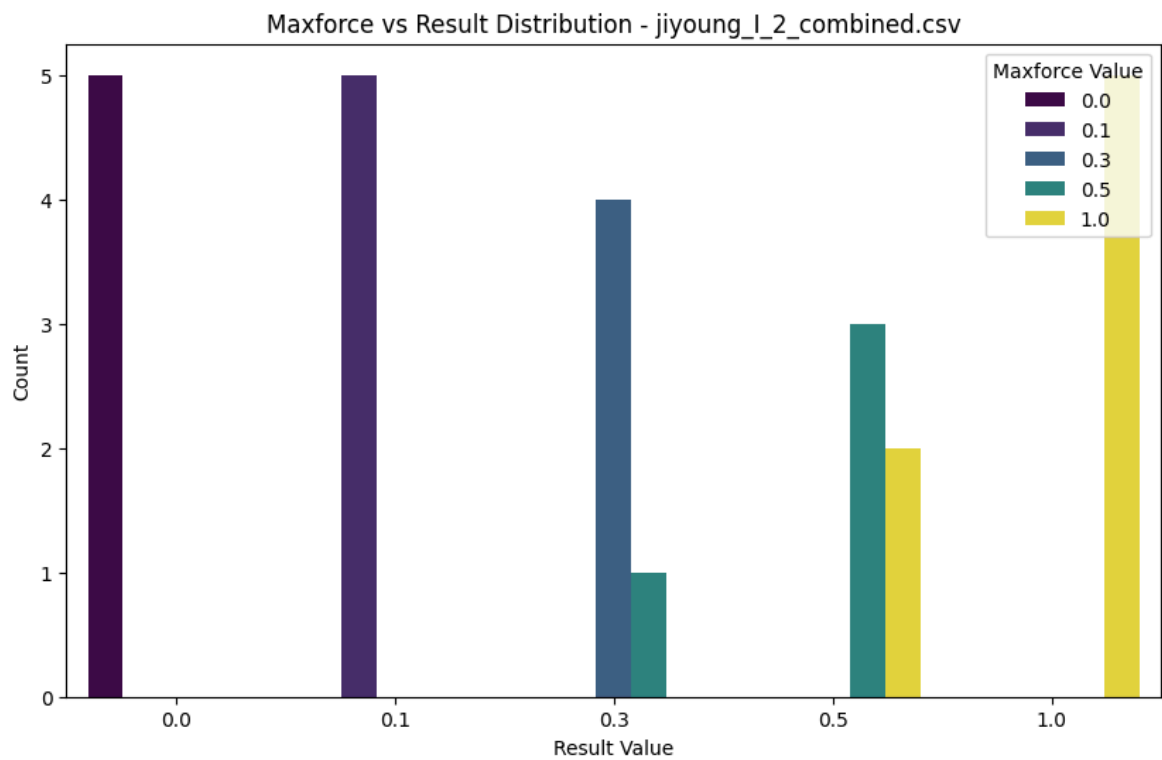
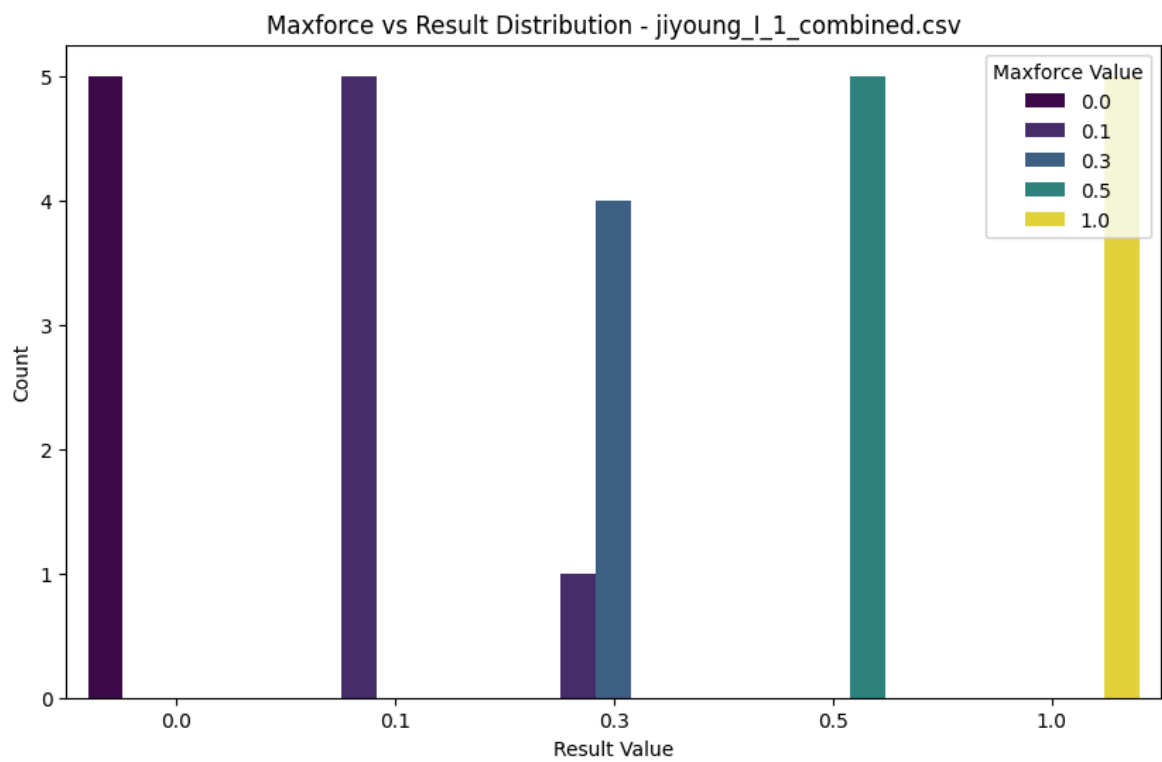


Maxforce vs Result Distribution - jaewan_S_3_combined.csv

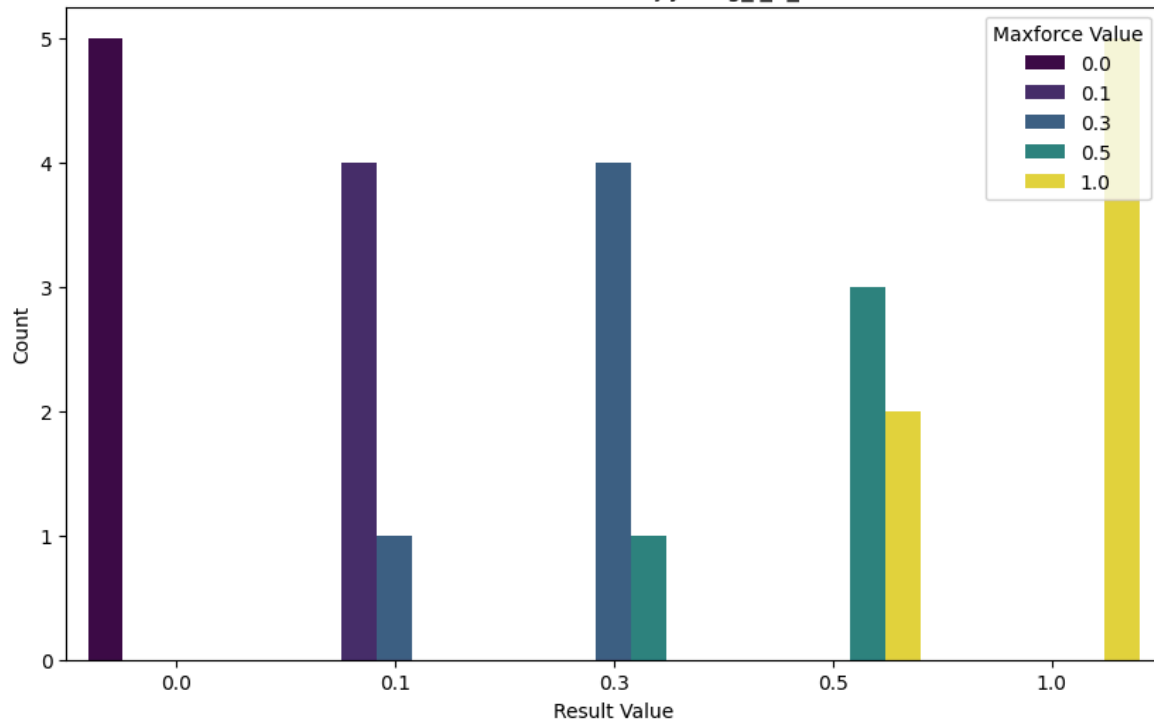


Maxforce vs Result Distribution - jaewan_S_4_combined.csv

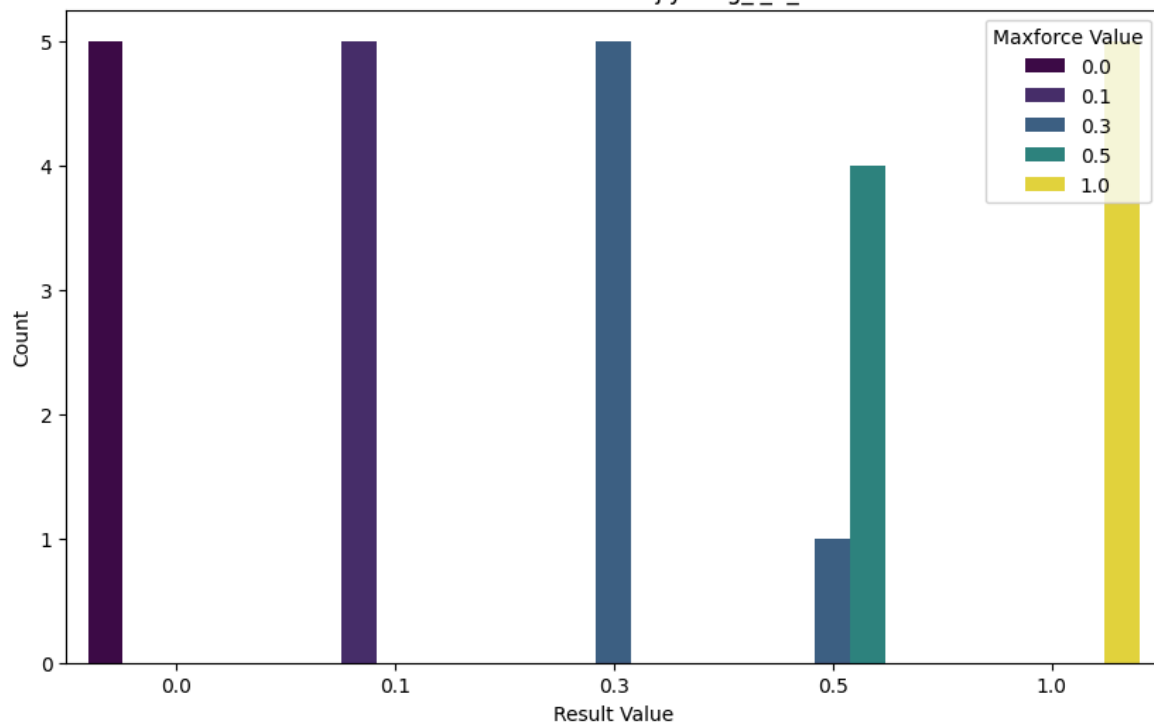




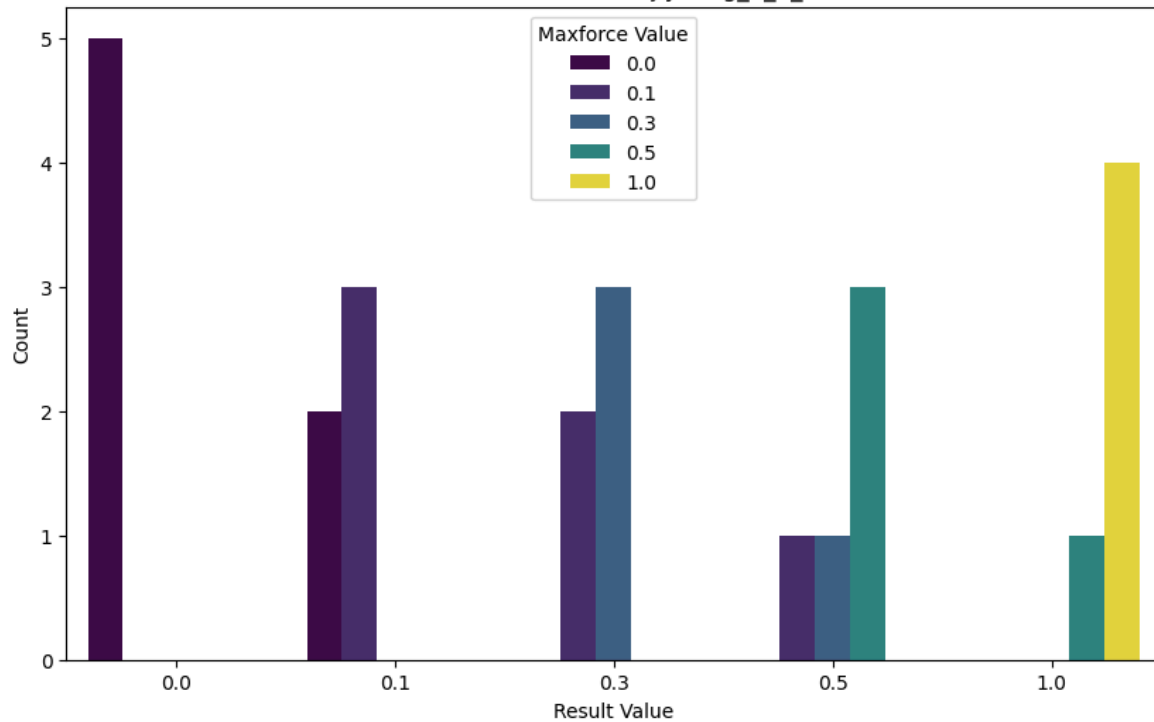
Maxforce vs Result Distribution - jiyoung_I_3_combined.csv



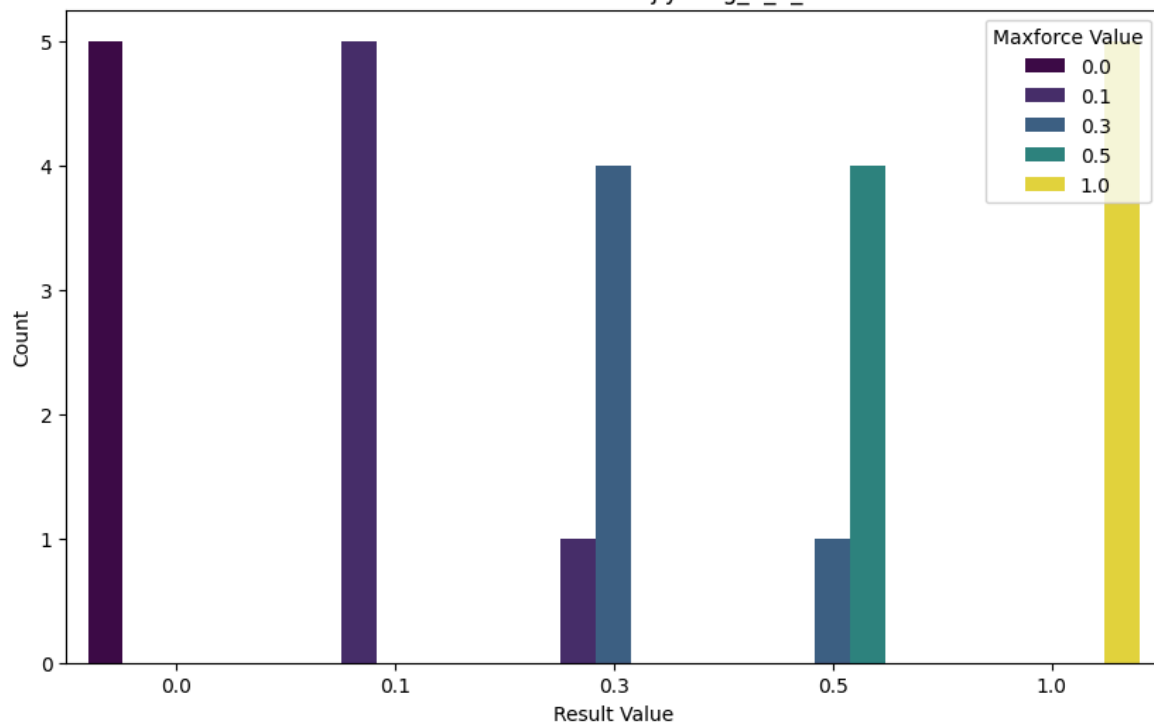
Maxforce vs Result Distribution - jiyoung_I_4_combined.csv



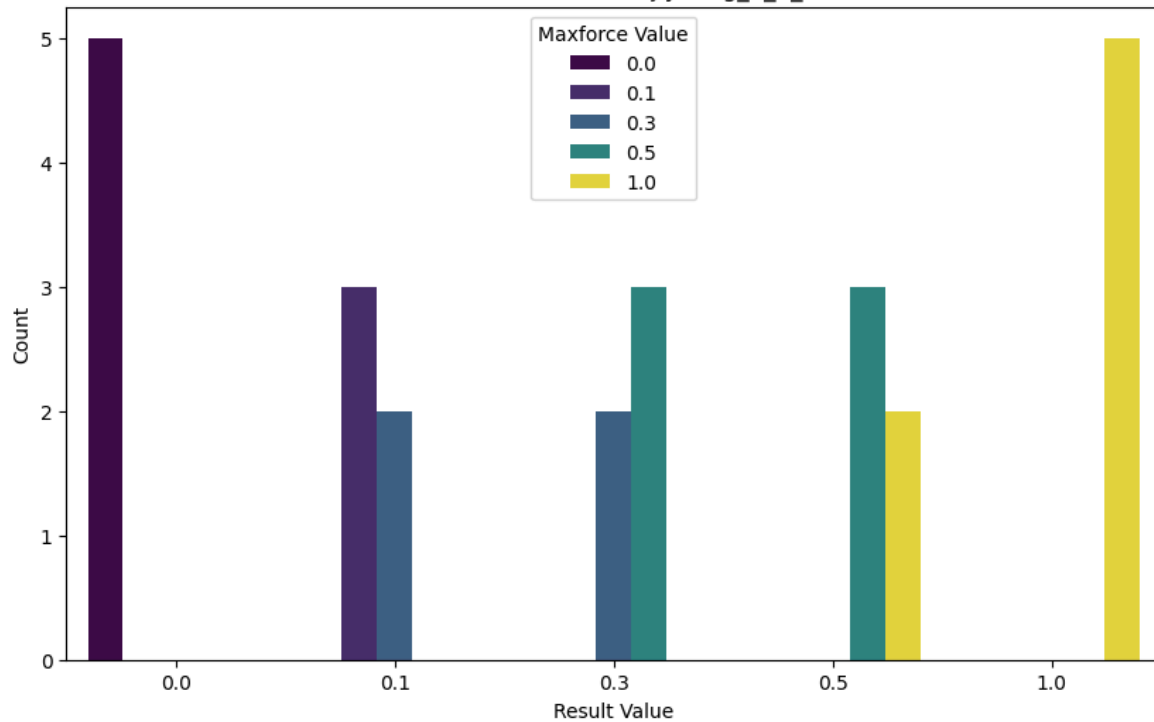
Maxforce vs Result Distribution - jiyoung_S_1_combined.csv



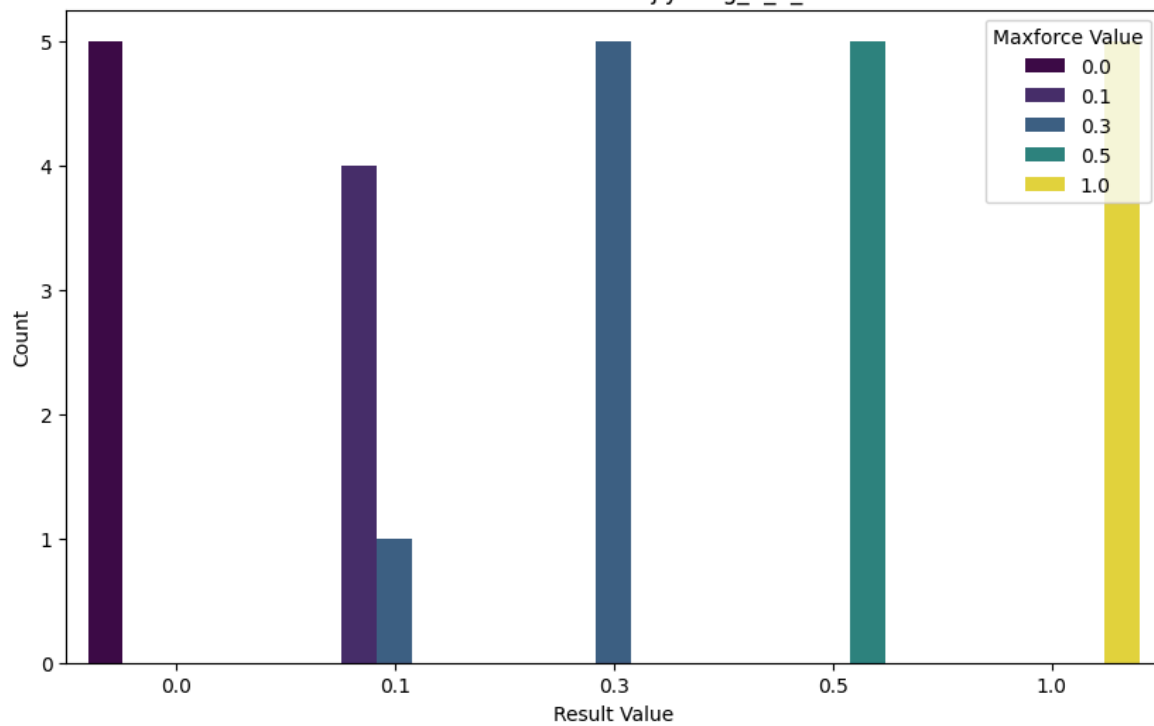
Maxforce vs Result Distribution - jiyoung_S_2_combined.csv



