

- DataFrame - create pd.DataFrame
 - value: { }
 - 3 rows: 3 dict { }
 - 2 cols: 2 items in 1 dict (key: col. name, value: col. value)

1. DataFrame merge

- add new col: new col. identifier - df[new col. name]
 - value - a list of values
 - missing value + None
 - uniq index - series { dict: {key: index: value} }
 - align
- join 2 large dataframe: pd.merge(df1, df2, method, left_index=True, right_index=True)
 - conflicts: location-x, location-y
 - left_on
 - right_on
 - left/right_on = { }

2. idiomatic pd → Pandorable

- method chaining .rename(columns={col.name: new col.name})
- apply = function - data - (a new row - pd.series {dict: {key: func.name, value: func}})
 - row - col
 - add row - row [new col.name]
 - compare col value per row → df.apply(func, axis=1) → row
- apply + lambda
 - df.apply(lambda x: function, axis=1)

3. Groupby - split up dataframe

- unique = for state in df['STATE'].unique():
 - groupby = for group, frame in df.groupby('STATE')
 - hp.average (df[df['STATE'] == state].dropna()['census2010pop'])
 - frame['census2010pop']
 - time it - speed different
- groupby (function) → split
 - df.reset_index()
 - df.groupby(func)
 - define func
- groupby + func → split - apply pattern
 - agg df.groupby(index).agg({col.name: func})

Apply

- df.groupby(index).apply(lambda df, a, b: func, col.name)
- groupby object (single, list)
 - df.reset_index(index).groupby(level=0)[col.name].agg({name: func}).
 - a list of values

4. Scale - data type

- col. data - categorical (nominal) - astype('category')
 - logical order - categories = { }, ordered = { }
- interval / ratio - categorical
 - pd.cut (col.value, bins)

5. pivot table: 2 cate. variables - 1 numeric variable - summary

- df.pivot_table(value, index, columns, aggfunc)
 - single
 - list