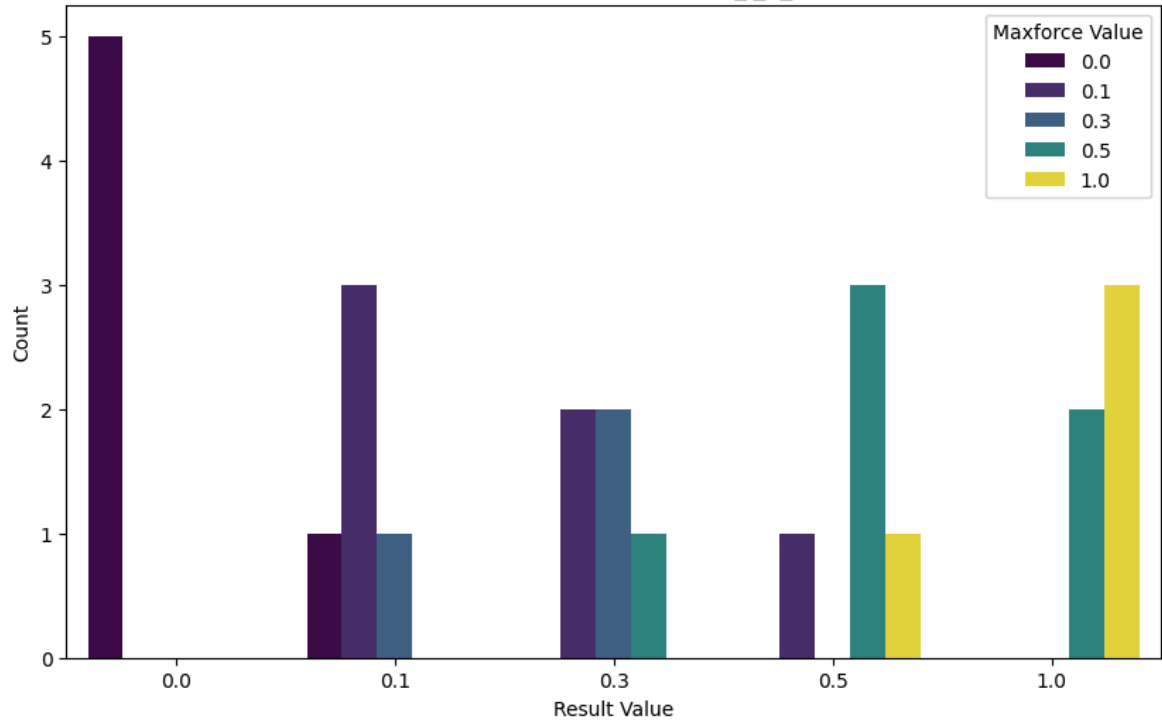
Maxforce vs Result Distribution - jiyoung_S_3_combined.csv



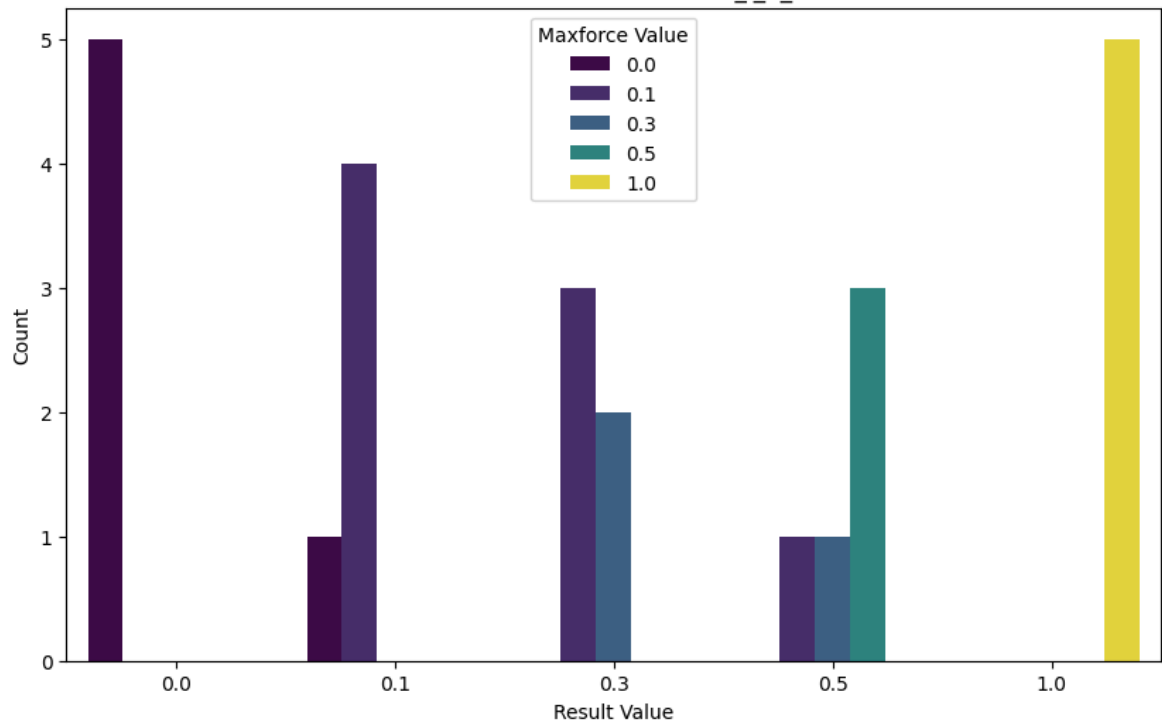
Maxforce vs Result Distribution - jiyoung_S_4_combined.csv



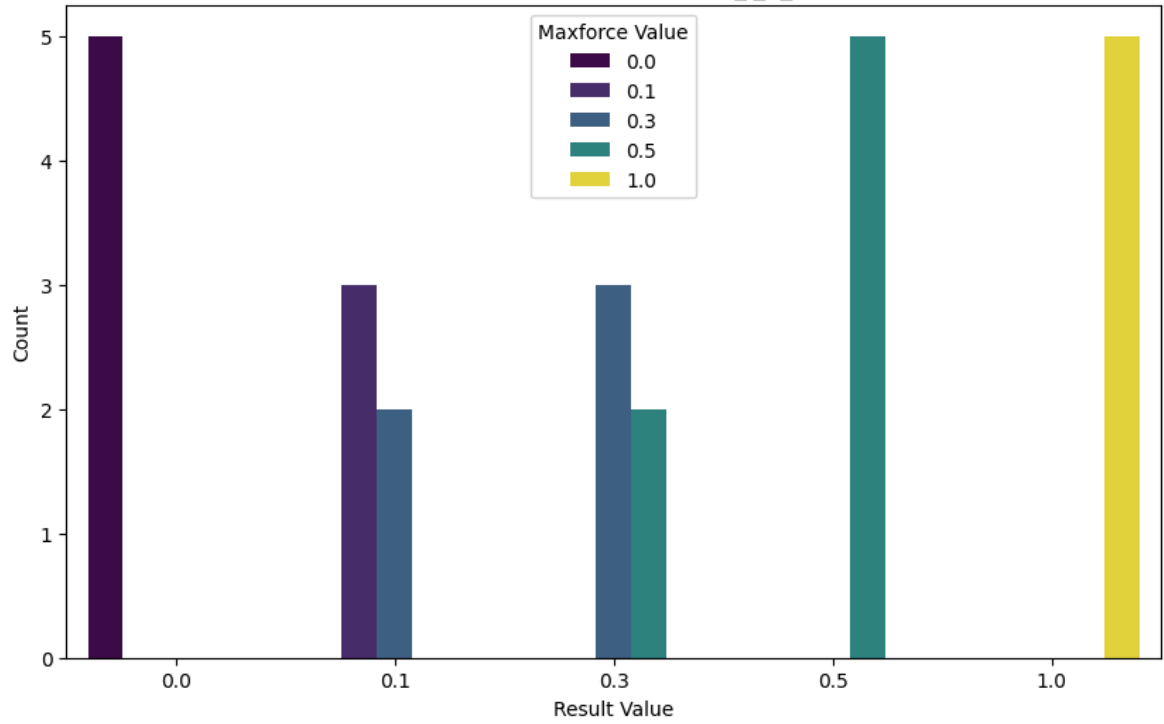
Maxforce vs Result Distribution - minho_I_1_combined.csv



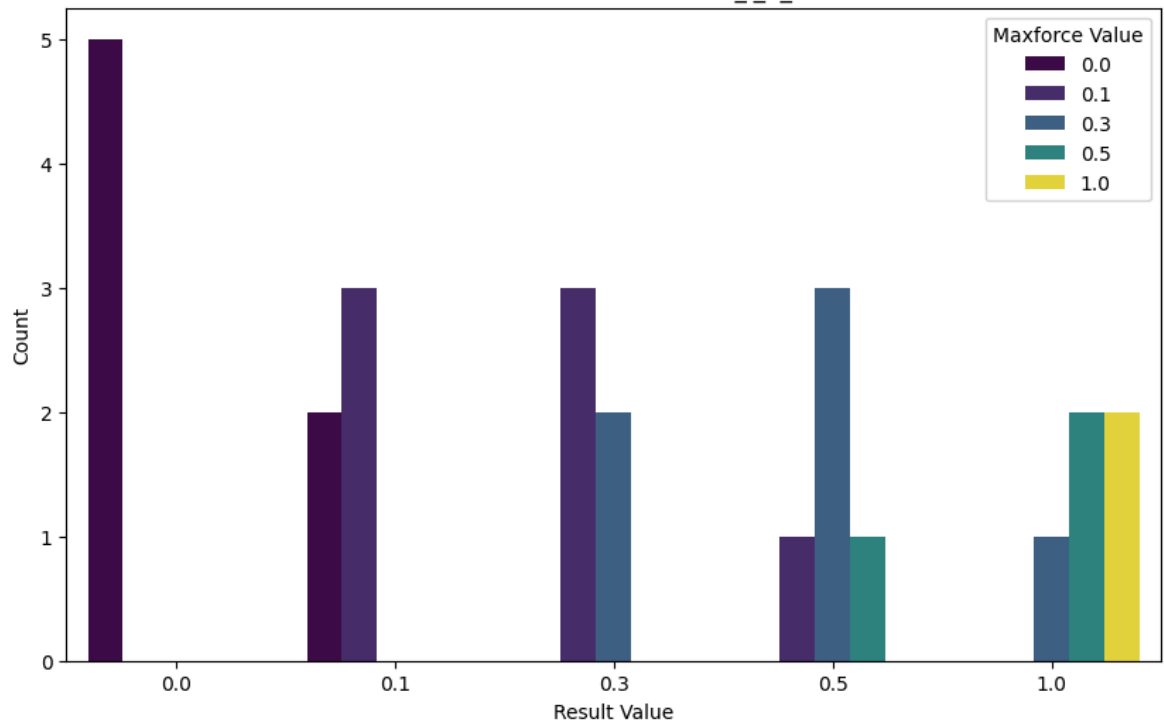
Maxforce vs Result Distribution - minho_I_2_combined.csv

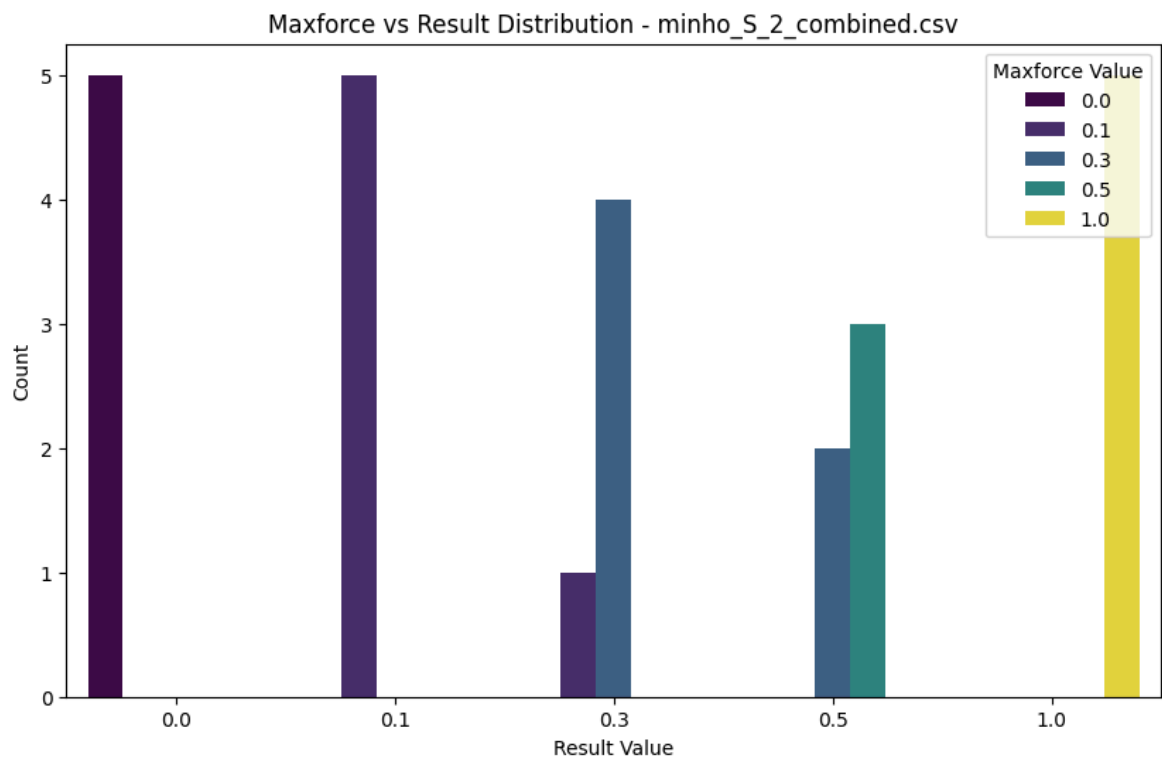
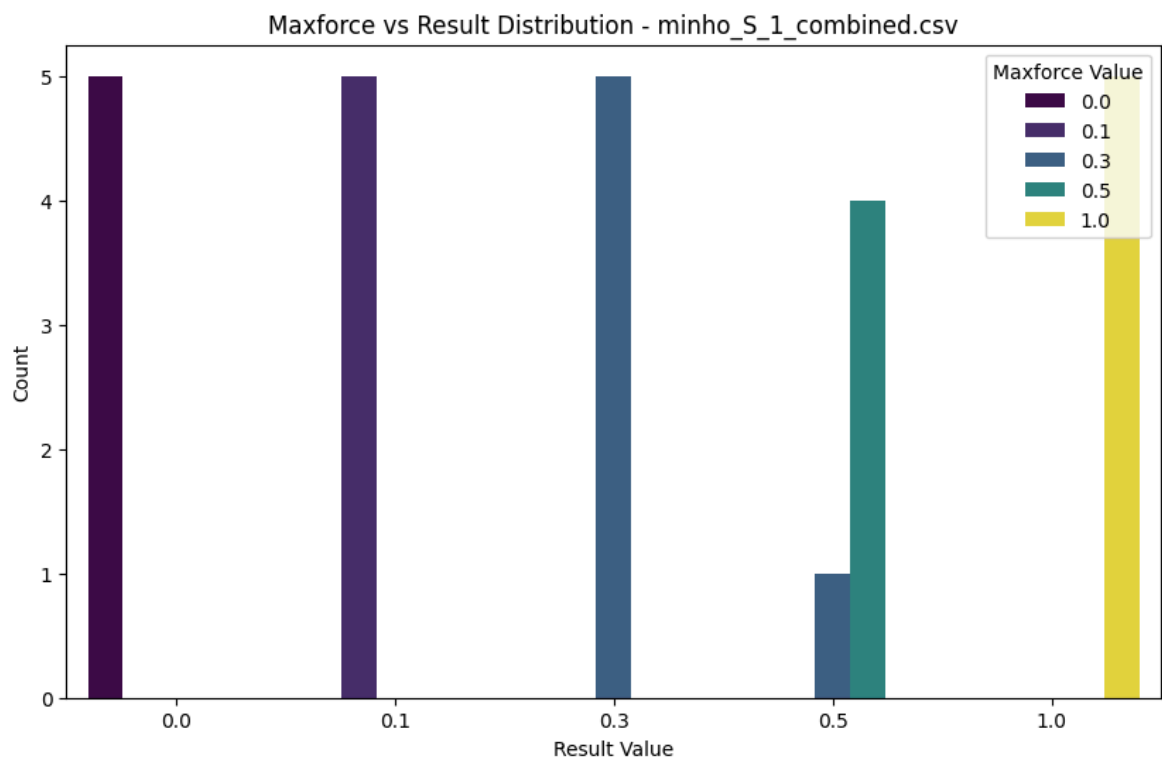


Maxforce vs Result Distribution - minho_I_3_combined.csv

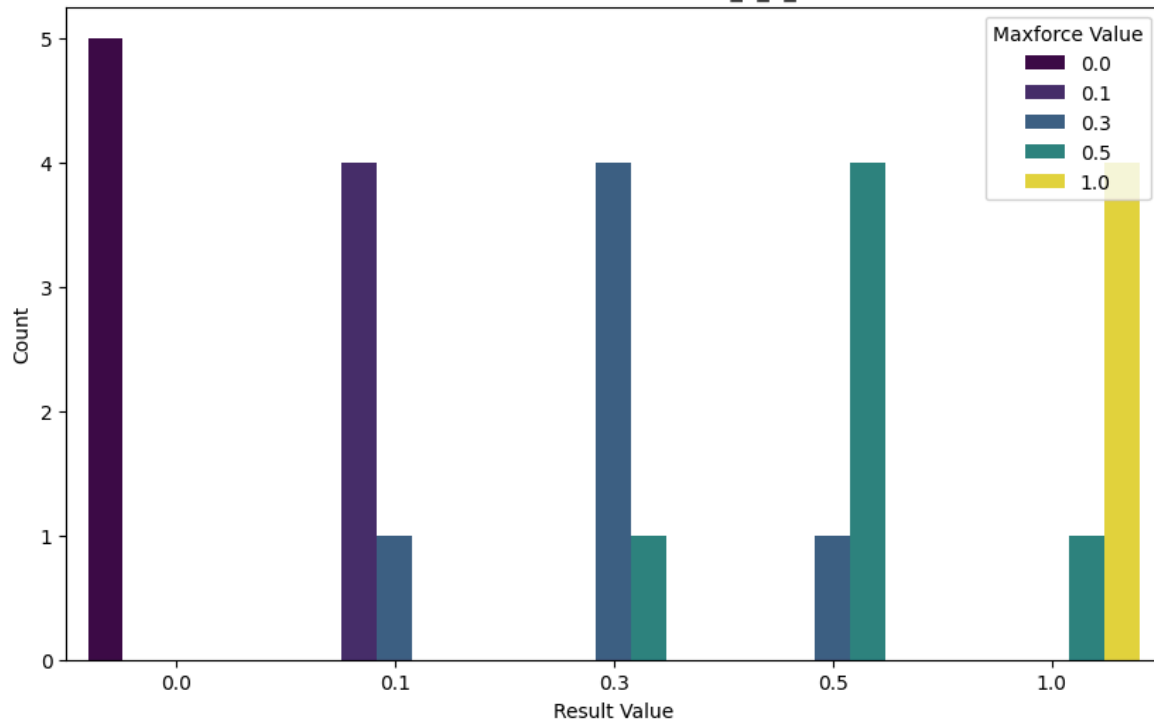


Maxforce vs Result Distribution - minho_I_4_combined.csv

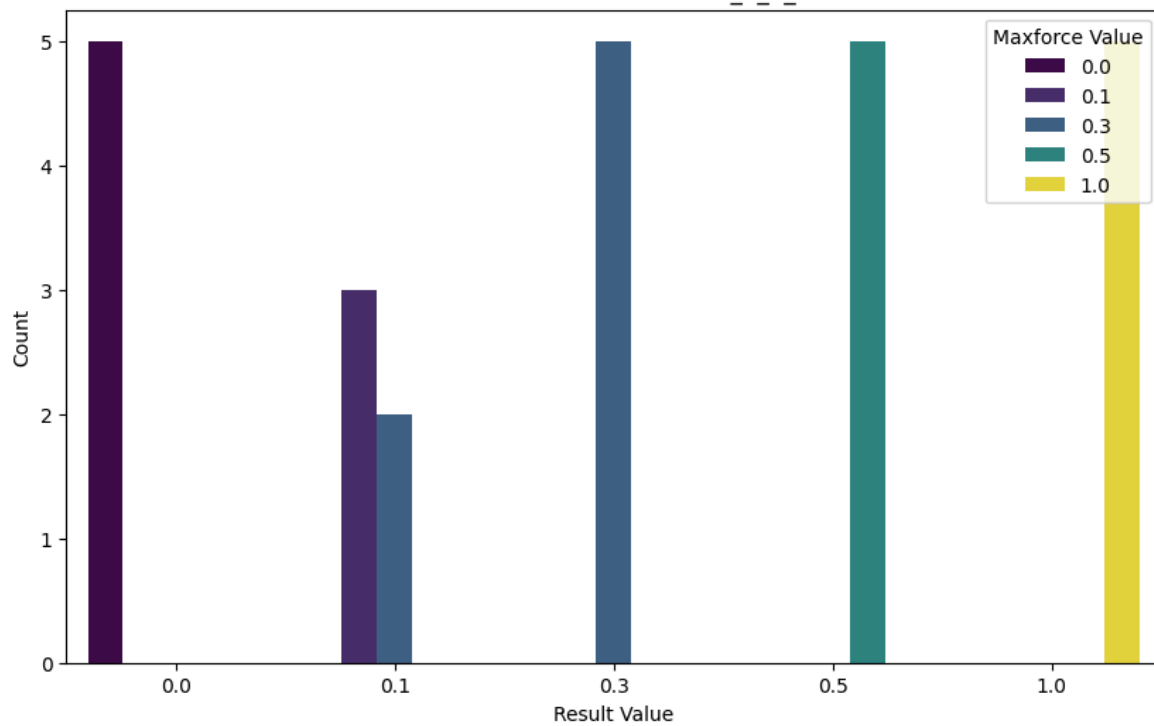




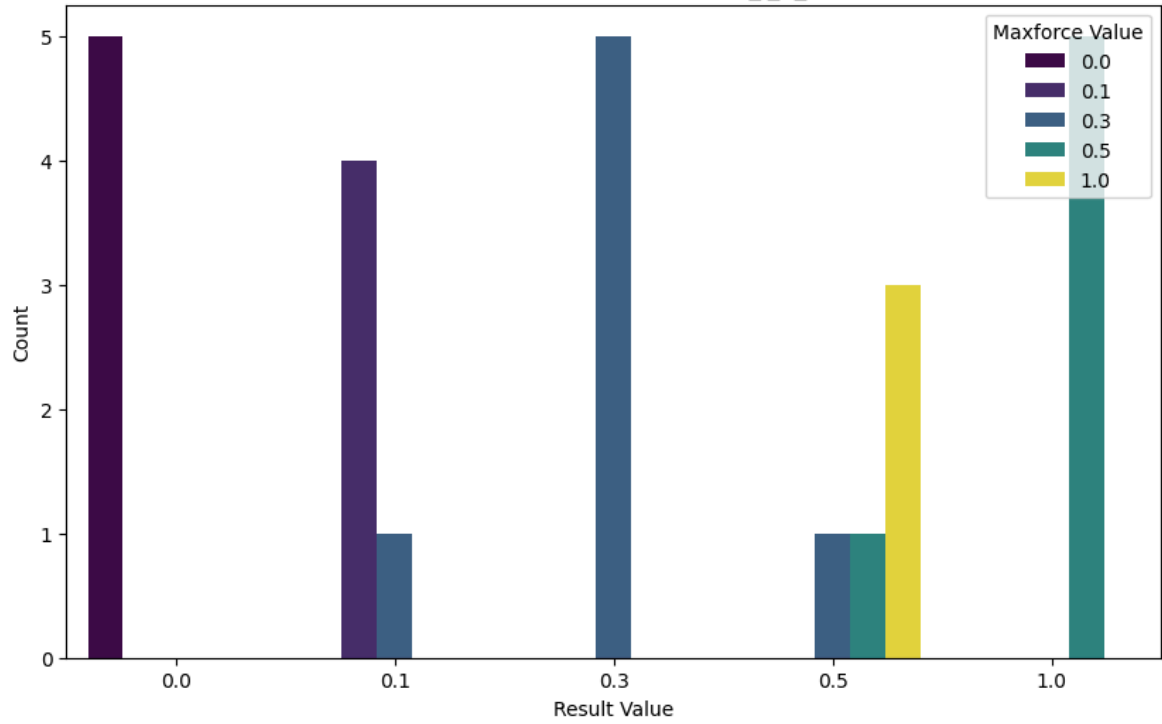
Maxforce vs Result Distribution - minho_S_3_combined.csv



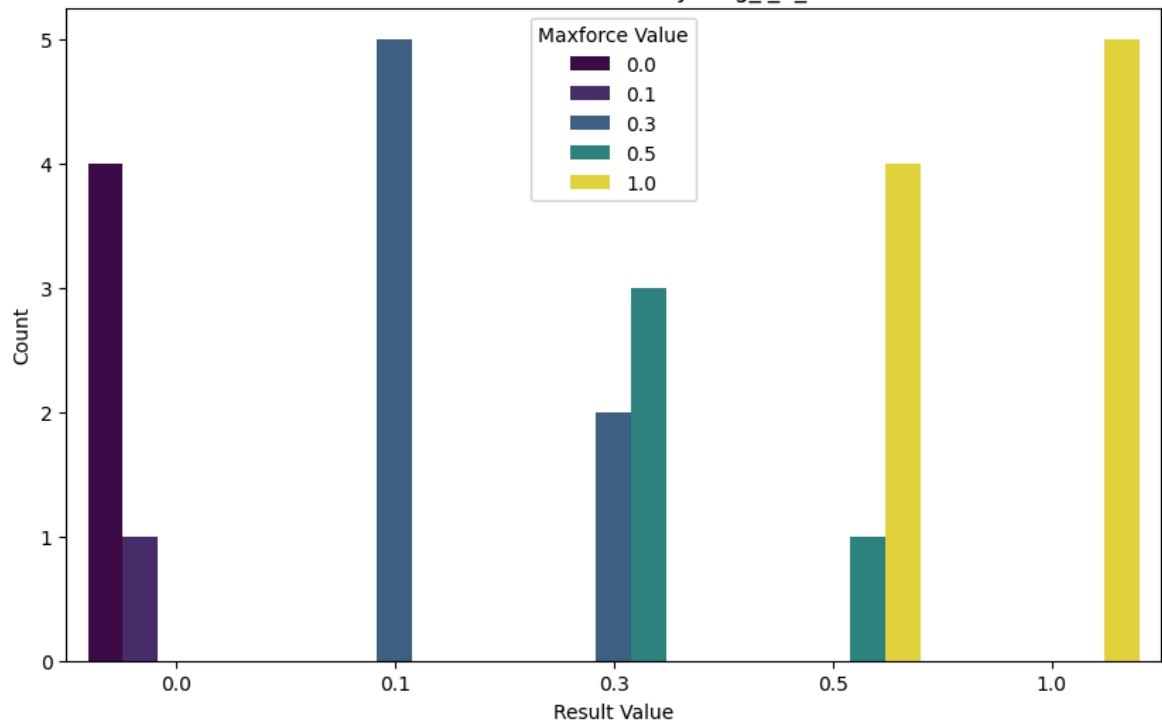
Maxforce vs Result Distribution - minho_S_4_combined.csv

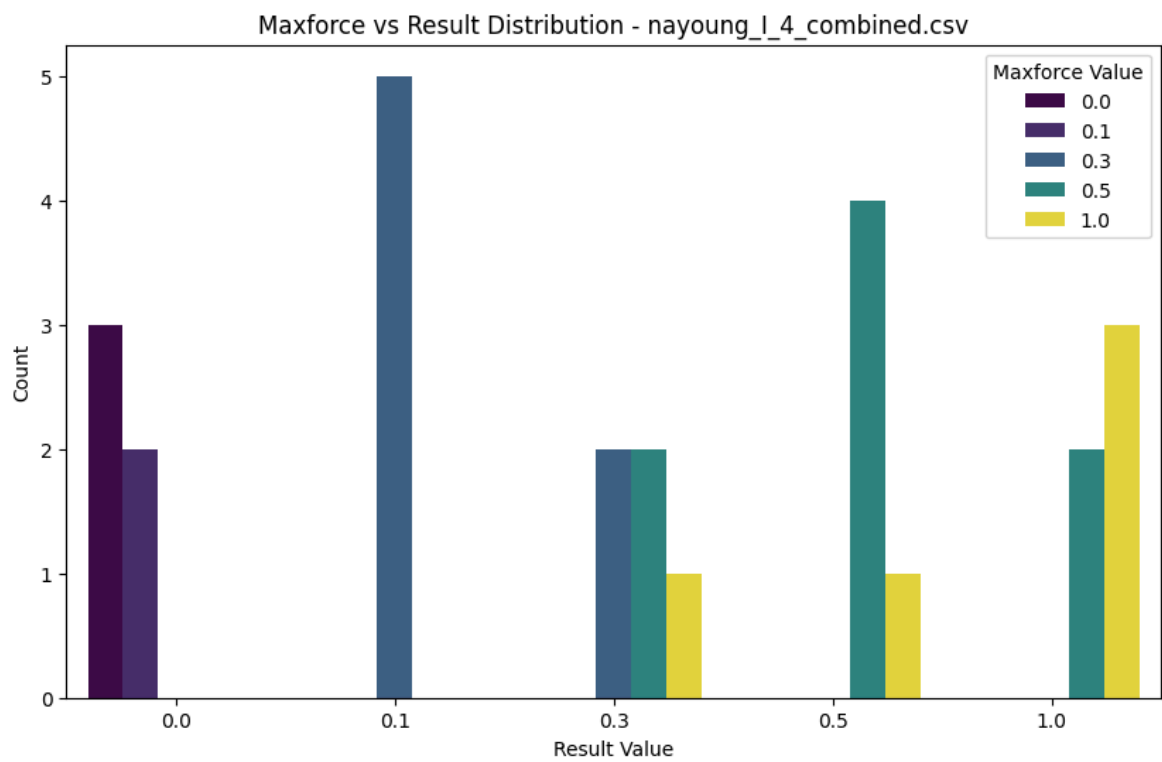
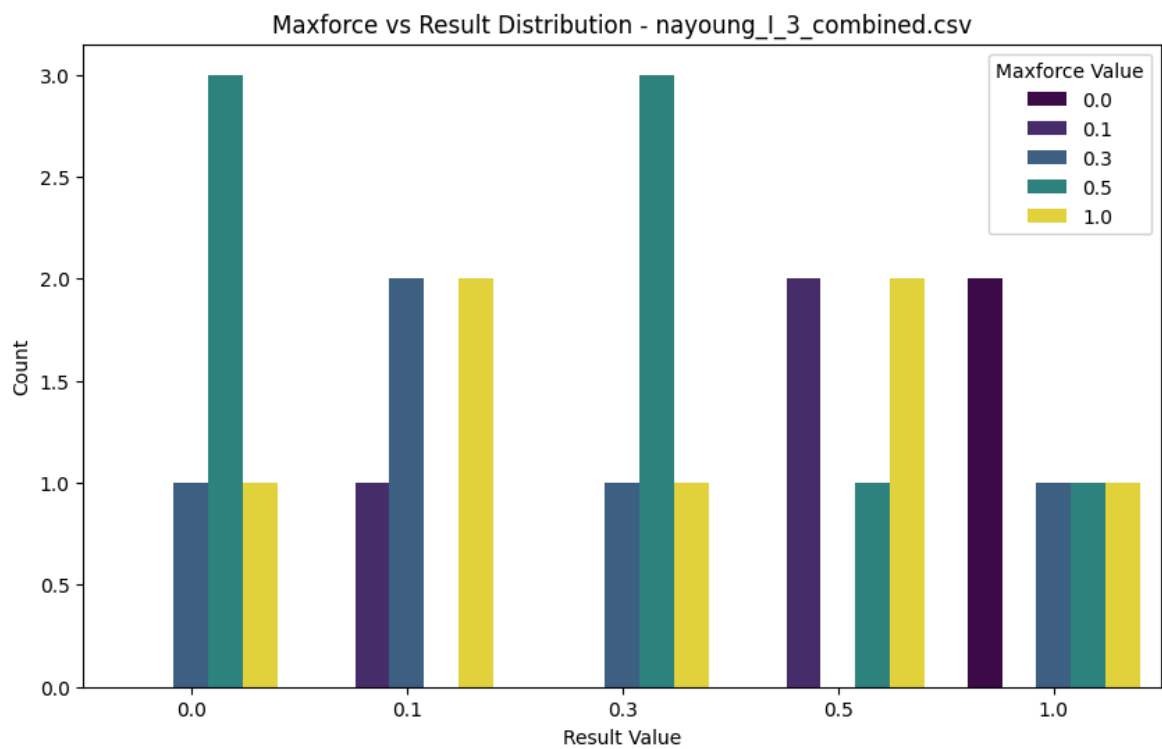


Maxforce vs Result Distribution - nayoung_I_1_combined.csv

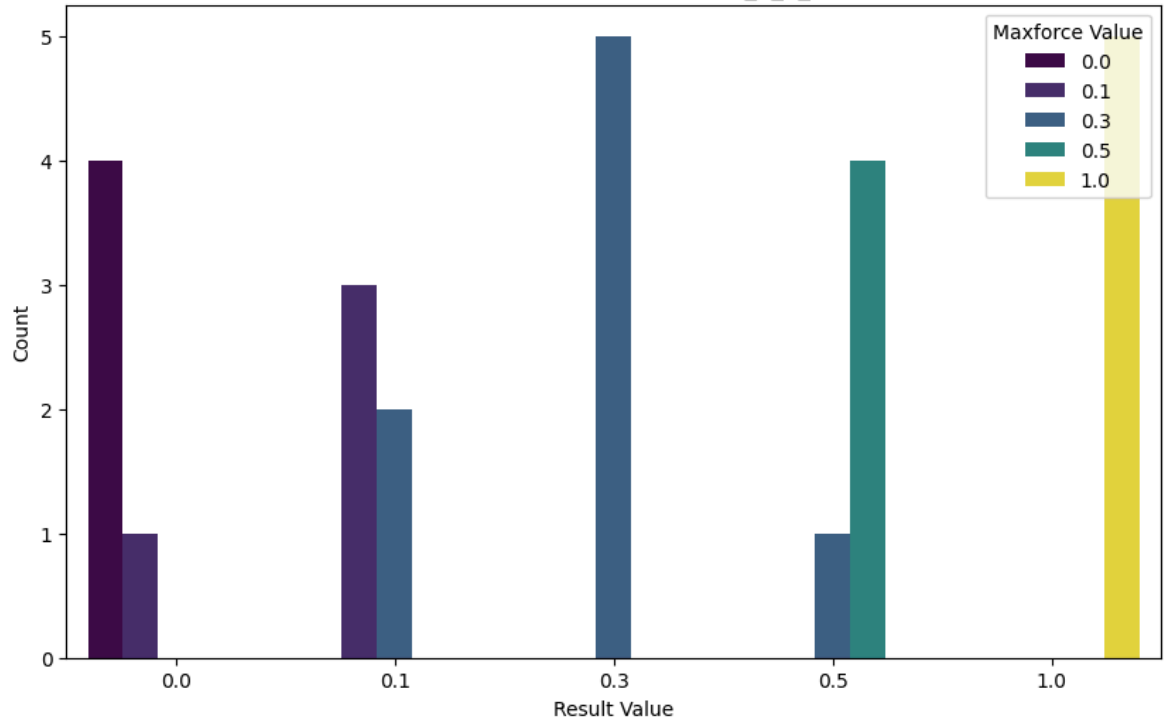


Maxforce vs Result Distribution - nayoung_I_2_combined.csv

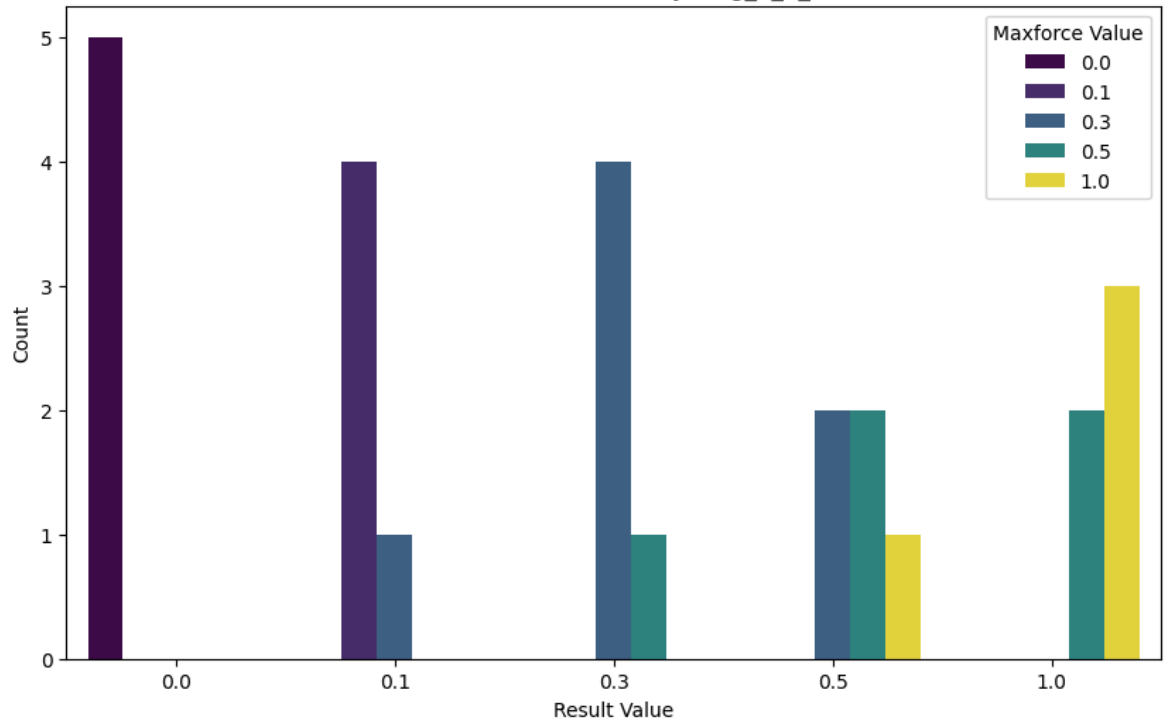




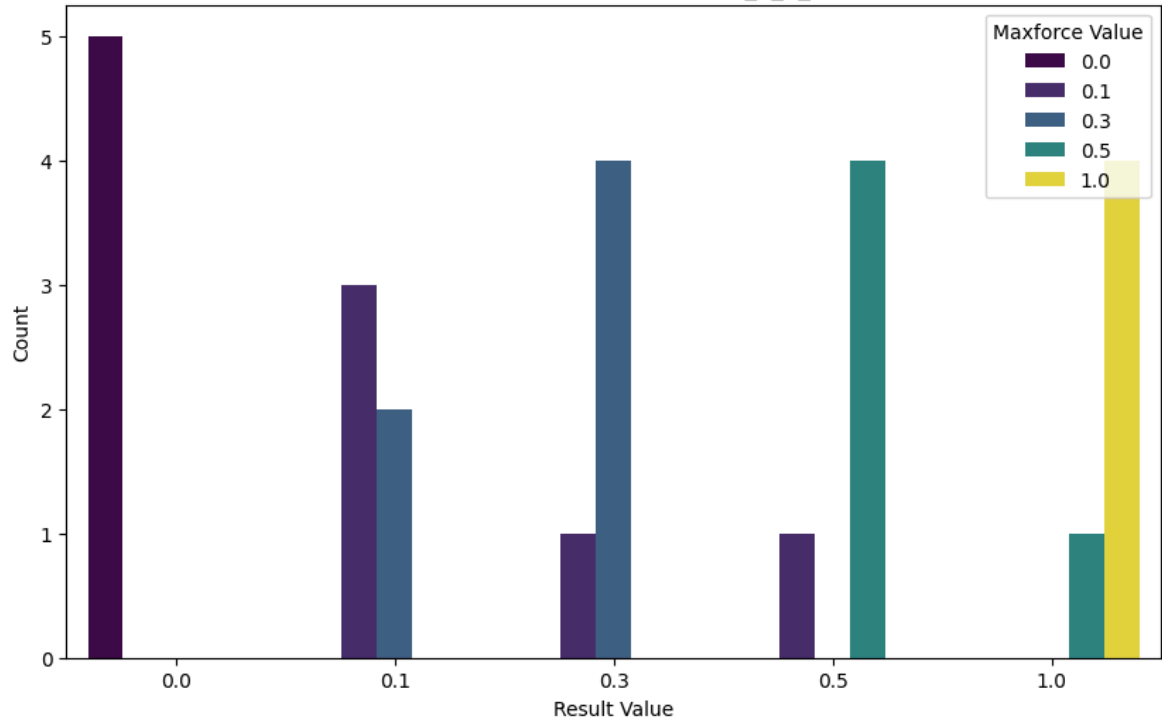
Maxforce vs Result Distribution - nayoung_S_1_combined.csv



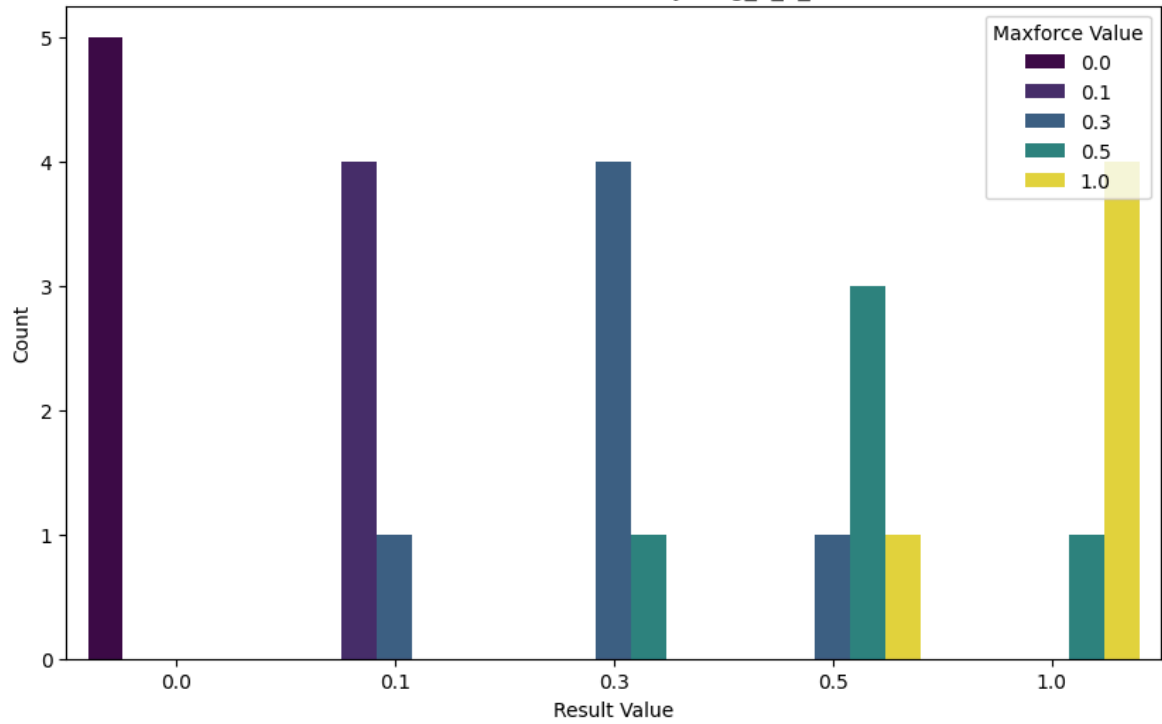
Maxforce vs Result Distribution - nayoung_S_2_combined.csv

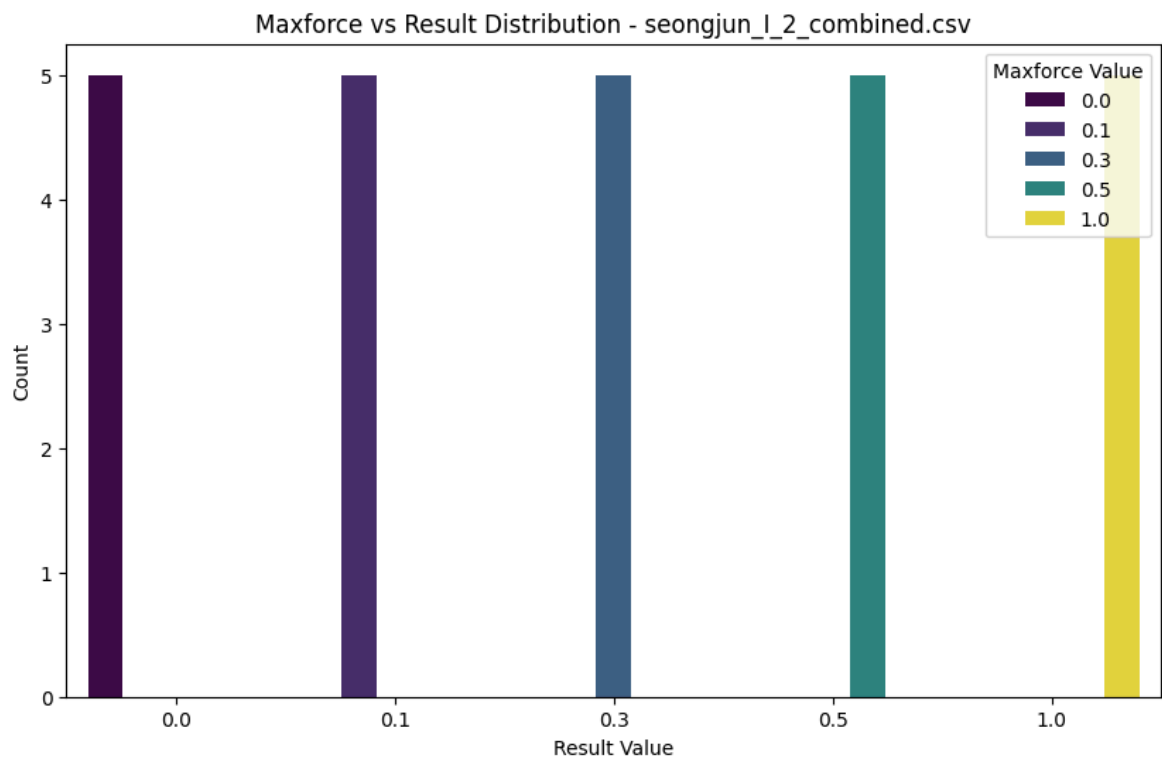
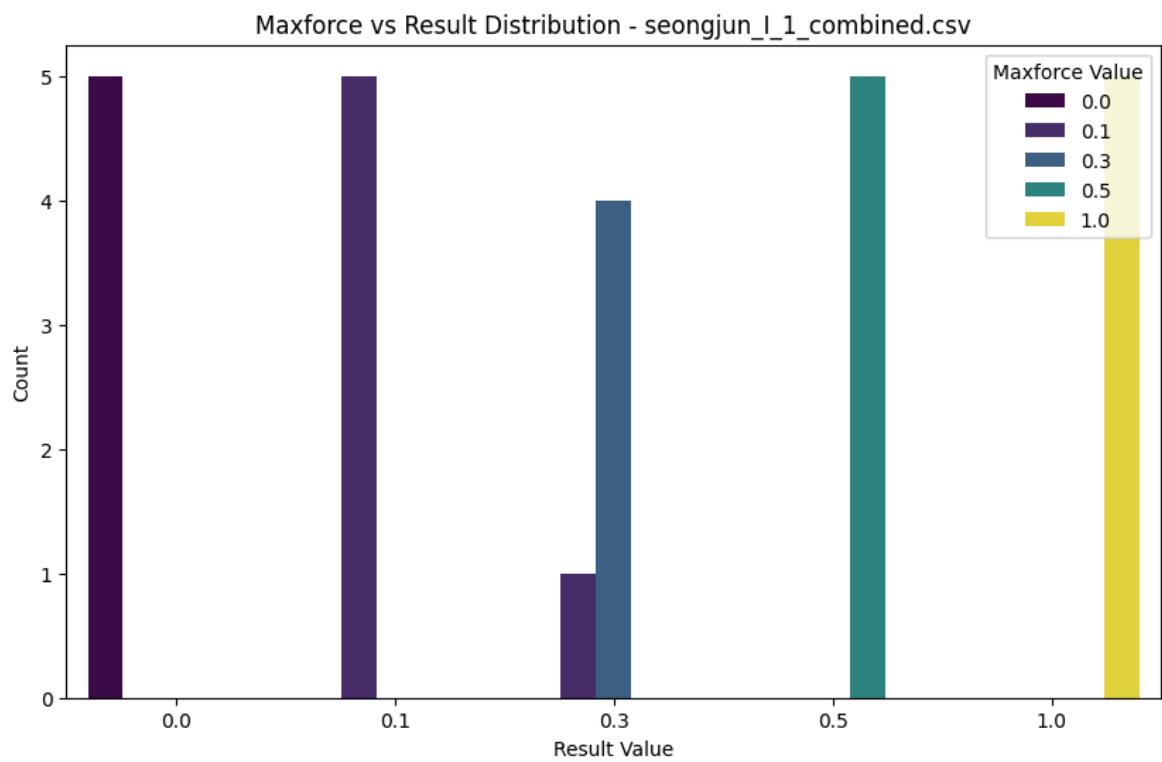


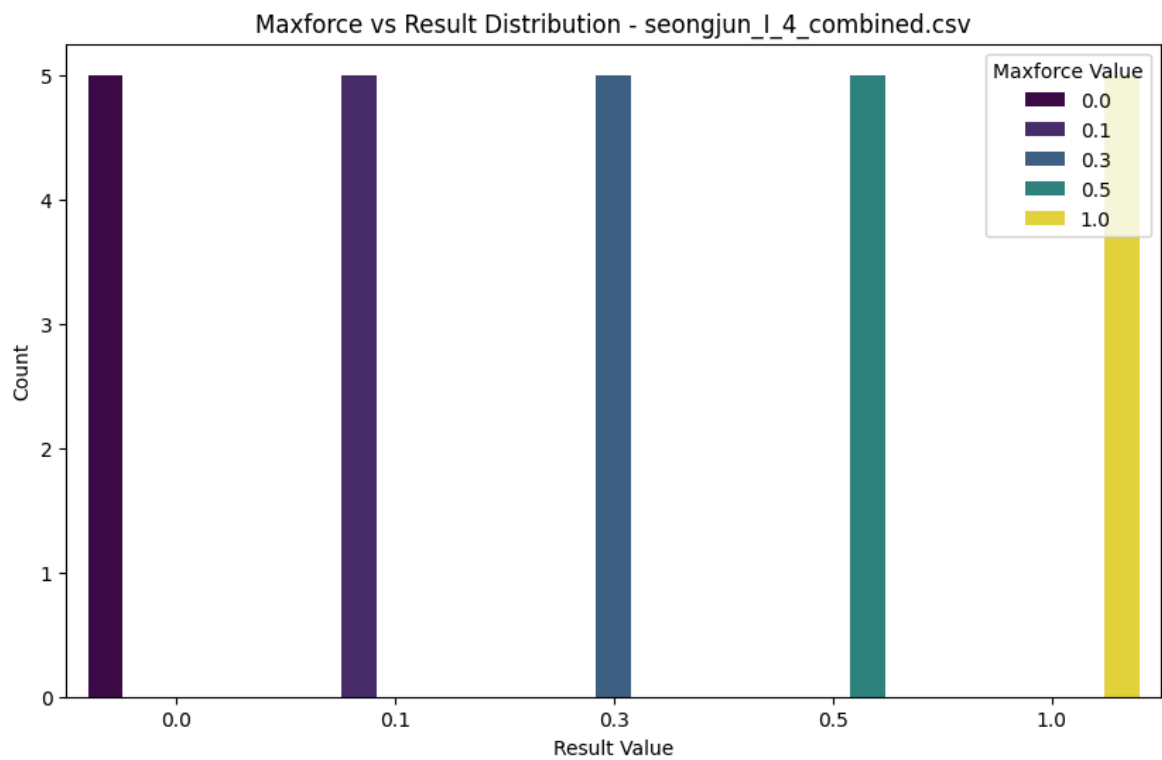
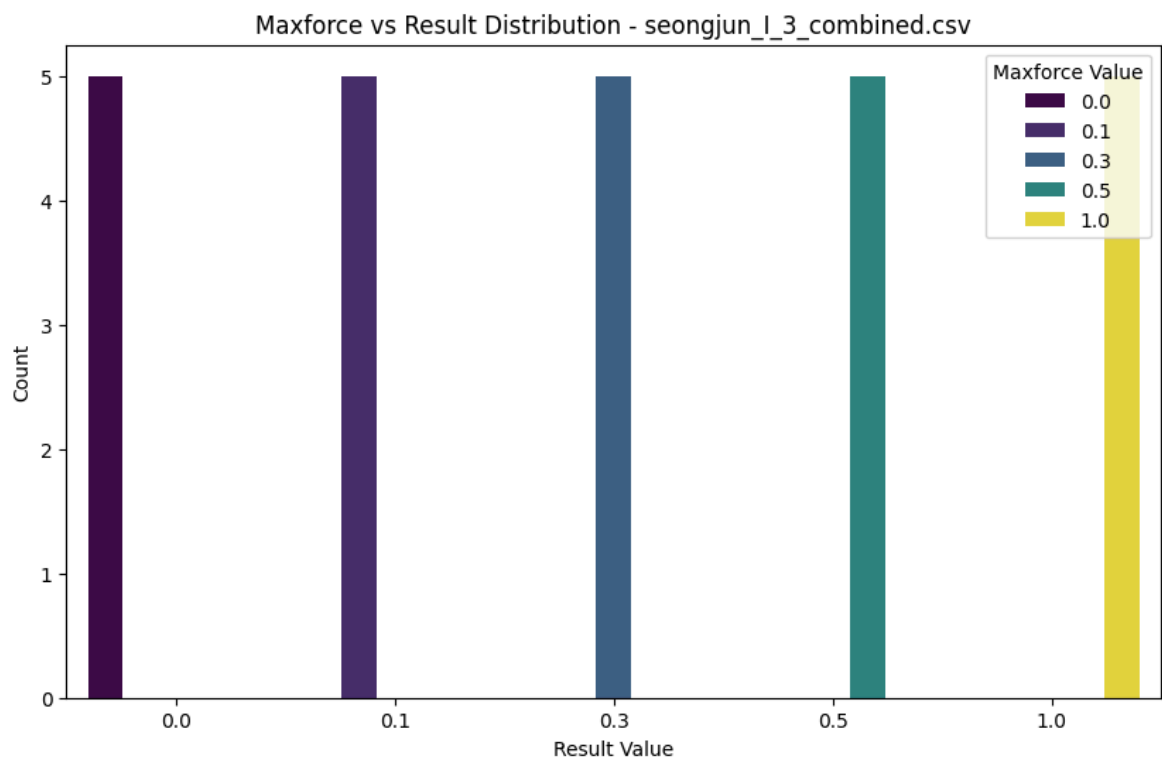
Maxforce vs Result Distribution - nayoung_S_3_combined.csv



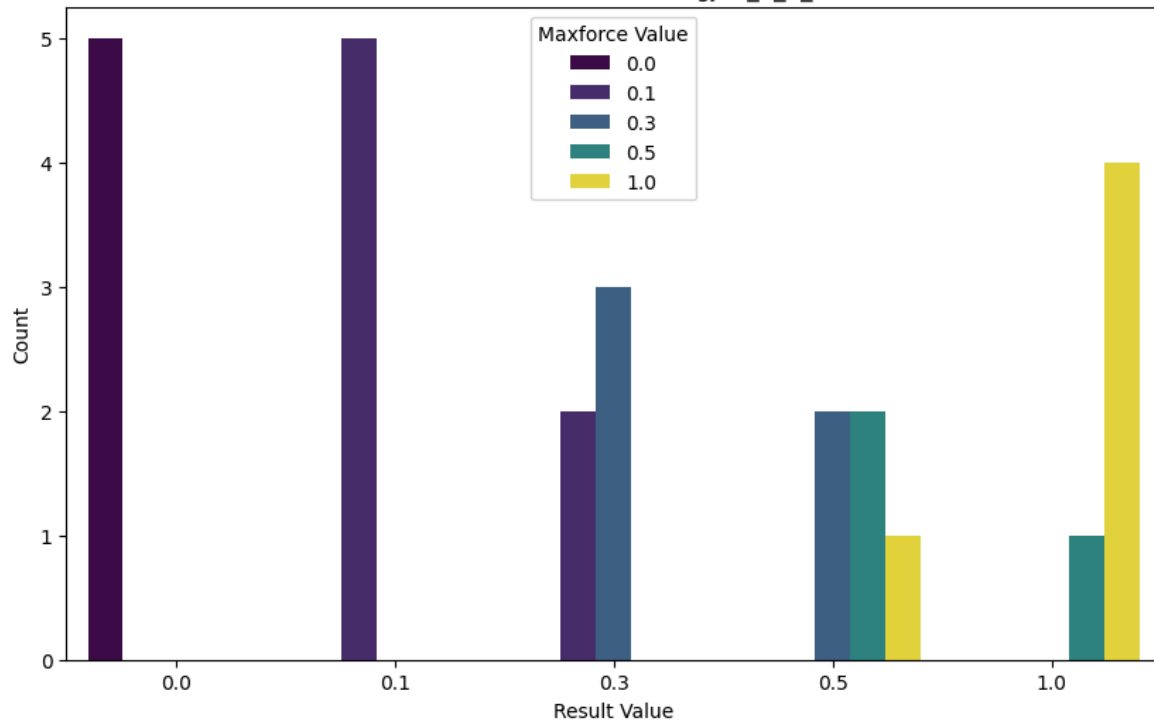
Maxforce vs Result Distribution - nayoung_S_4_combined.csv



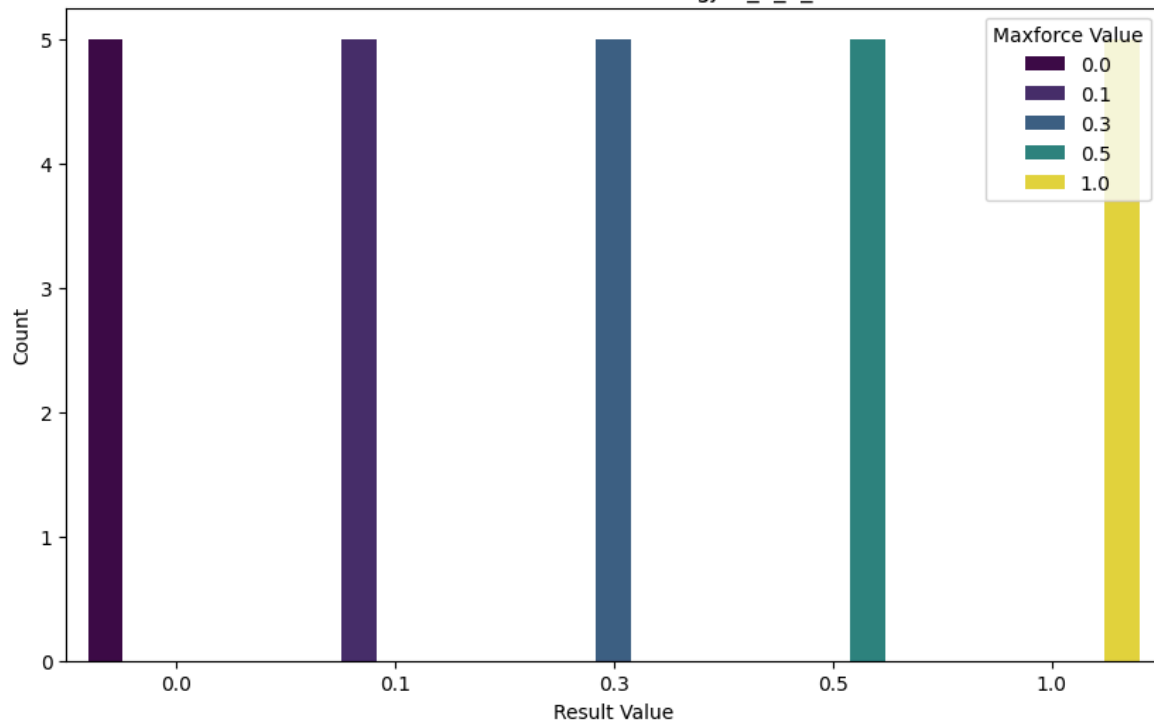




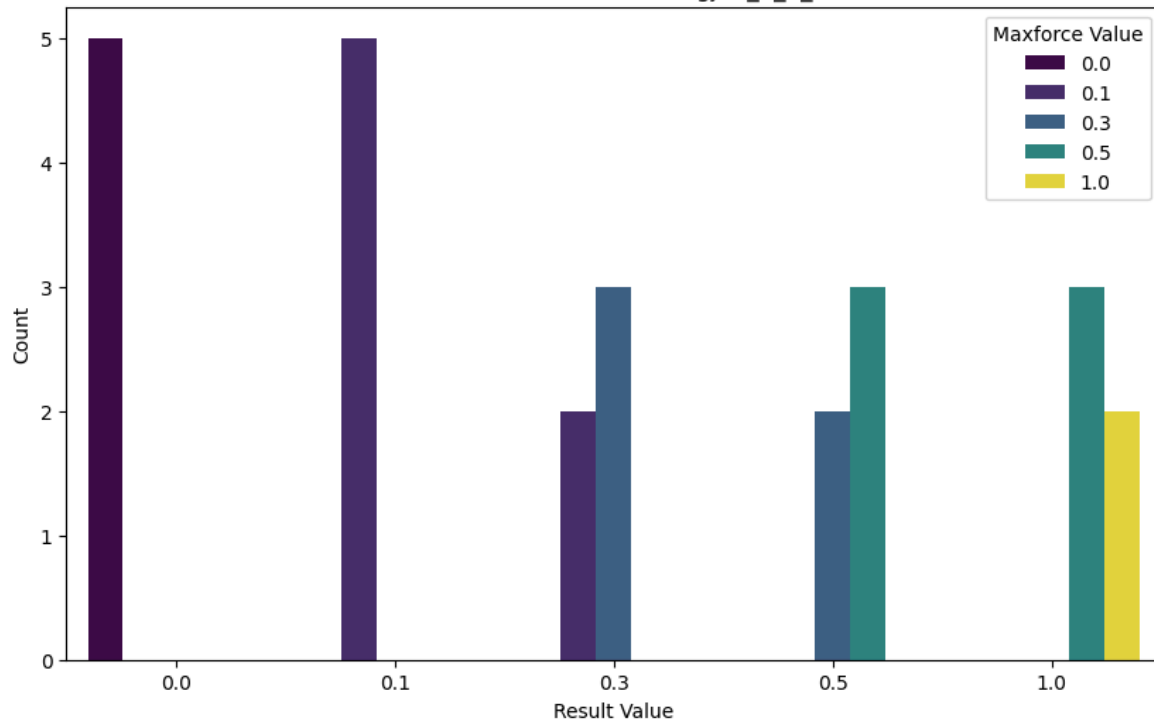
Maxforce vs Result Distribution - seongjun_S_1_combined.csv



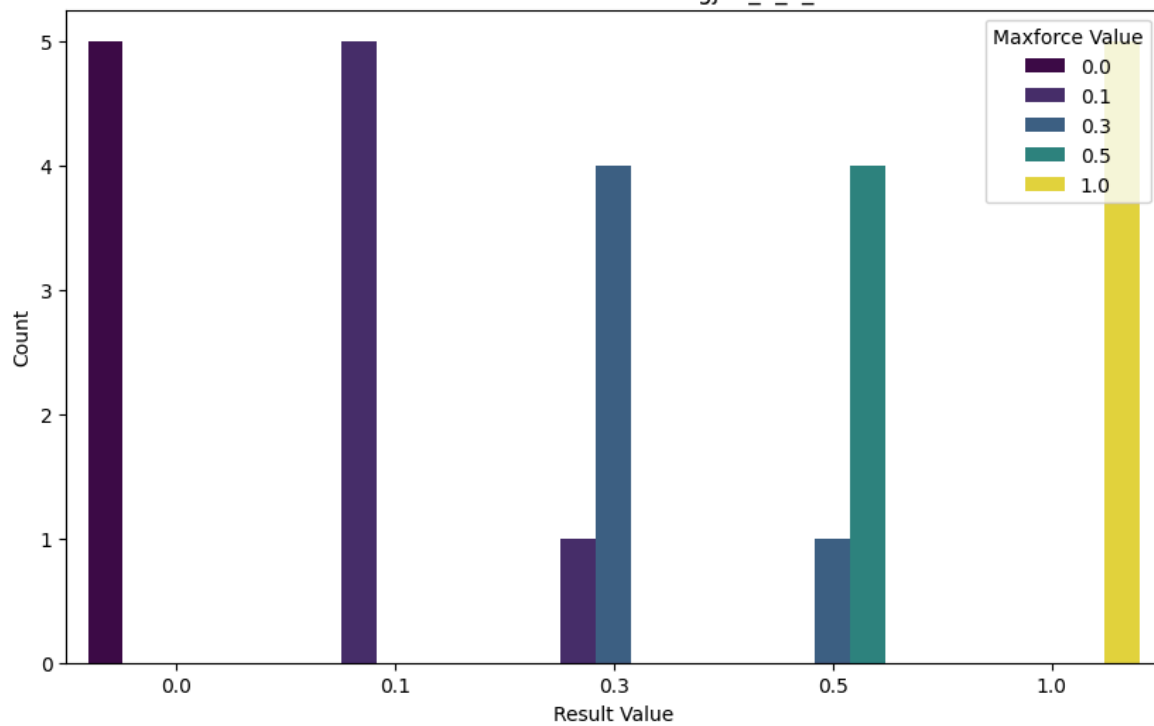
Maxforce vs Result Distribution - seongjun_S_2_combined.csv



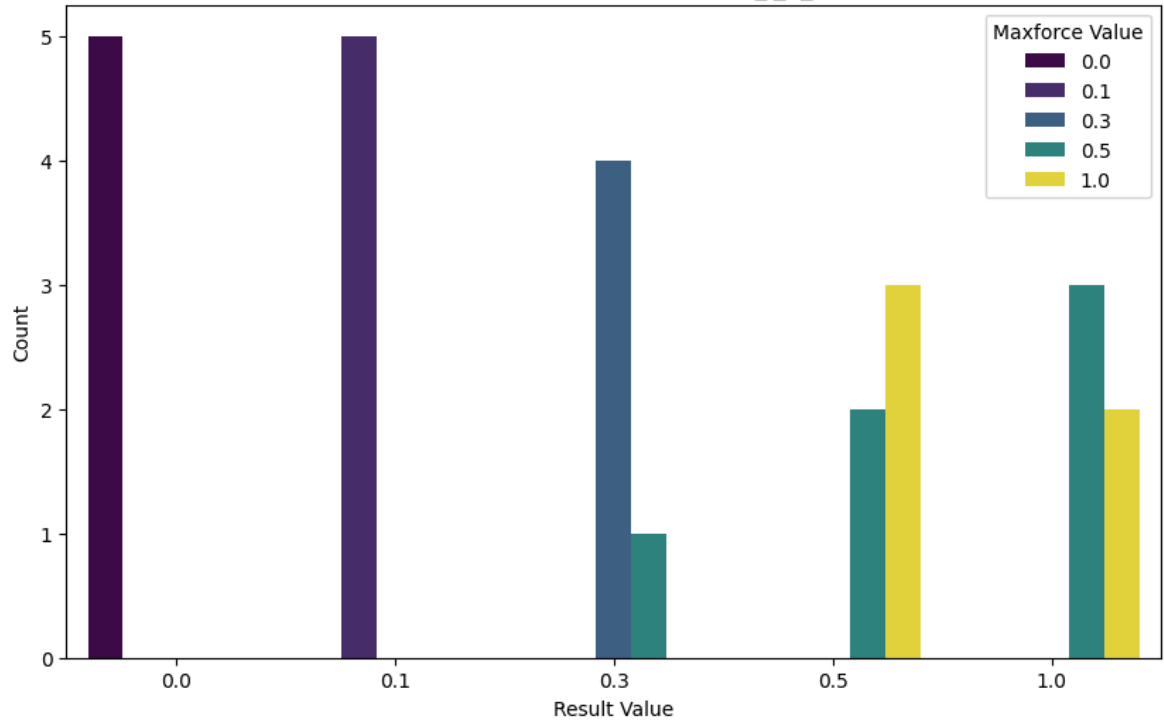
Maxforce vs Result Distribution - seongjun_S_3_combined.csv



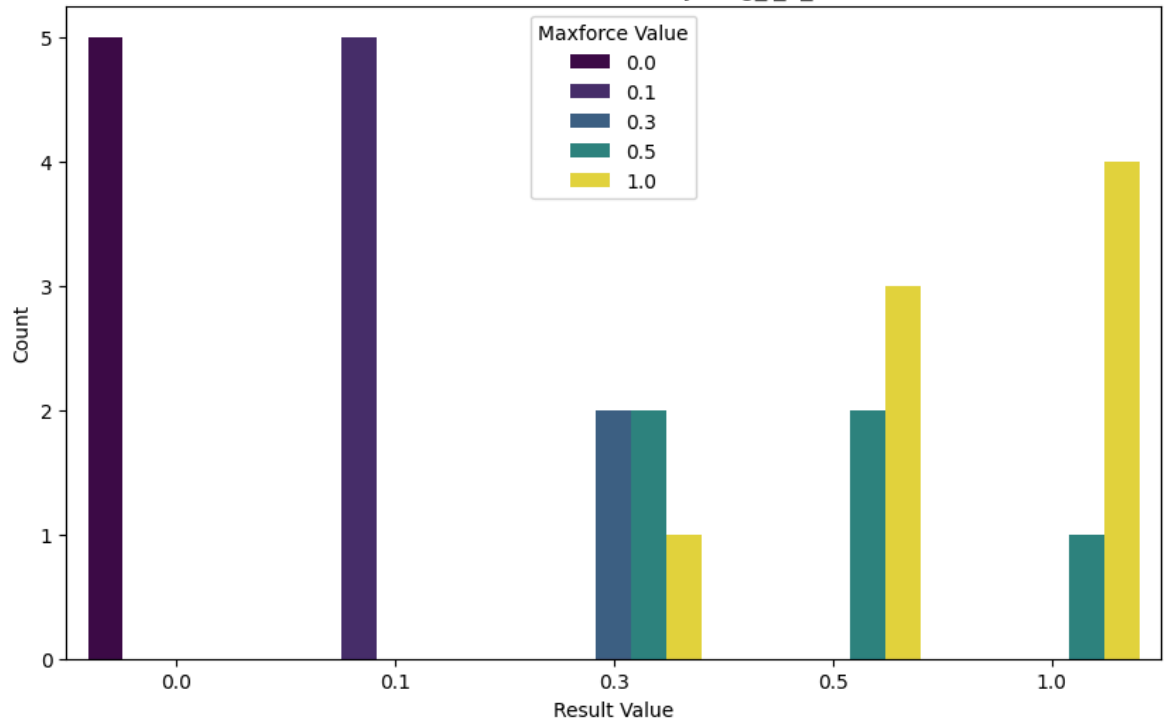
Maxforce vs Result Distribution - seongjun_S_4_combined.csv



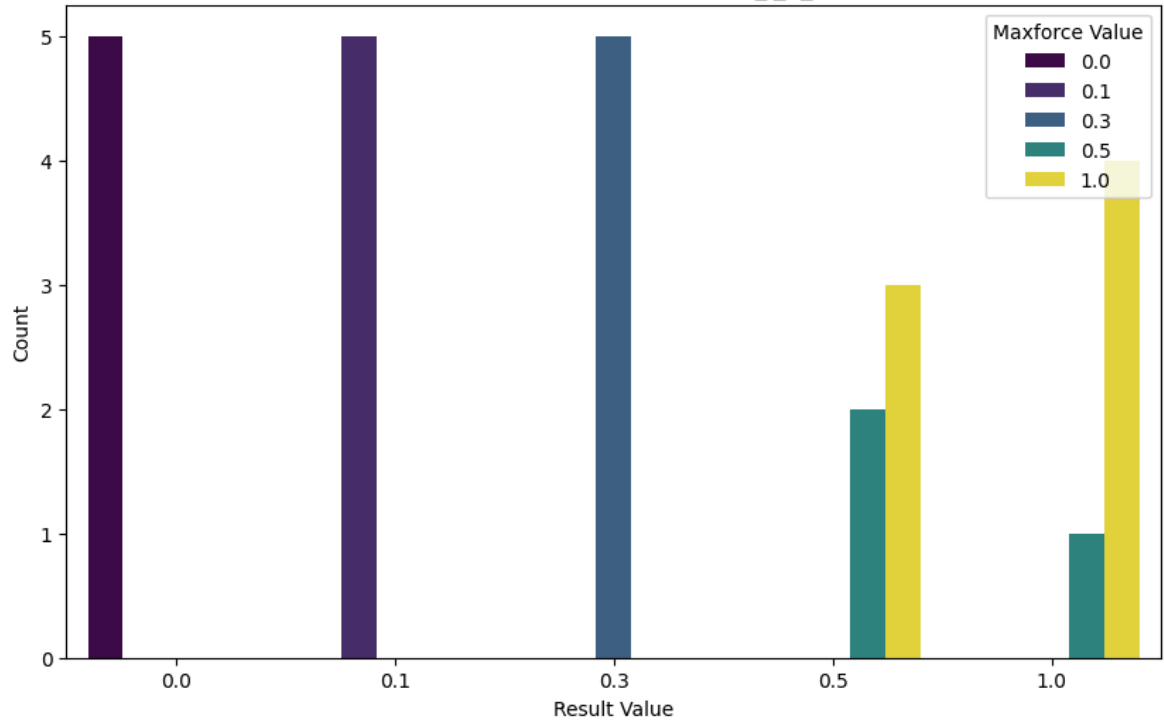
Maxforce vs Result Distribution - seoyeong_l_1_combined.csv



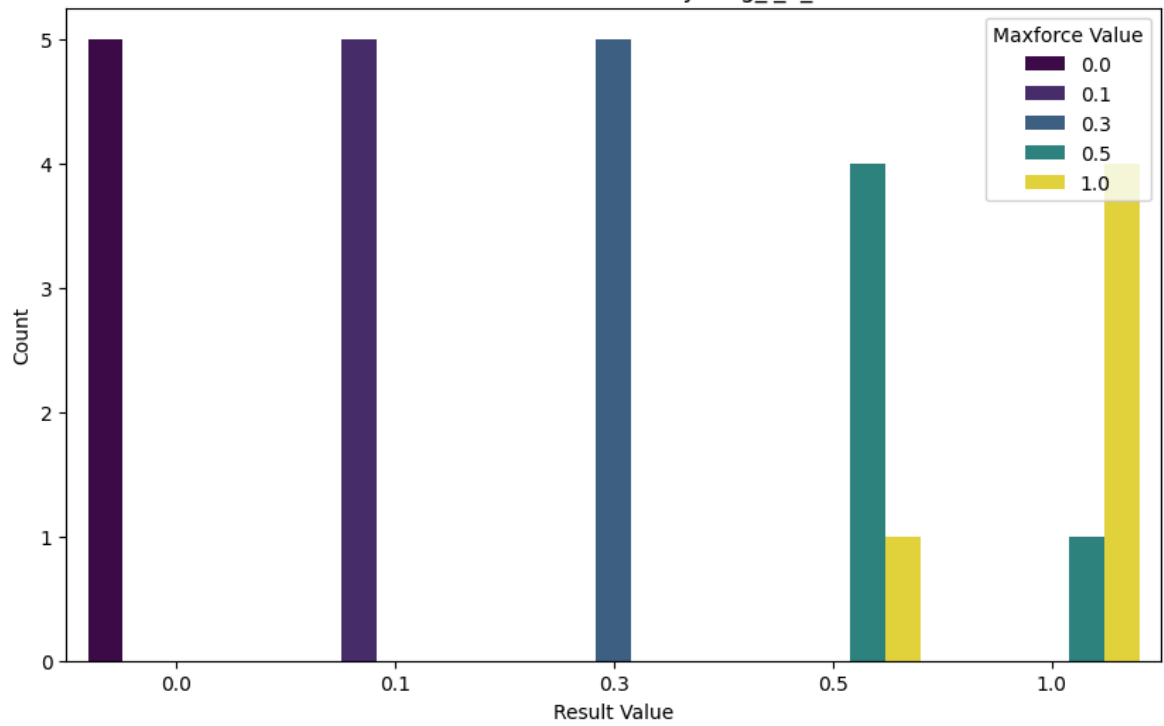
Maxforce vs Result Distribution - seoyeong_l_2_combined.csv



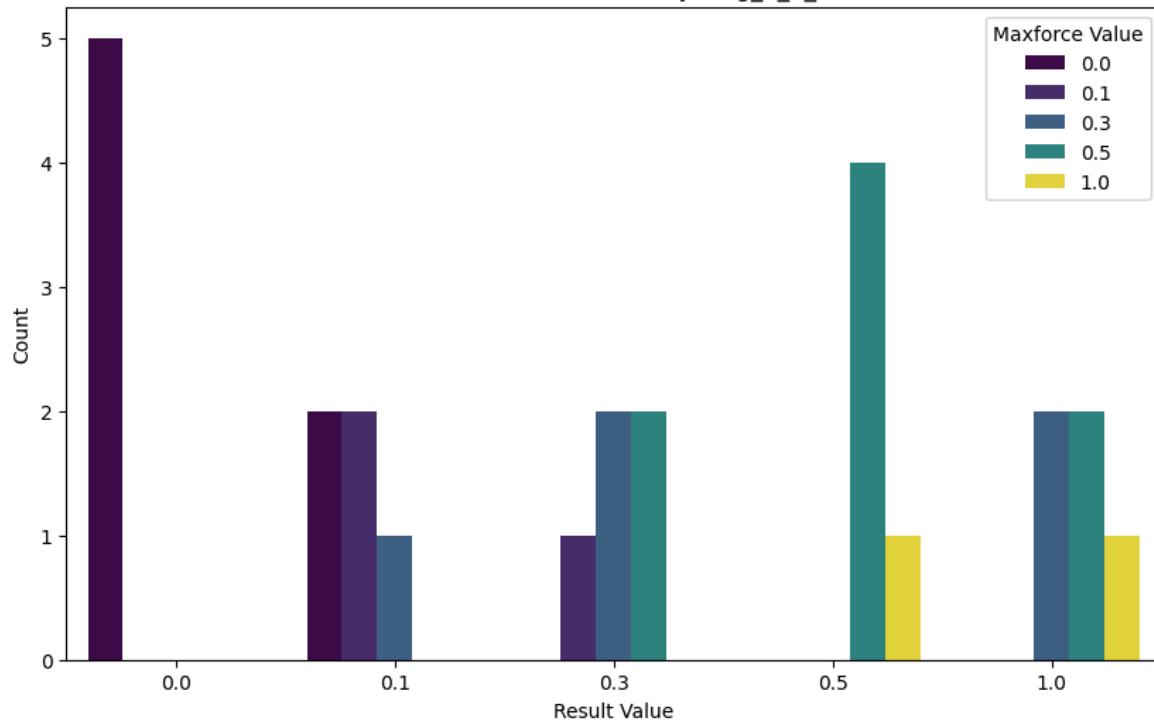
Maxforce vs Result Distribution - seoyeong_l_3_combined.csv



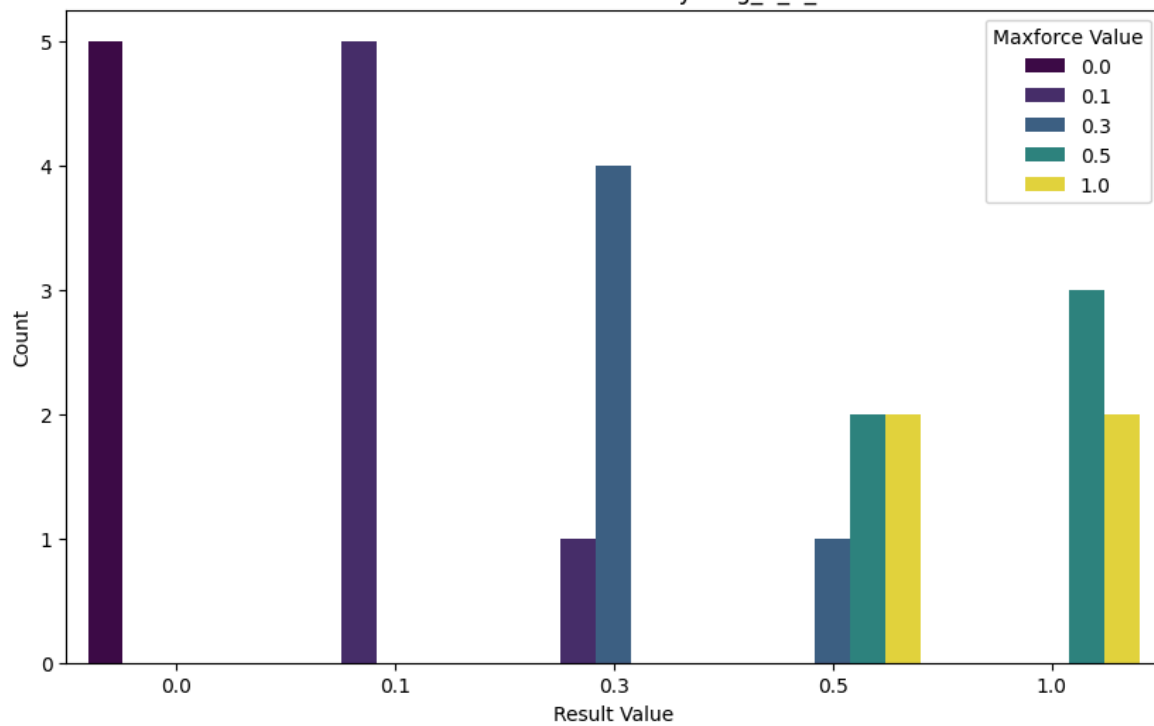
Maxforce vs Result Distribution - seoyeong_l_4_combined.csv



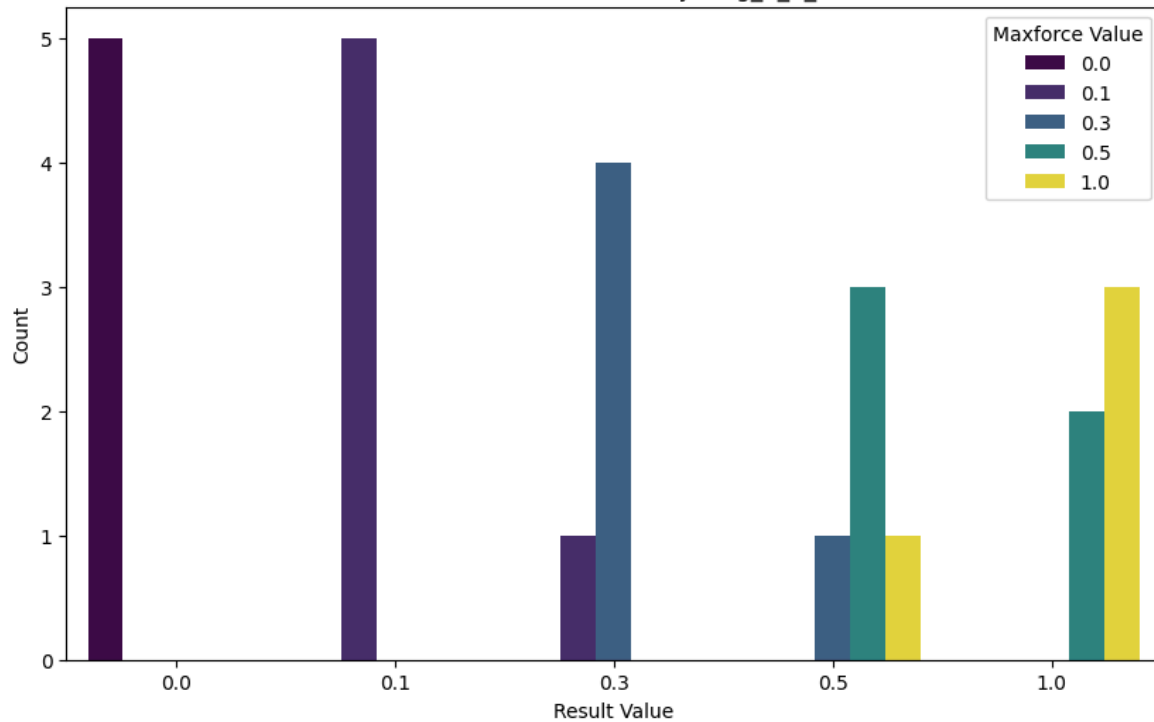
Maxforce vs Result Distribution - seoyeong_S_1_combined.csv



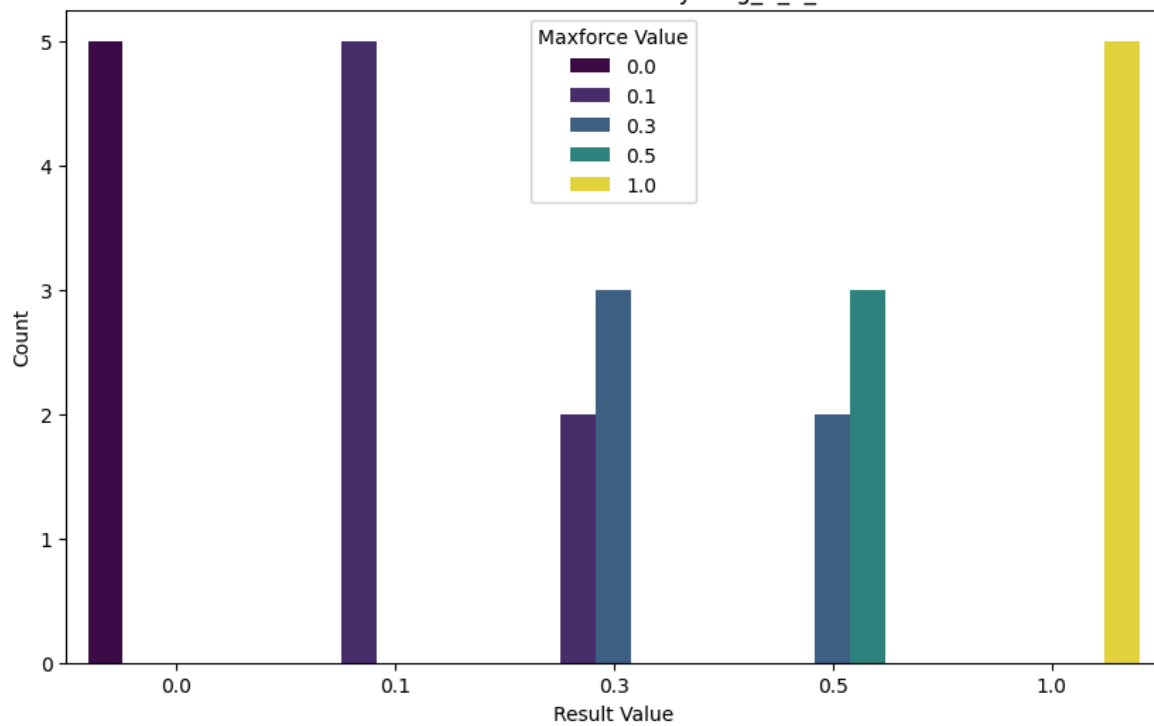
Maxforce vs Result Distribution - seoyeong_S_2_combined.csv



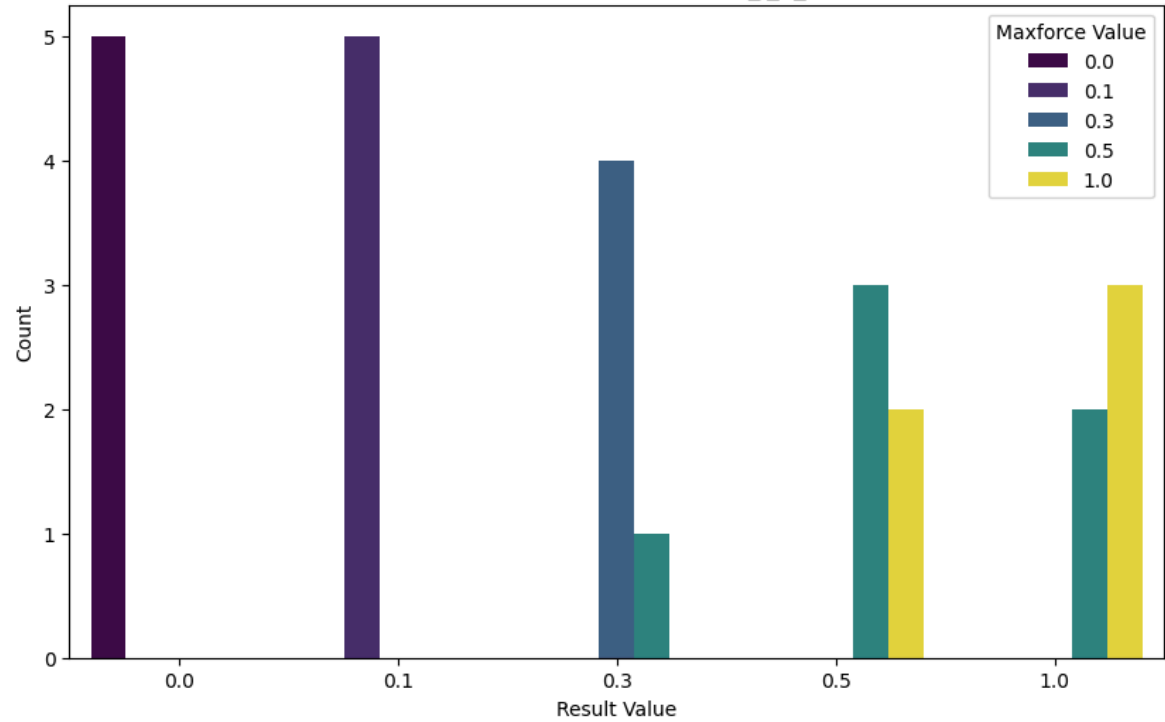
Maxforce vs Result Distribution - seoyeong_S_3_combined.csv



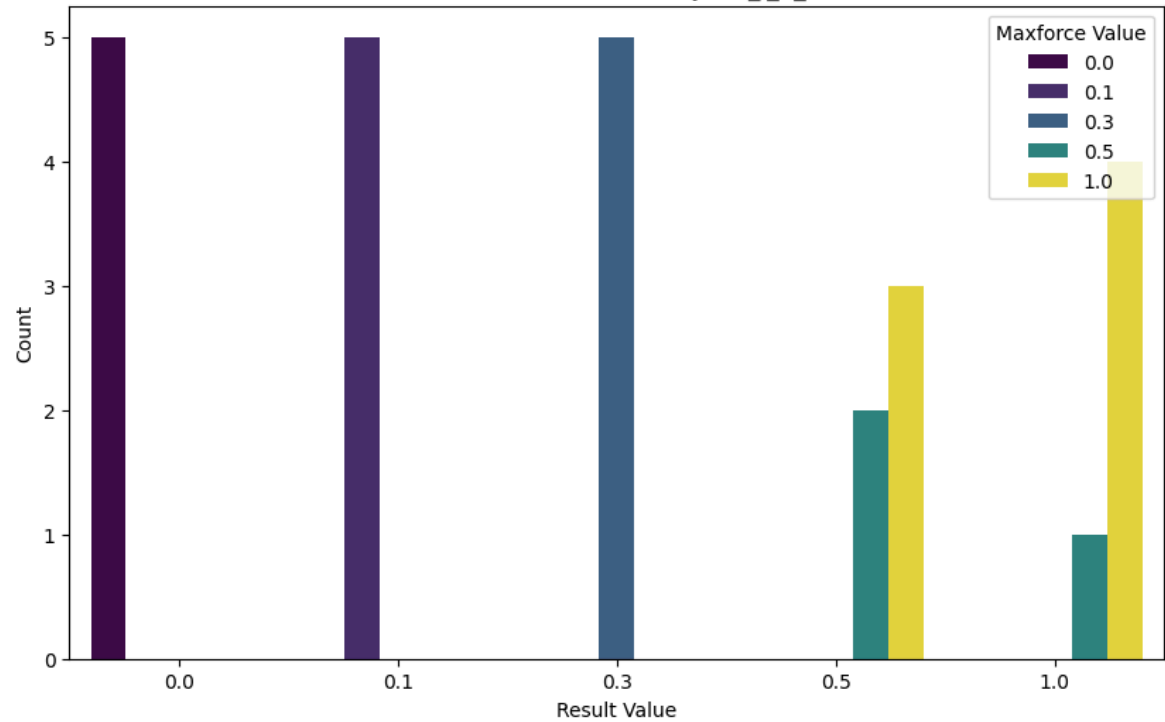
Maxforce vs Result Distribution - seoyeong_S_4_combined.csv

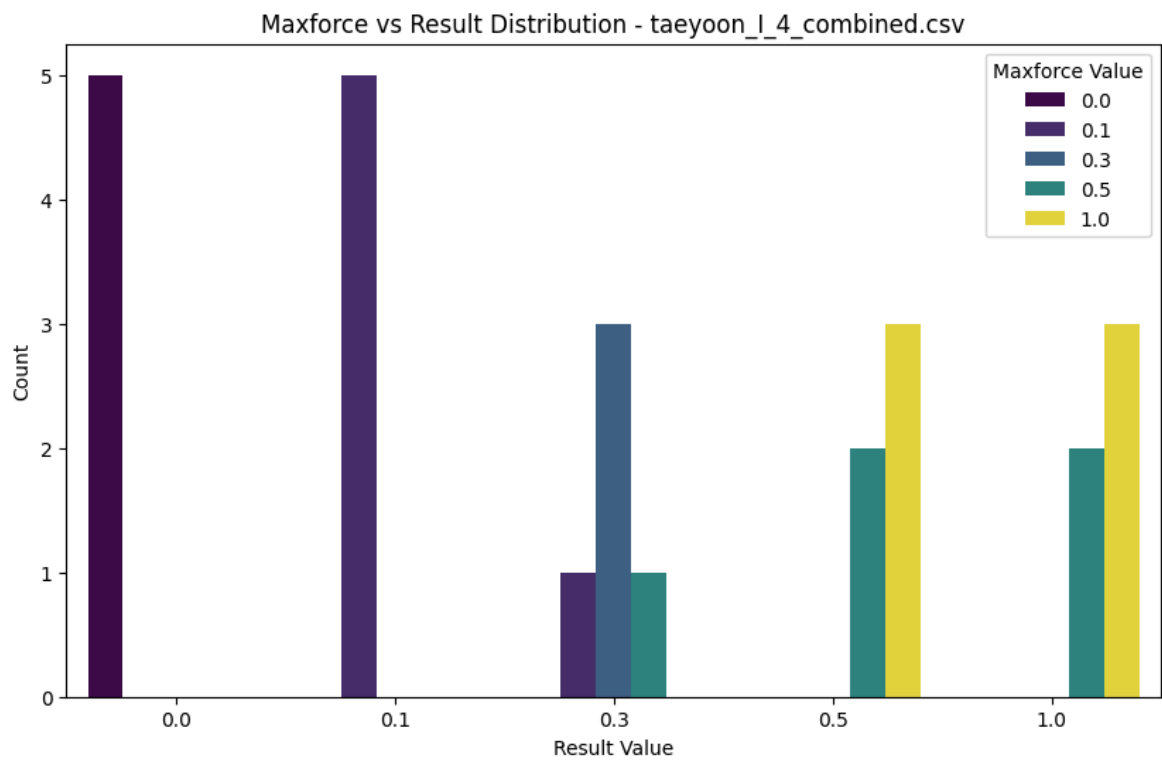
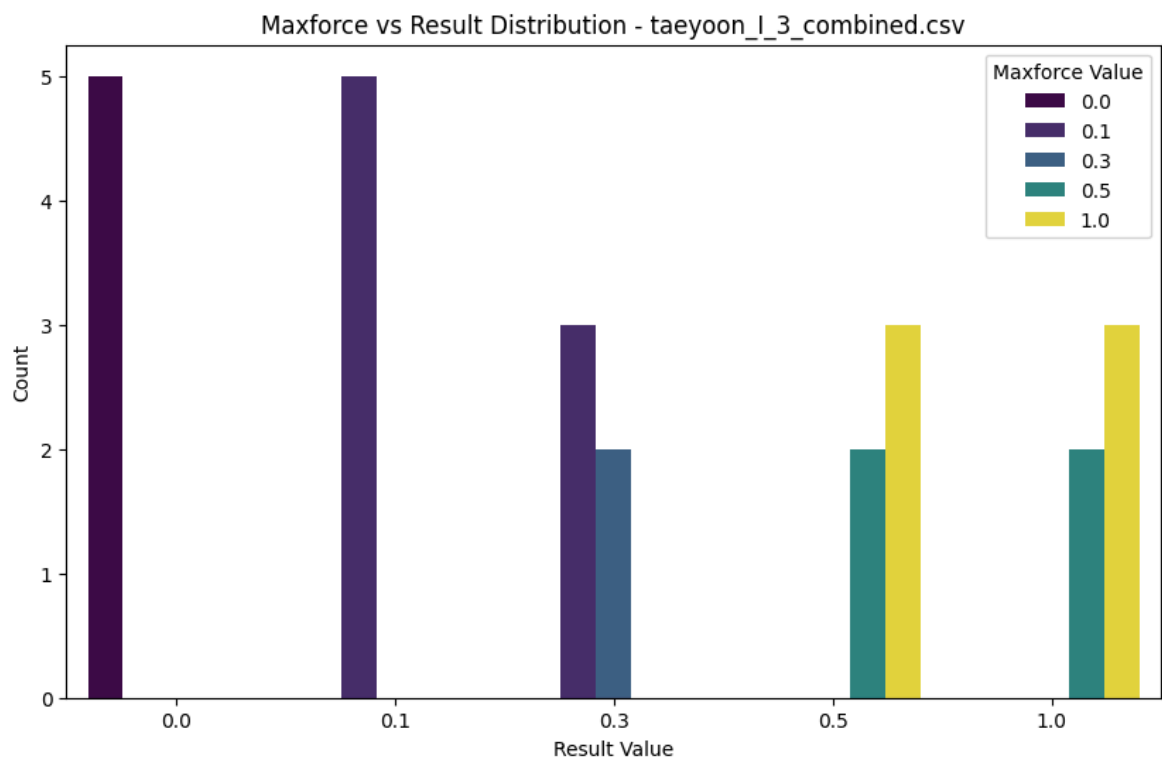


Maxforce vs Result Distribution - taeyoon_I_1_combined.csv

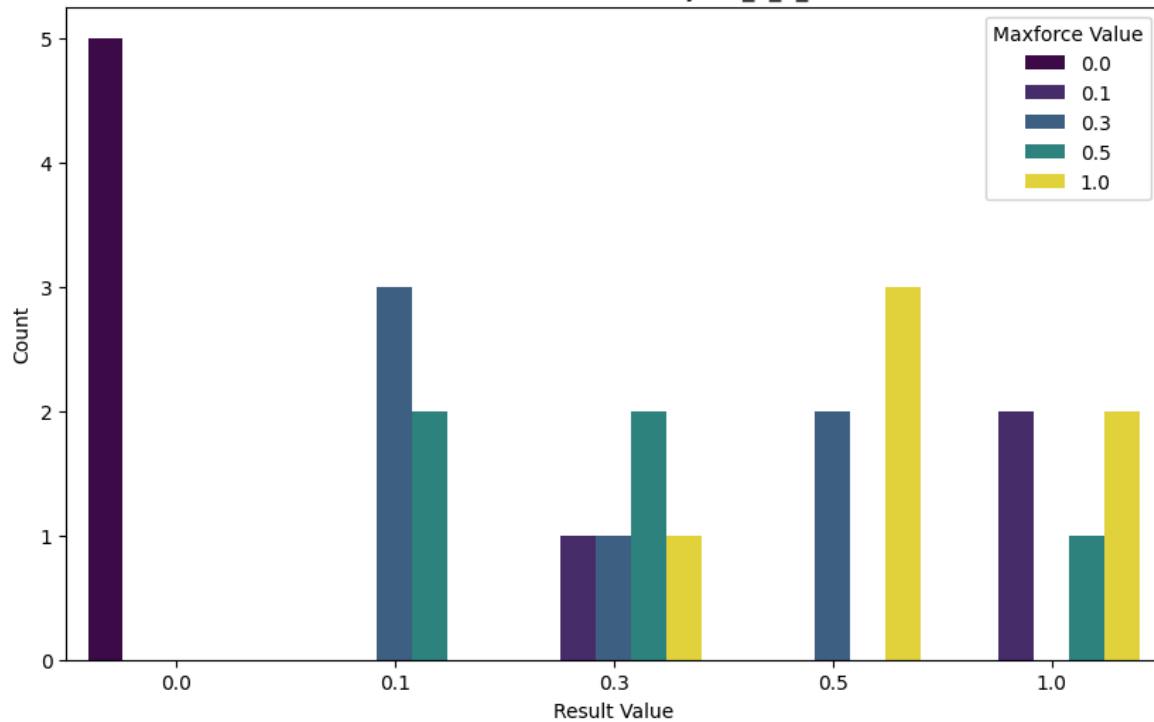


Maxforce vs Result Distribution - taeyoon_I_2_combined.csv

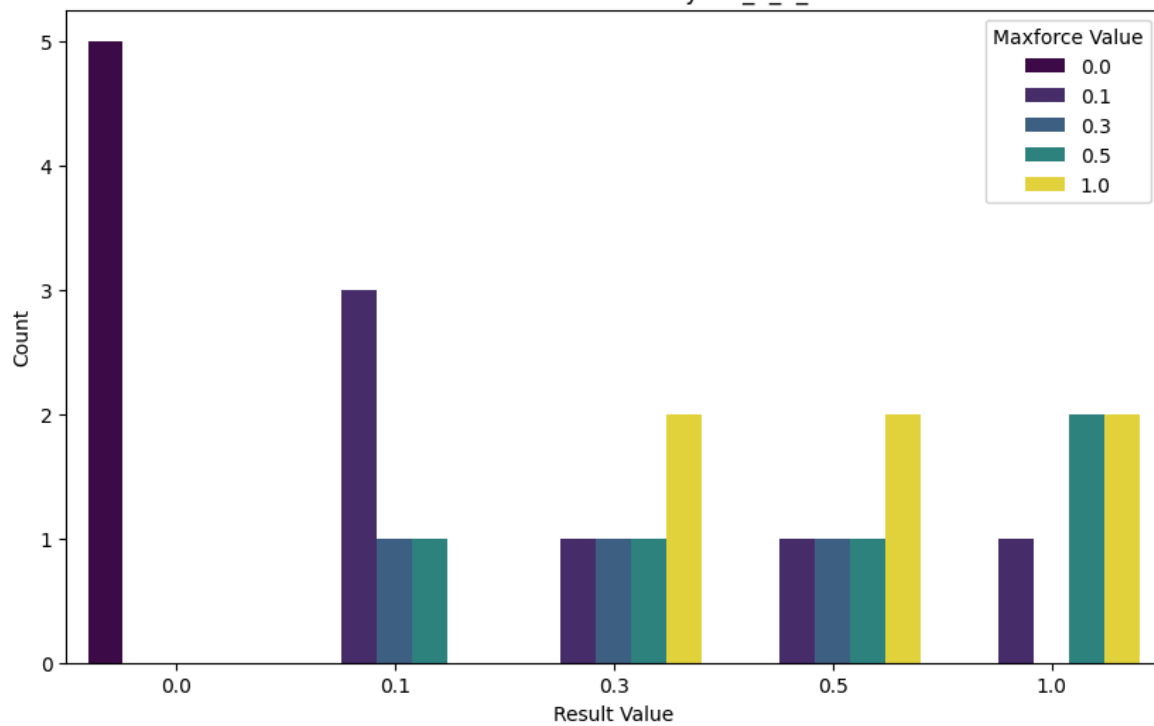


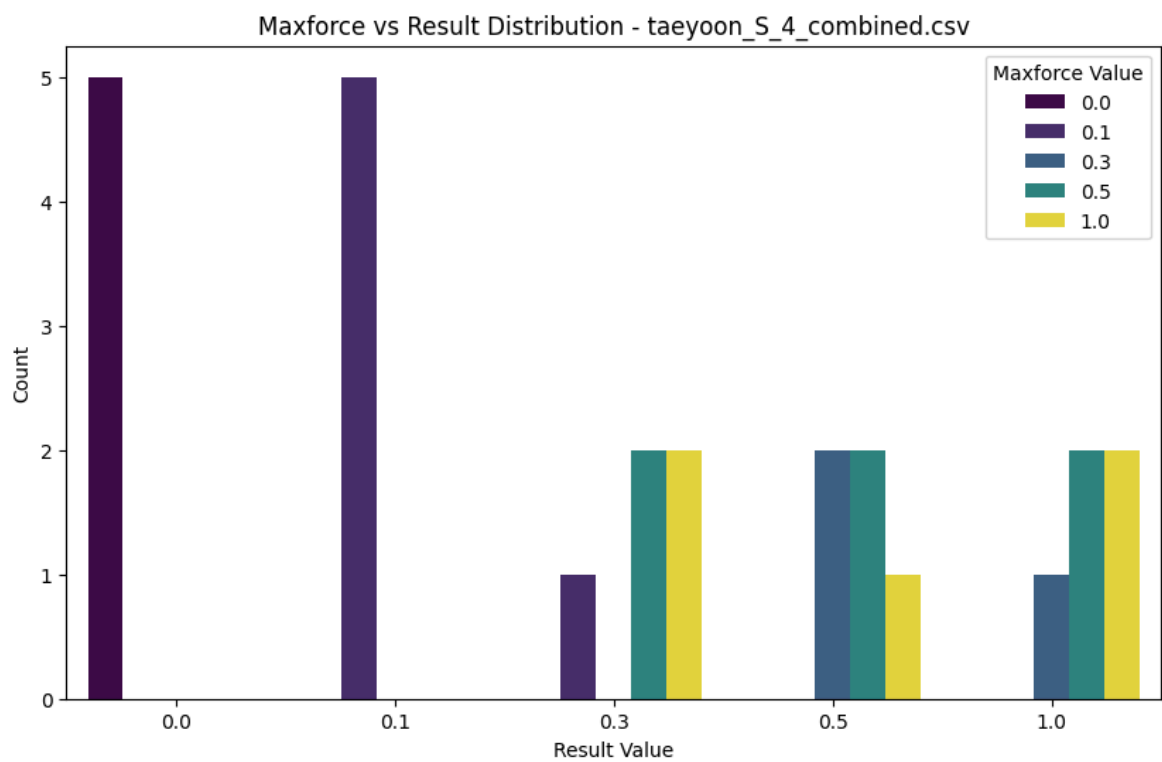
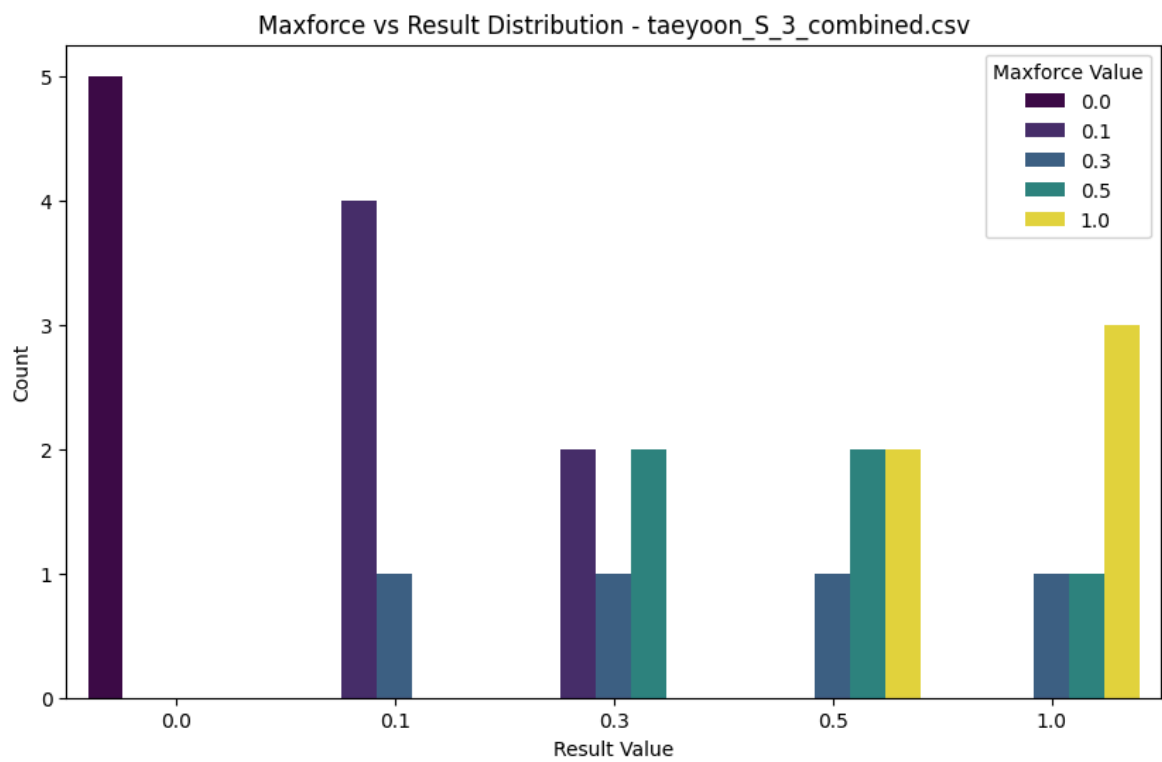


Maxforce vs Result Distribution - taeyoon_S_1_combined.csv



Maxforce vs Result Distribution - taeyoon_S_2_combined.csv





```
In [36]: pd.describe_option()
```

```

compute.use_bottleneck : bool
    Use the bottleneck library to accelerate if it is installed,
    the default is True
    Valid values: False,True
    [default: True] [currently: True]
compute.use_numba : bool
    Use the numba engine option for select operations if it is installed,
    the default is False
    Valid values: False,True
    [default: False] [currently: False]
compute.use_numexpr : bool
    Use the numexpr library to accelerate computation if it is installed,
    the default is True
    Valid values: False,True
    [default: True] [currently: True]
display.chop_threshold : float or None
    if set to a float value, all float values smaller than the given threshold
    will be displayed as exactly 0 by repr and friends.
    [default: None] [currently: None]
display.colheader_justify : 'left'/'right'
    Controls the justification of column headers. used by DataFrameFormatter.
    [default: right] [currently: right]
display.date_dayfirst : boolean
    When True, prints and parses dates with the day first, eg 20/01/2005
    [default: False] [currently: False]
display.date_yearfirst : boolean
    When True, prints and parses dates with the year first, eg 2005/01/20
    [default: False] [currently: False]
display.encoding : str/unicode
    Defaults to the detected encoding of the console.
    Specifies the encoding to be used for strings returned by to_string,
    these are generally strings meant to be displayed on the console.
    [default: UTF-8] [currently: UTF-8]
display.expand_frame_repr : boolean
    Whether to print out the full DataFrame repr for wide DataFrames across
    multiple lines, `max_columns` is still respected, but the output will
    wrap-around across multiple "pages" if its width exceeds `display.width`.
    [default: True] [currently: True]
display.float_format : callable
    The callable should accept a floating point number and return
    a string with the desired format of the number. This is used
    in some places like SeriesFormatter.
    See formats.format.EngFormatter for an example.
    [default: None] [currently: None]
display.html.border : int
    A ``border=value`` attribute is inserted in the ``<table>`` tag
    for the DataFrame HTML repr.
    [default: 1] [currently: 1]
display.html.table_schema : boolean
    Whether to publish a Table Schema representation for frontends
    that support it.
    (default: False)
    [default: False] [currently: False]
display.html.use_mathjax : boolean
    When True, Jupyter notebook will process table contents using MathJax,
    rendering mathematical expressions enclosed by the dollar symbol.
    (default: True)
    [default: True] [currently: True]
display.large_repr : 'truncate'/'info'
    For DataFrames exceeding max_rows/max_cols, the repr (and HTML repr) can

```

show a truncated table, or switch to the view from `df.info()` (the behaviour in earlier versions of pandas).
[default: truncate] [currently: truncate]

`display.max_categories` : int
This sets the maximum number of categories pandas should output when printing out a `Categorical` or a Series of dtype "category".
[default: 8] [currently: 8]

`display.max_columns` : int
If `max_cols` is exceeded, switch to truncate view. Depending on `large_repr`, objects are either centrally truncated or printed as a summary view. 'None' value means unlimited.

In case python/IPython is running in a terminal and `large_repr` equals 'truncate' this can be set to 0 or None and pandas will auto-detect the width of the terminal and print a truncated object which fits the screen width. The IPython notebook, IPython qtconsole, or IDLE do not run in a terminal and hence it is not possible to do correct auto-detection and defaults to 20.
[default: 20] [currently: 100]

`display.max_colwidth` : int or None
The maximum width in characters of a column in the repr of a pandas data structure. When the column overflows, a "..." placeholder is embedded in the output. A 'None' value means unlimited.
[default: 50] [currently: 50]

`display.max_dir_items` : int
The number of items that will be added to `dir(...)`. 'None' value means unlimited. Because dir is cached, changing this option will not immediately affect already existing dataframes until a column is deleted or added.

This is for instance used to suggest columns from a dataframe to tab completion.
[default: 100] [currently: 100]

`display.max_info_columns` : int
`max_info_columns` is used in `DataFrame.info` method to decide if per column information will be printed.
[default: 100] [currently: 100]

`display.max_info_rows` : int
`df.info()` will usually show null-counts for each column. For large frames this can be quite slow. `max_info_rows` and `max_info_cols` limit this null check only to frames with smaller dimensions than specified.
[default: 1690785] [currently: 1690785]

`display.max_rows` : int
If `max_rows` is exceeded, switch to truncate view. Depending on `large_repr`, objects are either centrally truncated or printed as a summary view. 'None' value means unlimited.

In case python/IPython is running in a terminal and `large_repr` equals 'truncate' this can be set to 0 and pandas will auto-detect the height of the terminal and print a truncated object which fits the screen height. The IPython notebook, IPython qtconsole, or IDLE do not run in a terminal and hence it is not possible to do correct auto-detection.
[default: 60] [currently: None]

`display.max_seq_items` : int or None
When pretty-printing a long sequence, no more than `max_seq_items` will be printed. If items are omitted, they will be denoted by the addition of "..." to the resulting string.

If set to None, the number of items to be printed is unlimited.

```

    [default: 100] [currently: 100]
display.memory_usage : bool, string or None
    This specifies if the memory usage of a DataFrame should be displayed when
    df.info() is called. Valid values True,False,'deep'
    [default: True] [currently: True]
display.min_rows : int
    The numbers of rows to show in a truncated view (when `max_rows` is
    exceeded). Ignored when `max_rows` is set to None or 0. When set to
    None, follows the value of `max_rows`.
    [default: 10] [currently: 10]
display.multi_sparse : boolean
    "sparsify" MultiIndex display (don't display repeated
    elements in outer levels within groups)
    [default: True] [currently: True]
display.notebook_repr_html : boolean
    When True, IPython notebook will use html representation for
    pandas objects (if it is available).
    [default: True] [currently: True]
display.pprint_nest_depth : int
    Controls the number of nested levels to process when pretty-printing
    [default: 3] [currently: 3]
display.precision : int
    Floating point output precision in terms of number of places after the
    decimal, for regular formatting as well as scientific notation. Similar
    to ``precision`` in :meth:`numpy.set_printoptions`.
    [default: 6] [currently: 6]
display.show_dimensions : boolean or 'truncate'
    Whether to print out dimensions at the end of DataFrame repr.
    If 'truncate' is specified, only print out the dimensions if the
    frame is truncated (e.g. not display all rows and/or columns)
    [default: truncate] [currently: truncate]
display.unicode.ambiguous_as_wide : boolean
    Whether to use the Unicode East Asian Width to calculate the display text
    width.
    Enabling this may affect to the performance (default: False)
    [default: False] [currently: False]
display.unicode.east_asian_width : boolean
    Whether to use the Unicode East Asian Width to calculate the display text
    width.
    Enabling this may affect to the performance (default: False)
    [default: False] [currently: False]
display.width : int
    Width of the display in characters. In case python/IPython is running in
    a terminal this can be set to None and pandas will correctly auto-detect
    the width.
    Note that the IPython notebook, IPython qtconsole, or IDLE do not run in a
    terminal and hence it is not possible to correctly detect the width.
    [default: 80] [currently: 150]
future.infer_string Whether to infer sequence of str objects as pyarrow string dt
ype, which will be the default in pandas 3.0 (at which point this option will be
deprecated).
    [default: False] [currently: False]
future.no_silent_downcasting Whether to opt-in to the future behavior which will
*not* silently downcast results from Series and DataFrame `where`, `mask`, and `c
lip` methods. Silent downcasting will be removed in pandas 3.0 (at which point th
is option will be deprecated).
    [default: False] [currently: False]
io.excel.ods.reader : string
    The default Excel reader engine for 'ods' files. Available options:
    auto, odf, calamine.

```

```

    [default: auto] [currently: auto]
io.excel.ods.writer : string
    The default Excel writer engine for 'ods' files. Available options:
    auto, odf.
    [default: auto] [currently: auto]
io.excel.xls.reader : string
    The default Excel reader engine for 'xls' files. Available options:
    auto, xlrd, calamine.
    [default: auto] [currently: auto]
io.excel.xlsb.reader : string
    The default Excel reader engine for 'xlsb' files. Available options:
    auto, pyxlsb, calamine.
    [default: auto] [currently: auto]
io.excel.xlsm.reader : string
    The default Excel reader engine for 'xlsm' files. Available options:
    auto, xlrd, openpyxl, calamine.
    [default: auto] [currently: auto]
io.excel.xlsm.writer : string
    The default Excel writer engine for 'xlsm' files. Available options:
    auto, openpyxl.
    [default: auto] [currently: auto]
io.excel.xlsx.reader : string
    The default Excel reader engine for 'xlsx' files. Available options:
    auto, xlrd, openpyxl, calamine.
    [default: auto] [currently: auto]
io.excel.xlsx.writer : string
    The default Excel writer engine for 'xlsx' files. Available options:
    auto, openpyxl, xlsxwriter.
    [default: auto] [currently: auto]
io.hdf.default_format : format
    default format writing format, if None, then
    put will default to 'fixed' and append will default to 'table'
    [default: None] [currently: None]
io.hdf.dropna_table : boolean
    drop ALL nan rows when appending to a table
    [default: False] [currently: False]
io.parquet.engine : string
    The default parquet reader/writer engine. Available options:
    'auto', 'pyarrow', 'fastparquet', the default is 'auto'
    [default: auto] [currently: auto]
io.sql.engine : string
    The default sql reader/writer engine. Available options:
    'auto', 'sqlalchemy', the default is 'auto'
    [default: auto] [currently: auto]
mode.chained_assignment : string
    Raise an exception, warn, or no action if trying to use chained assignment,
    The default is warn
    [default: warn] [currently: warn]
mode.copy_on_write : bool
    Use new copy-view behaviour using Copy-on-Write. Defaults to False,
    unless overridden by the 'PANDAS_COPY_ON_WRITE' environment variable
    (if set to "1" for True, needs to be set before pandas is imported).
    [default: False] [currently: False]
mode.data_manager : string
    Internal data manager type; can be "block" or "array". Defaults to "block",
    unless overridden by the 'PANDAS_DATA_MANAGER' environment variable (needs
    to be set before pandas is imported).
    [default: block] [currently: block]
    (Deprecated, use `` instead.)
mode.sim_interactive : boolean

```

Whether to simulate interactive mode for purposes of testing
[default: False] [currently: False]

mode.string_storage : string
The default storage for StringDtype. This option is ignored if
`future.infer_string` is set to True.
[default: python] [currently: python]

mode.use_inf_as_na : boolean
True means treat None, NaN, INF, -INF as NA (old way),
False means None and NaN are null, but INF, -INF are not NA
(new way).

This option is deprecated in pandas 2.1.0 and will be removed in 3.0.
[default: False] [currently: False]
(Deprecated, use `` instead.)

plotting.backend : str
The plotting backend to use. The default value is "matplotlib", the
backend provided with pandas. Other backends can be specified by
providing the name of the module that implements the backend.
[default: matplotlib] [currently: matplotlib]

plotting.matplotlib.register_converters : bool or 'auto'.
Whether to register converters with matplotlib's units registry for
dates, times, datetimes, and Periods. Toggling to False will remove
the converters, restoring any converters that pandas overwrote.
[default: auto] [currently: auto]

styler.format.decimal : str
The character representation for the decimal separator for floats and comple
x.
[default: .] [currently: .]

styler.format.escape : str, optional
Whether to escape certain characters according to the given context; html or
latex.
[default: None] [currently: None]

styler.format.formatter : str, callable, dict, optional
A formatter object to be used as default within ``Styler.format``.
[default: None] [currently: None]

styler.format.na_rep : str, optional
The string representation for values identified as missing.
[default: None] [currently: None]

styler.format.precision : int
The precision for floats and complex numbers.
[default: 6] [currently: 6]

styler.format.thousands : str, optional
The character representation for thousands separator for floats, int and comp
lex.
[default: None] [currently: None]

styler.html.mathjax : bool
If False will render special CSS classes to table attributes that indicate Ma
thjax
will not be used in Jupyter Notebook.
[default: True] [currently: True]

styler.latex.environment : str
The environment to replace ``\begin{table}``. If "longtable" is used results
in a specific longtable environment format.
[default: None] [currently: None]

styler.latex.hrules : bool
Whether to add horizontal rules on top and bottom and below the headers.
[default: False] [currently: False]

styler.latex.multicol_align : {"r", "c", "l", "naive-l", "naive-r"}
The specifier for horizontal alignment of sparsified LaTeX multicolumns. Pipe
decorators can also be added to non-naive values to draw vertical

rules, e.g. "`\|r`" will draw a rule on the left side of right aligned merged cells.

[default: r] [currently: r]
styler.latex.multiprow_align : {"c", "t", "b"}
The specifier for vertical alignment of sparsified LaTeX multirows.
[default: c] [currently: c]
styler.render.encoding : str
The encoding used for output HTML and LaTeX files.
[default: utf-8] [currently: utf-8]
styler.render.max_columns : int, optional
The maximum number of columns that will be rendered. May still be reduced to satisfy `max_elements`, which takes precedence.
[default: None] [currently: None]
styler.render.max_elements : int
The maximum number of data-cell (<td>) elements that will be rendered before trimming will occur over columns, rows or both if needed.
[default: 262144] [currently: 262144]
styler.render.max_rows : int, optional
The maximum number of rows that will be rendered. May still be reduced to satisfy `max_elements`, which takes precedence.
[default: None] [currently: None]
styler.render.repr : str
Determine which output to use in Jupyter Notebook in {"html", "latex"}.
[default: html] [currently: html]
styler.sparse.columns : bool
Whether to sparsify the display of hierarchical columns. Setting to False will
1
display each explicit level element in a hierarchical key for each column.
[default: True] [currently: True]
styler.sparse.index : bool
Whether to sparsify the display of a hierarchical index. Setting to False will
1
display each explicit level element in a hierarchical key for each row.
[default: True] [currently: True]

```
In [37]: pd.set_option('display.max_columns', 100)
```

```
In [38]: pd.set_option('display.width', 150)
```

```
In [39]: pd.set_option('display.max_rows', None)
```

세션별 정확도(I-Impedance Model, S-Siffness shifting algorithm)

```
In [40]: # 결과를 저장할 리스트 초기화
comparison_results = []

# 각 CSV 파일에 대해 작업 수행
for csv_file in csv_files:
    file_path = os.path.join(directory_path, csv_file)

    # CSV 파일 읽기
    df = pd.read_csv(file_path)

    # maxforce와 result 열이 있는지 확인
    if 'maxforce' in df.columns and 'result' in df.columns:
        # 상관관계수 계산
        correlation = df['maxforce'].corr(df['result'])

        # 차이의 절대값 평균 계산
```

```

mean_absolute_difference = (df['maxforce'] - df['result']).abs().mean()

# maxforce와 result가 동일한 값의 비율 계산
identical_percentage = (df['maxforce'] == df['result']).mean() * 100

# 결과 저장
comparison_results.append({
    'filename': csv_file,
    'correlation': correlation,
    'mean_absolute_difference': mean_absolute_difference,
    'identical_percentage': identical_percentage
})

# # 그래프 생성
# plt.figure(figsize=(10, 6))

# # 산점도 (scatter plot)
# plt.scatter(df.index, df['maxforce'], label='Maxforce', color='blue',
# # plt.scatter(df.index, df['result'], label='Result', color='orange', al

# # 선 그래프 (line plot)
# plt.plot(df.index, df['maxforce'], color='blue', alpha=0.3)
# plt.plot(df.index, df['result'], color='orange', alpha=0.3)

# # 제목 및 라벨 추가
# plt.title(f'Result vs Maxforce - {csv_file}')
# plt.xlabel('Index')
# plt.ylabel('Values')
# plt.legend()

else:
    print(f"'maxforce' or 'result' column missing in {csv_file}")

# 결과를 데이터프레임으로 변환
results_df = pd.DataFrame(comparison_results)

# 결과 출력
print(results_df)

```

	filename	correlation	mean_absolute_difference	identical_p
percentage				
0	dohoon_I_1_combined.csv	0.993892	0.008	
96.0				
1	dohoon_I_2_combined.csv	1.000000	0.000	
100.0				
2	dohoon_I_3_combined.csv	0.993877	0.008	
96.0				
3	dohoon_I_4_combined.csv	1.000000	0.000	
100.0				
4	dohoon_S_1_combined.csv	0.994014	0.008	
96.0				
5	dohoon_S_2_combined.csv	0.960668	0.028	
92.0				
6	dohoon_S_3_combined.csv	0.993892	0.008	
96.0				
7	dohoon_S_4_combined.csv	0.965275	0.020	
96.0				
8	jaeho_I_1_combined.csv	0.993883	0.008	
96.0				
9	jaeho_I_2_combined.csv	0.993877	0.008	
96.0				
10	jaeho_I_3_combined.csv	0.901517	0.088	
80.0				
11	jaeho_I_4_combined.csv	0.918953	0.088	
68.0				
12	jaeho_S_1_combined.csv	0.843943	0.104	
68.0				
13	jaeho_S_2_combined.csv	1.000000	0.000	
100.0				
14	jaeho_S_3_combined.csv	0.987500	0.016	
92.0				
15	jaeho_S_4_combined.csv	0.920785	0.072	
76.0				
16	jaewan_I_1_combined.csv	0.907143	0.064	
80.0				
17	jaewan_I_2_combined.csv	0.746634	0.136	
68.0				
18	jaewan_I_3_combined.csv	0.656684	0.184	
56.0				
19	jaewan_I_4_combined.csv	0.796246	0.112	
68.0				
20	jaewan_S_1_combined.csv	0.946608	0.036	
88.0				
21	jaewan_S_2_combined.csv	0.983258	0.024	
88.0				
22	jaewan_S_3_combined.csv	0.981416	0.024	
88.0				
23	jaewan_S_4_combined.csv	0.974279	0.040	
80.0				
24	jiyoung_I_1_combined.csv	0.994014	0.008	
96.0				
25	jiyoung_I_2_combined.csv	0.935393	0.048	
88.0				
26	jiyoung_I_3_combined.csv	0.930587	0.056	
84.0				
27	jiyoung_I_4_combined.csv	0.993877	0.008	
96.0				
28	jiyoung_S_1_combined.csv	0.929386	0.068	
72.0				

29	jiyoung_S_2_combined.csv	0.988428	0.016
92.0			
30	jiyoung_S_3_combined.csv	0.922031	0.080
72.0			
31	jiyoung_S_4_combined.csv	0.993883	0.008
96.0			
32	minho_I_1_combined.csv	0.829396	0.112
64.0			
33	minho_I_2_combined.csv	0.960760	0.052
76.0			
34	minho_I_3_combined.csv	0.978370	0.032
84.0			
35	minho_I_4_combined.csv	0.843053	0.140
52.0			
36	minho_S_1_combined.csv	0.993877	0.008
96.0			
37	minho_S_2_combined.csv	0.983317	0.024
88.0			
38	minho_S_3_combined.csv	0.939945	0.044
84.0			
39	minho_S_4_combined.csv	0.988290	0.016
92.0			
40	nayoung_I_1_combined.csv	0.631618	0.176
60.0			
41	nayoung_I_2_combined.csv	0.887813	0.148
48.0			
42	nayoung_I_3_combined.csv	-0.199253	0.452
16.0			
43	nayoung_I_4_combined.csv	0.755987	0.152
48.0			
44	nayoung_S_1_combined.csv	0.979792	0.028
84.0			
45	nayoung_S_2_combined.csv	0.848014	0.092
72.0			
46	nayoung_S_3_combined.csv	0.915326	0.060
80.0			
47	nayoung_S_4_combined.csv	0.900424	0.064
80.0			
48	seongjun_I_1_combined.csv	0.994014	0.008
96.0			
49	seongjun_I_2_combined.csv	1.000000	0.000
100.0			
50	seongjun_I_3_combined.csv	1.000000	0.000
100.0			
51	seongjun_I_4_combined.csv	1.000000	0.000
100.0			
52	seongjun_S_1_combined.csv	0.900718	0.072
76.0			
53	seongjun_S_2_combined.csv	1.000000	0.000
100.0			
54	seongjun_S_3_combined.csv	0.887474	0.092
72.0			
55	seongjun_S_4_combined.csv	0.988428	0.016
92.0			
56	seoyeong_I_1_combined.csv	0.755282	0.128
72.0			
57	seoyeong_I_2_combined.csv	0.814612	0.124
72.0			
58	seoyeong_I_3_combined.csv	0.866834	0.080
84.0			

59	seoyeong_I_4_combined.csv	0.920382	0.040
92.0			
60	seoyeong_S_1_combined.csv	0.669061	0.156
56.0			
61	seoyeong_S_2_combined.csv	0.782878	0.116
72.0			
62	seoyeong_S_3_combined.csv	0.867168	0.076
80.0			
63	seoyeong_S_4_combined.csv	0.979068	0.032
84.0			
64	taeyoon_I_1_combined.csv	0.834818	0.088
80.0			
65	taeyoon_I_2_combined.csv	0.866834	0.080
84.0			
66	taeyoon_I_3_combined.csv	0.803441	0.124
68.0			
67	taeyoon_I_4_combined.csv	0.805299	0.116
72.0			
68	taeyoon_S_1_combined.csv	0.427669	0.276
32.0			
69	taeyoon_S_2_combined.csv	0.515988	0.236
48.0			
70	taeyoon_S_3_combined.csv	0.767243	0.136
60.0			
71	taeyoon_S_4_combined.csv	0.619546	0.184
56.0			

```
In [41]: # S와 I에 해당하는 identical_percentage 저장 딕셔너리
identical_percentages = {'S1': [], 'S2': [], 'S3': [], 'S4': [],
                          'I1': [], 'I2': [], 'I3': [], 'I4': []}

# 각 CSV 파일에 대해 작업 수행
for csv_file in csv_files:
    file_path = os.path.join(directory_path, csv_file)

    # CSV 파일 읽기
    df = pd.read_csv(file_path)

    # maxforce와 result 열이 있는지 확인
    if 'maxforce' in df.columns and 'result' in df.columns:
        # maxforce와 result가 동일한 값의 비율 계산
        identical_percentage = (df['maxforce'] == df['result']).mean() * 100

    # 파일 이름에서 S1, S2, S3, S4 또는 I1, I2, I3, I4를 추출하여 해당 리스트
    if '_S_1_' in csv_file:
        identical_percentages['S1'].append(identical_percentage)
    elif '_S_2_' in csv_file:
        identical_percentages['S2'].append(identical_percentage)
    elif '_S_3_' in csv_file:
        identical_percentages['S3'].append(identical_percentage)
    elif '_S_4_' in csv_file:
        identical_percentages['S4'].append(identical_percentage)
    elif '_I_1_' in csv_file:
        identical_percentages['I1'].append(identical_percentage)
    elif '_I_2_' in csv_file:
        identical_percentages['I2'].append(identical_percentage)
    elif '_I_3_' in csv_file:
        identical_percentages['I3'].append(identical_percentage)
    elif '_I_4_' in csv_file:
        identical_percentages['I4'].append(identical_percentage)
```

```

else:
    print(f"'maxforce' or 'result' column missing in {csv_file}")

# S1, S2, S3, S4, I1, I2, I3, I4의 평균 계산
average_S1 = sum(identical_percentages['S1']) / len(identical_percentages['S1'])
average_S2 = sum(identical_percentages['S2']) / len(identical_percentages['S2'])
average_S3 = sum(identical_percentages['S3']) / len(identical_percentages['S3'])
average_S4 = sum(identical_percentages['S4']) / len(identical_percentages['S4'])

average_I1 = sum(identical_percentages['I1']) / len(identical_percentages['I1'])
average_I2 = sum(identical_percentages['I2']) / len(identical_percentages['I2'])
average_I3 = sum(identical_percentages['I3']) / len(identical_percentages['I3'])
average_I4 = sum(identical_percentages['I4']) / len(identical_percentages['I4'])
# S와 I의 평균을 리스트로 저장
averages_S = [average_S1, average_S2, average_S3, average_S4]
averages_I = [average_I1, average_I2, average_I3, average_I4]

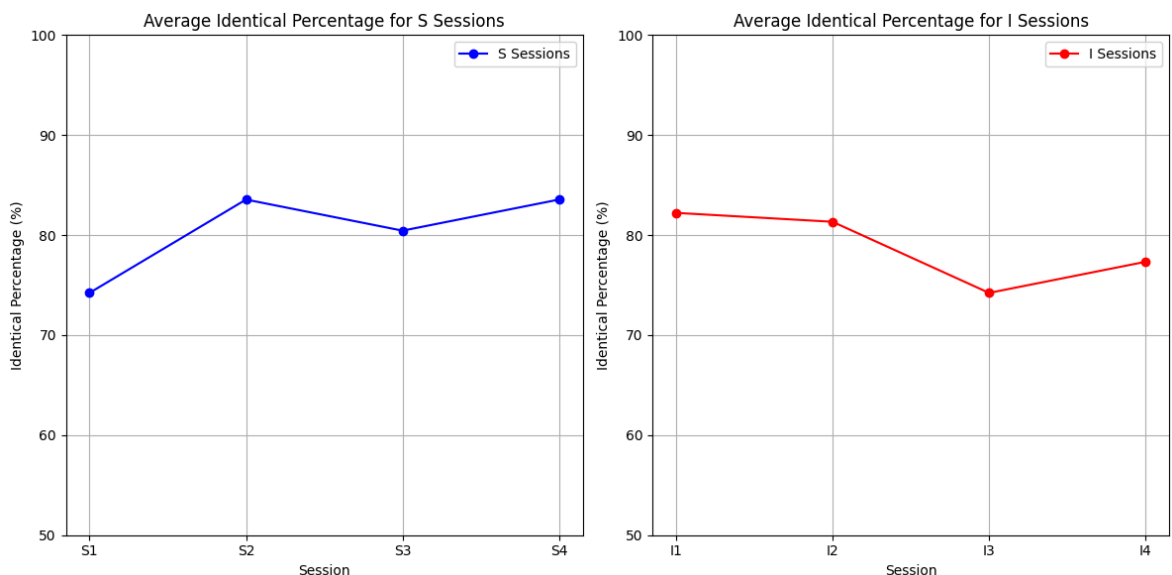
# 두 개의 그래프 생성
fig, axs = plt.subplots(1, 2, figsize=(12, 6))

# S 세션 그래프 (왼쪽) - 파란색
axs[0].plot(['S1', 'S2', 'S3', 'S4'], averages_S, marker='o', color='blue', label='S Sessions')
axs[0].set_title('Average Identical Percentage for S Sessions')
axs[0].set_xlabel('Session')
axs[0].set_ylabel('Identical Percentage (%)')
axs[0].grid(True)
axs[0].legend()
axs[0].set_ylim(50, 100) # y축 범위를 50%~100%로 설정

# I 세션 그래프 (오른쪽) - 빨간색
axs[1].plot(['I1', 'I2', 'I3', 'I4'], averages_I, marker='o', color='red', label='I Sessions')
axs[1].set_title('Average Identical Percentage for I Sessions')
axs[1].set_xlabel('Session')
axs[1].set_ylabel('Identical Percentage (%)')
axs[1].grid(True)
axs[1].legend()
axs[1].set_ylim(50, 100) # y축 범위를 50%~100%로 설정

# 그래프 출력
plt.tight_layout()
plt.show()

```



```

In [42]: # 각 횟수별로 identical_percentage 저장 딕셔너리
identical_percentages = {'Session': [], 'Percentage': [], 'Count': []}

# 각 CSV 파일에 대해 작업 수행
for csv_file in csv_files:
    file_path = os.path.join(directory_path, csv_file)

    # CSV 파일 읽기
    df = pd.read_csv(file_path)

    # maxforce와 result 열이 있는지 확인
    if 'maxforce' in df.columns and 'result' in df.columns:
        # maxforce와 result가 동일한 값의 비율 계산
        identical_percentage = (df['maxforce'] == df['result']).mean() * 100

    # 파일 이름에서 S1, S2, S3, S4 또는 I1, I2, I3, I4를 추출하여 데이터 추가
    if '_S_1_' in csv_file:
        identical_percentages['Session'].append('S')
        identical_percentages['Count'].append('1')
    elif '_S_2_' in csv_file:
        identical_percentages['Session'].append('S')
        identical_percentages['Count'].append('2')
    elif '_S_3_' in csv_file:
        identical_percentages['Session'].append('S')
        identical_percentages['Count'].append('3')
    elif '_S_4_' in csv_file:
        identical_percentages['Session'].append('S')
        identical_percentages['Count'].append('4')
    elif '_I_1_' in csv_file:
        identical_percentages['Session'].append('I')
        identical_percentages['Count'].append('1')
    elif '_I_2_' in csv_file:
        identical_percentages['Session'].append('I')
        identical_percentages['Count'].append('2')
    elif '_I_3_' in csv_file:
        identical_percentages['Session'].append('I')
        identical_percentages['Count'].append('3')
    elif '_I_4_' in csv_file:
        identical_percentages['Session'].append('I')
        identical_percentages['Count'].append('4')

    identical_percentages['Percentage'].append(identical_percentage)

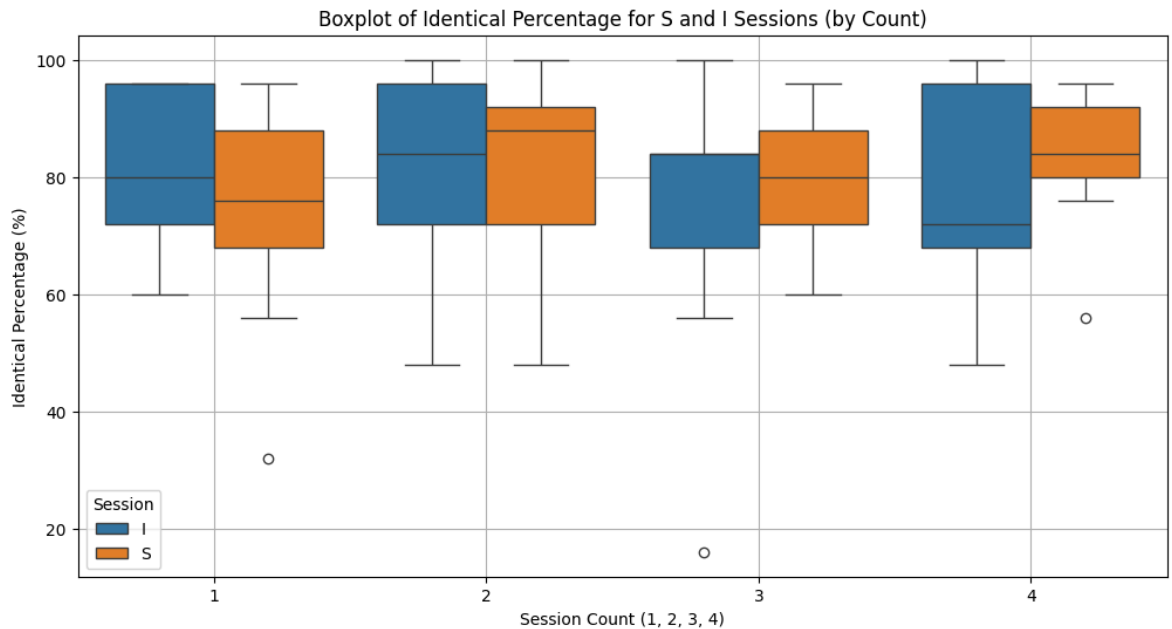
# 데이터프레임으로 변환
df_percentages = pd.DataFrame(identical_percentages)

# Boxplot 생성
plt.figure(figsize=(12, 6))
sns.boxplot(x='Count', y='Percentage', hue='Session', data=df_percentages)

# 그래프 세부 설정
plt.title('Boxplot of Identical Percentage for S and I Sessions (by Count)')
plt.xlabel('Session Count (1, 2, 3, 4)')
plt.ylabel('Identical Percentage (%)')
plt.grid(True)

# 그래프 출력
plt.show()

```



In []:

🔗 렌더링별 각 데이터 합산 결과

```
In [43]: import re

# E_combined와 S_combined 파일 리스트 필터링
i_combined_files = [f for f in os.listdir(directory_path) if re.search(r'I_\d+_c', f)]
s_combined_files = [f for f in os.listdir(directory_path) if re.search(r'S_\d+_c', f)]

# E_combined와 S_combined 데이터프레임 각각 합치기
df_i_combined = pd.concat([pd.read_csv(os.path.join(directory_path, file)) for f in i_combined_files])
df_s_combined = pd.concat([pd.read_csv(os.path.join(directory_path, file)) for f in s_combined_files])

# 데이터프레임을 합친 후 각각 동일한 작업 수행
for df_combined, combined_name in [(df_i_combined, 'I_combined'), (df_s_combined, 'S_combined')]:

    # maxforce와 result 열이 있는지 확인
    if 'maxforce' in df_combined.columns and 'result' in df_combined.columns:
        maxforce_values = [0.0, 0.1, 0.3, 0.5, 1.0]
        result_values = [0.0, 0.1, 0.3, 0.5, 1.0]

        distribution = []

        for maxforce_value in maxforce_values:
            for result_value in result_values:
                count = len(df_combined[(df_combined['maxforce'] == maxforce_value) & (df_combined['result'] == result_value)])
                distribution.append({
                    'maxforce_value': maxforce_value,
                    'result_value': result_value,
                    'count': count
                })

        distribution_df = pd.DataFrame(distribution)

# 막대 그래프 생성
plt.figure(figsize=(10, 6))
sns.barplot(x='result_value', y='count', hue='maxforce_value', data=distribution_df)
```



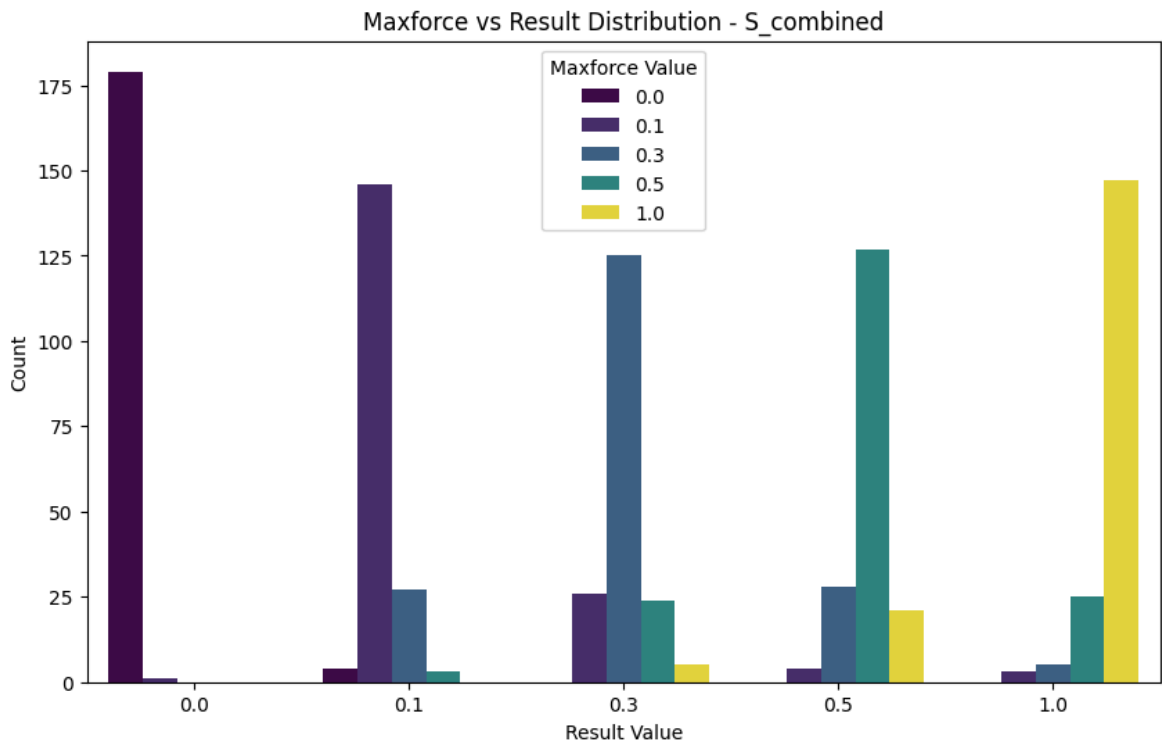
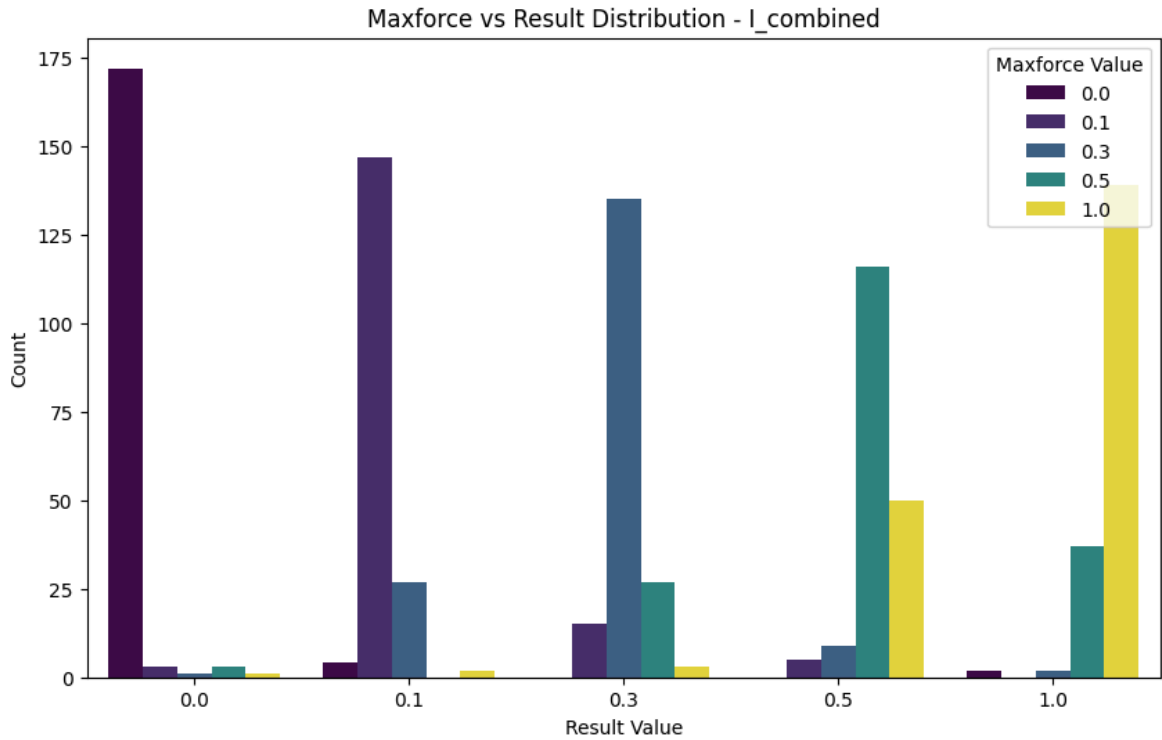
```

# 그래프 세부 설정
plt.title(f'Maxforce vs Result Distribution - {combined_name}')
plt.xlabel('Result Value')
plt.ylabel('Count')
plt.legend(title='Maxforce Value')

# 그래프 출력
plt.show()

else:
    print(f"'maxforce' or 'result' column missing in {combined_name}")

```



```

In [44]: # 결과를 저장할 리스트 초기화
comparison_results = []

```

```

# 각 데이터프레임에 대해 작업 수행
for df_combined, combined_name in [(df_i_combined, 'I_combined'), (df_s_combined, 'S_combined')]:

    # maxforce와 result 열이 있는지 확인
    if 'maxforce' in df_combined.columns and 'result' in df_combined.columns:
        # 상관계수 계산
        correlation = df_combined['maxforce'].corr(df_combined['result'])

        # 차이의 절대값 평균 계산
        mean_absolute_difference = (df_combined['maxforce'] - df_combined['result']).abs().mean()

        # maxforce와 result가 동일한 값의 비율 계산
        identical_percentage = (df_combined['maxforce'] == df_combined['result']).sum() / df_combined['maxforce'].count()

        # 결과 저장
        comparison_results.append({
            'filename': combined_name,
            'correlation': correlation,
            'mean_absolute_difference': mean_absolute_difference,
            'identical_percentage': identical_percentage
        })

    else:
        print(f"'maxforce' or 'result' column missing in {combined_name}")

# 결과를 데이터프레임으로 변환
results_df = pd.DataFrame(comparison_results)

# 결과 출력
print(results_df)

```

	filename	correlation	mean_absolute_difference	identical_percentage
0	I_combined	0.855124	0.079889	78.777778
1	S_combined	0.892093	0.063333	80.444444

In []:

In []:

In []:

In []: