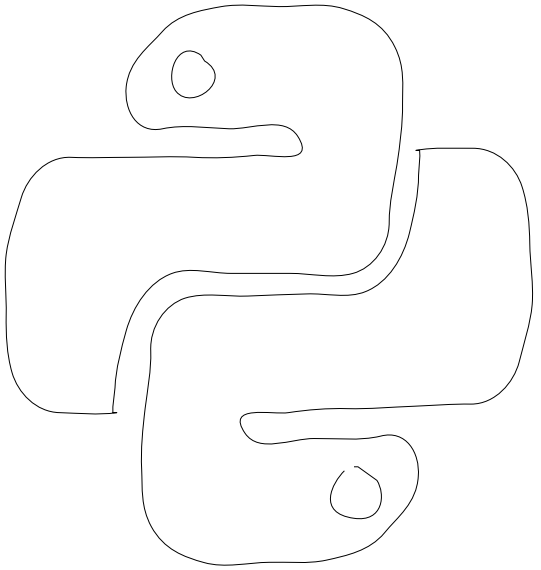


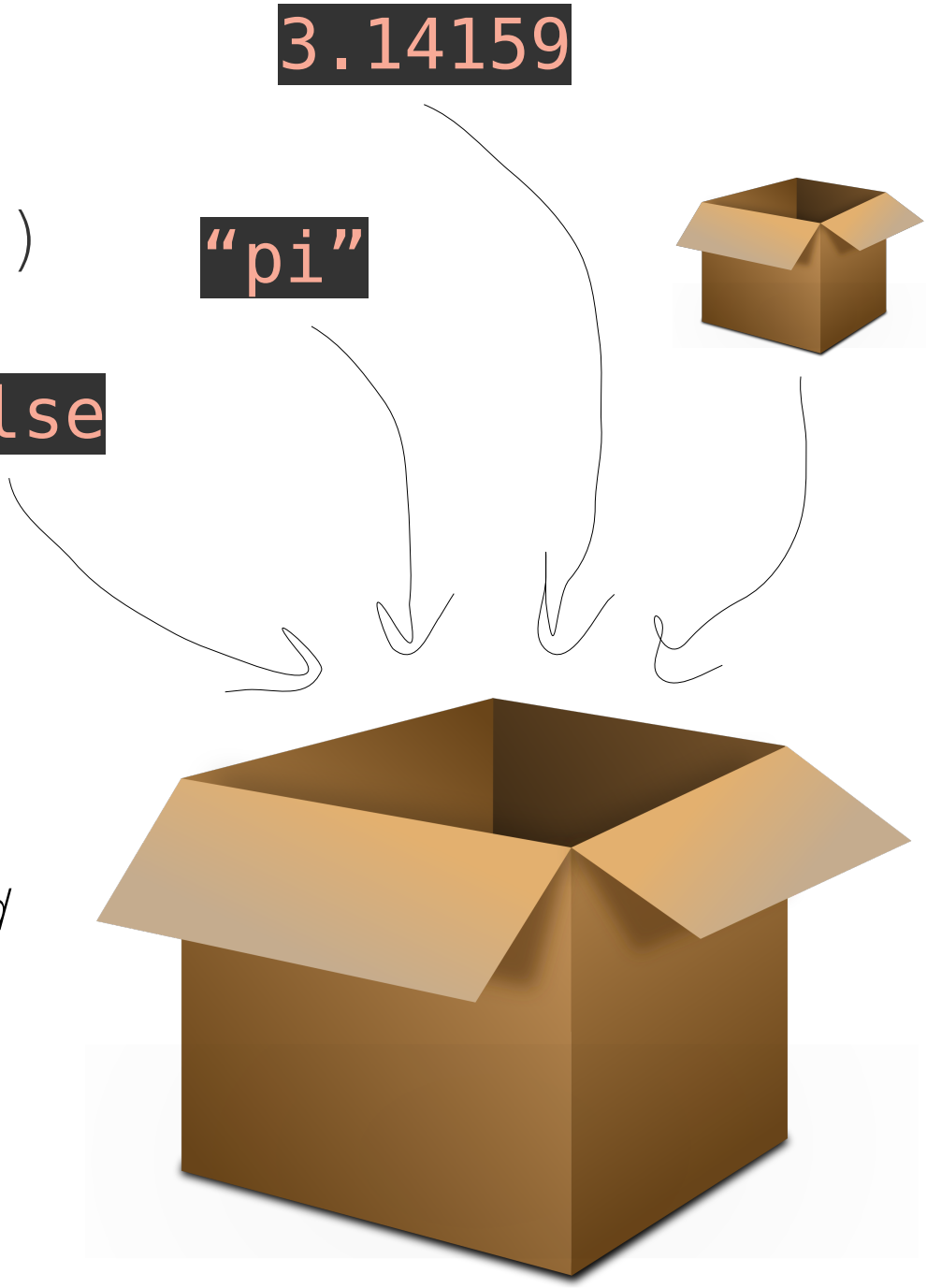
Collection choice by context

Met Office *Python Guild*, August ‘18

Sadie Bartholomew

Terminology

- *collection* (= *container*) data structures
- i.e. non-primitive
- not just the **`collections`** library!



"an object that groups multiple elements into a single unit… used to store, retrieve, manipulate, & communicate aggregate data"

[from Oracle Java Tutorials]

Aims

- review (some) collections available for Python
- highlight characteristics & variety
- summarise considerations for choosing wisely from the *Collections Zoo* for a given context
- ultimately: encourage us to think about containers in Python & how we use them

Scope (for this talk)

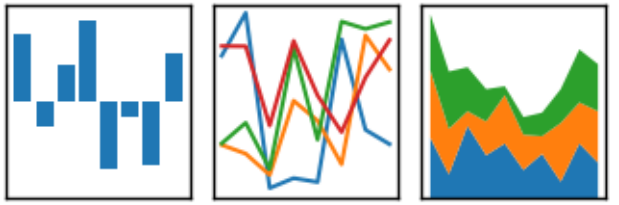
- *built-in* NB exclude strings as collections of characters
- *standard library* dedicated collections libraries:

  1) `collections`
  2) `array`

  NB exclude compound/serialisation data formats e.g. JSON
- established *external* numeric/data libraries (x2):

pandas

$$y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$$



(Limited scope) catalogue

Built-in	From <*module*> import ... collections	array	numpy	pandas
list	deque	array	array	
tuple	namedtuple()			
dict	defaultdict OrderedDict ChainMap Counter			Series DataFrame
set				
frozenset				

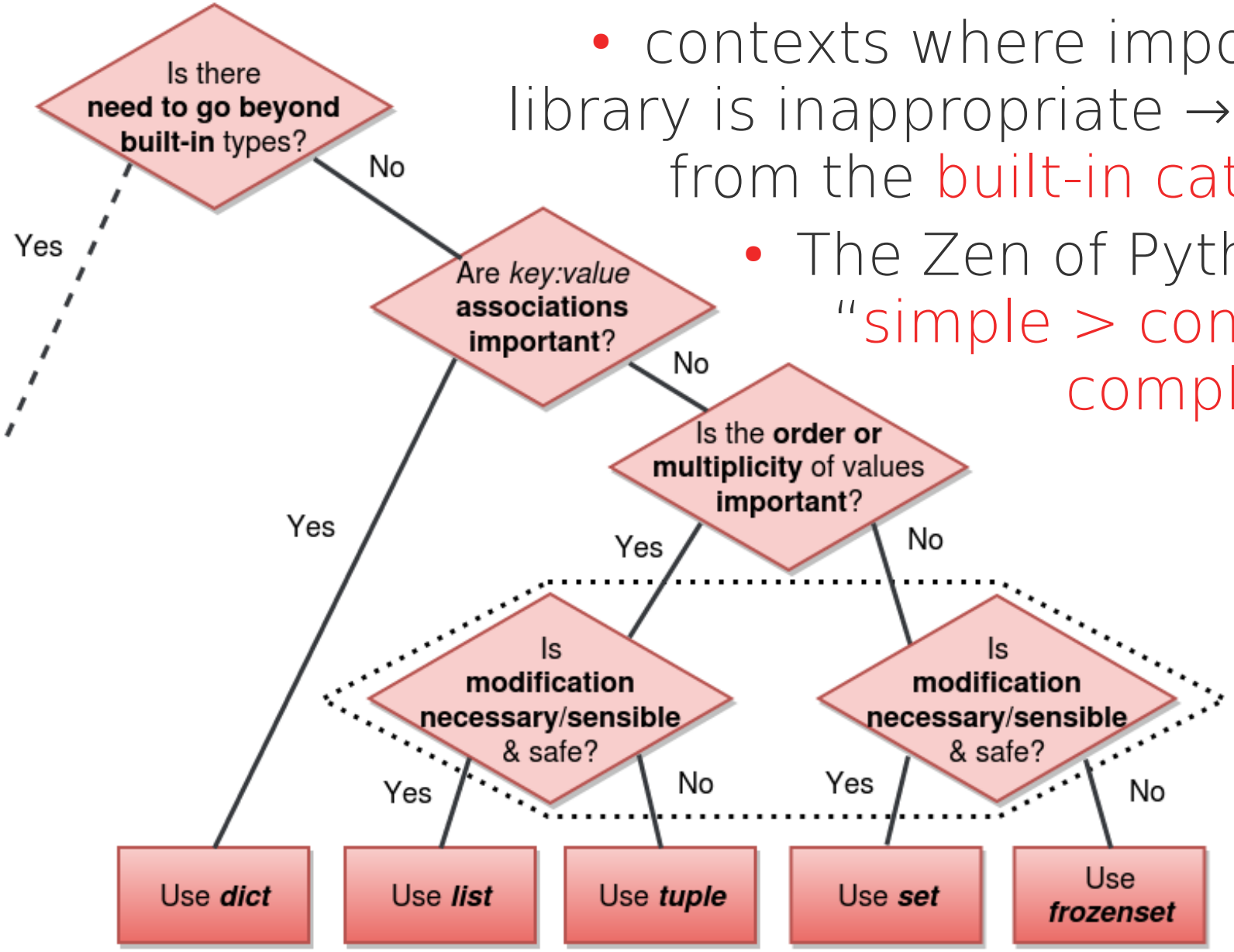
distinguish

based on & extends

+ **UserDict** & **UserList** wrappers (not covered)

- non-built-in collections to be outlined throughout

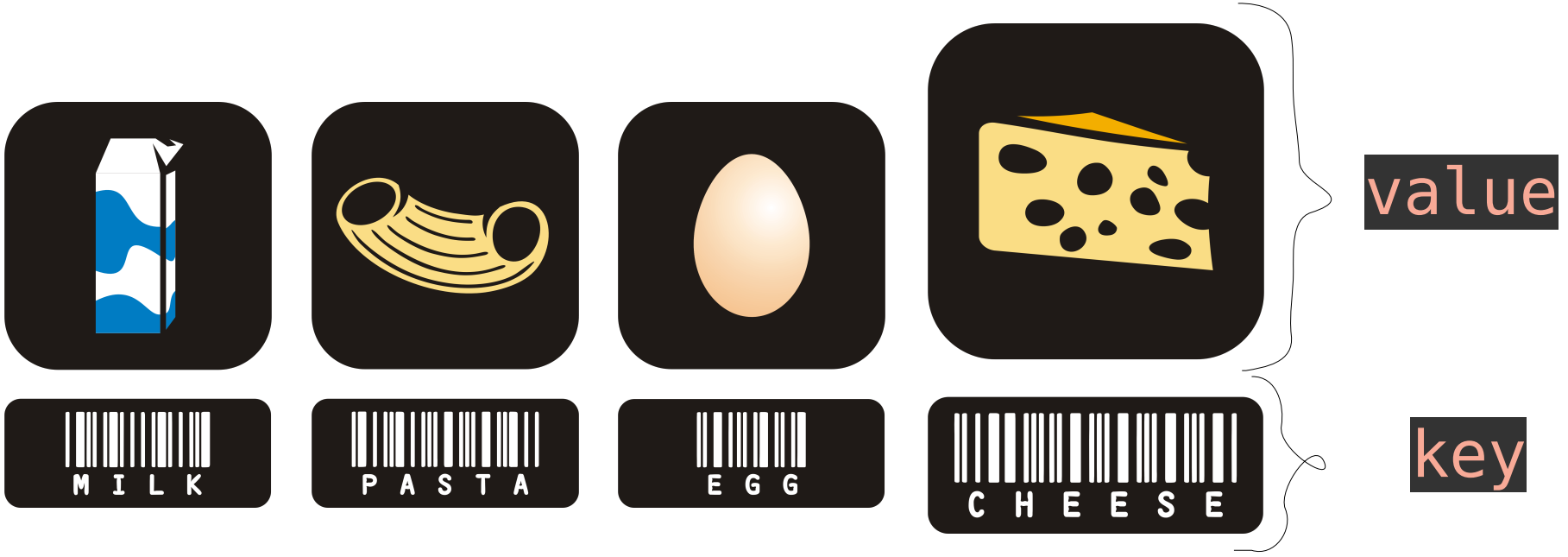
Considerations: Simplicity



- contexts where importing a library is inappropriate → choice from the built-in catalogue
- The Zen of Python #3: "simple > complex > complicated"

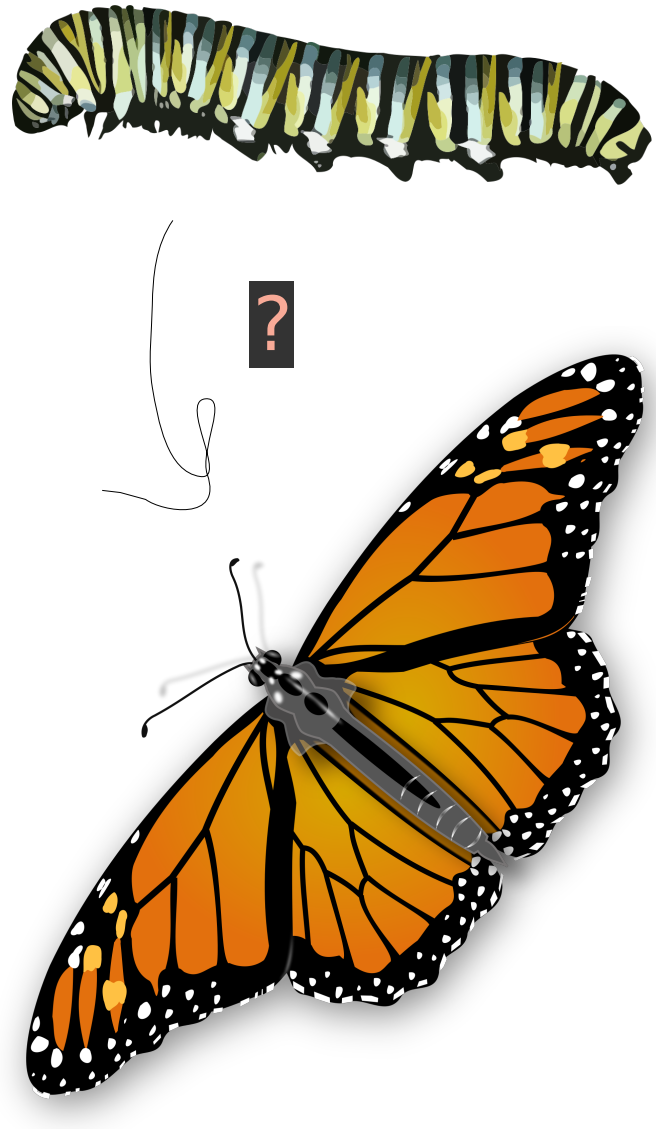
Considerations: Associations

- to associate, or *not to* associate (your values with keys)? keys essential as distinct mappings
- but… also consider as descriptors/labels:
  - ➔ require extra verbosity & memory space (but)
  - ➔ minimise lookup time & improve readability.



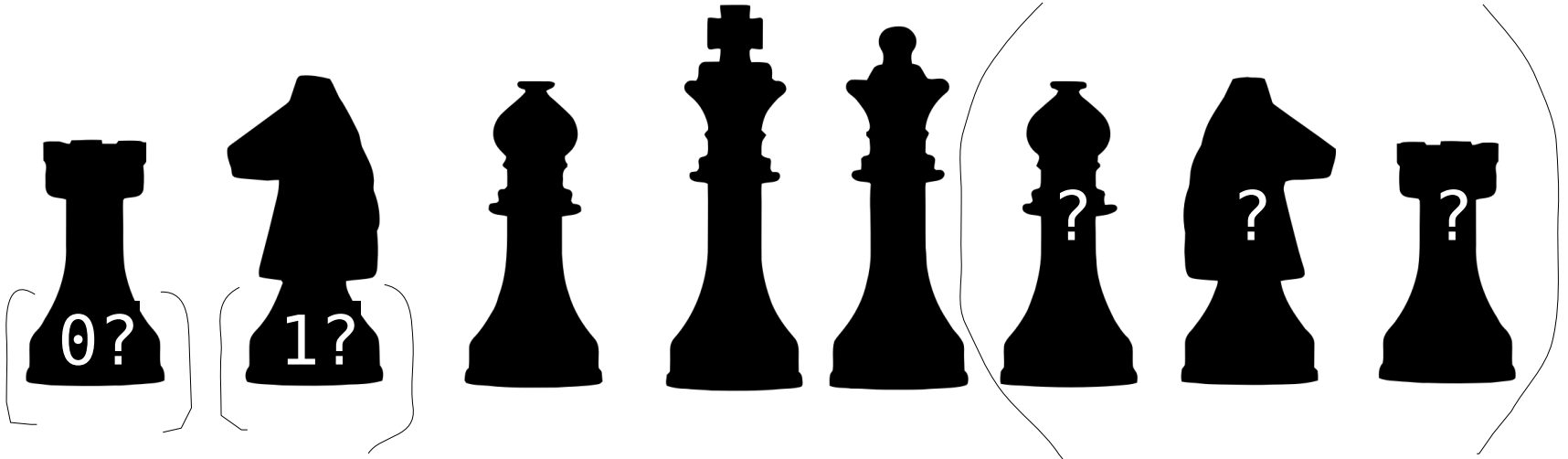
Considerations: Mutability

- to bring about a modification:
  - ➔ *directly alter* the object instance (mutable)
  - ➔ create a *new* object as a *revised copy* (immutable)
- mutable data structures offer:
  - ➔ increased flexibility & avoid expensive copying for alterations (but)
  - ➔ not thread-safe, no identity by state & referencing requires defensive copying



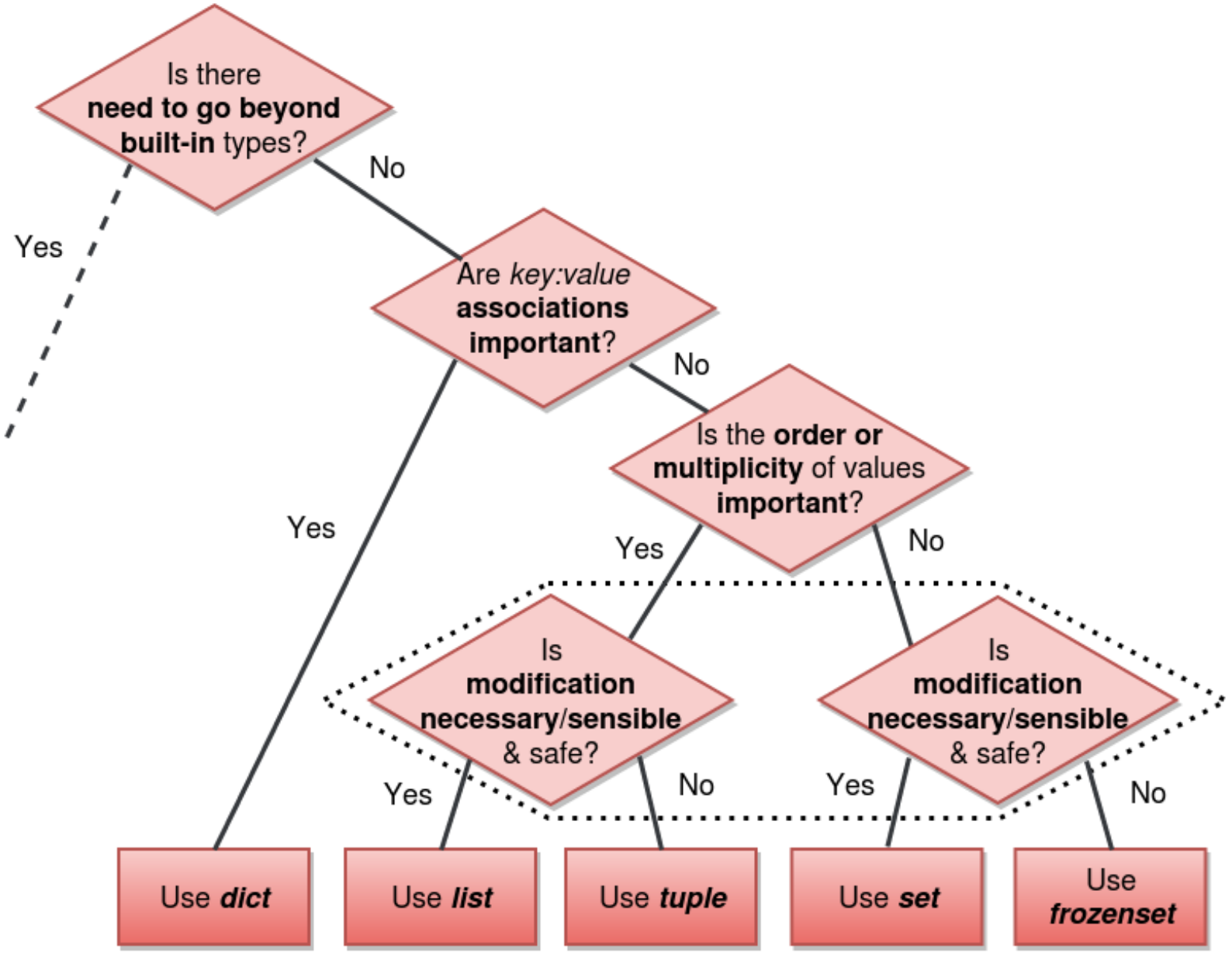
Considerations: Order & Tally

- element order: *position* in the collection
- tally/multiplicity: identical element *count*



- if both irrelevant, consider using **`set`** (or **`dict`**): *unordered* with only *unique* elements (keys)
- **`OrderedDict`** remembers insertion order
- **`Counter`** is a **`dict`** storing counts as values

Considerations: *pause & recap*



Considerations: Readability

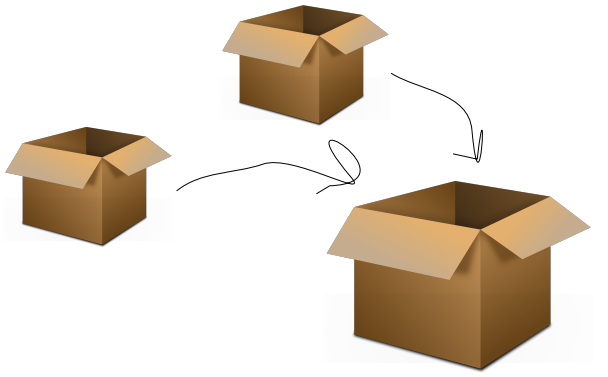
- The Zen of Python #7: "Readability counts."
- difficult to keep track of field identity via index for *long* or *nested* structures or for *like* fields
- named fields & **dict** keys make code easier to read & comprehend (self-documenting)

```
from collections import namedtuple
Box = namedtuple('Box', 'height width length')
box_A = Box(height=10, width=12, length=14)
```

```
box_A = (10, 12, 14)
```

compare to **tuple**

Considerations: Nesting



- The Zen of Python #5:

  "Flat is better than nested."

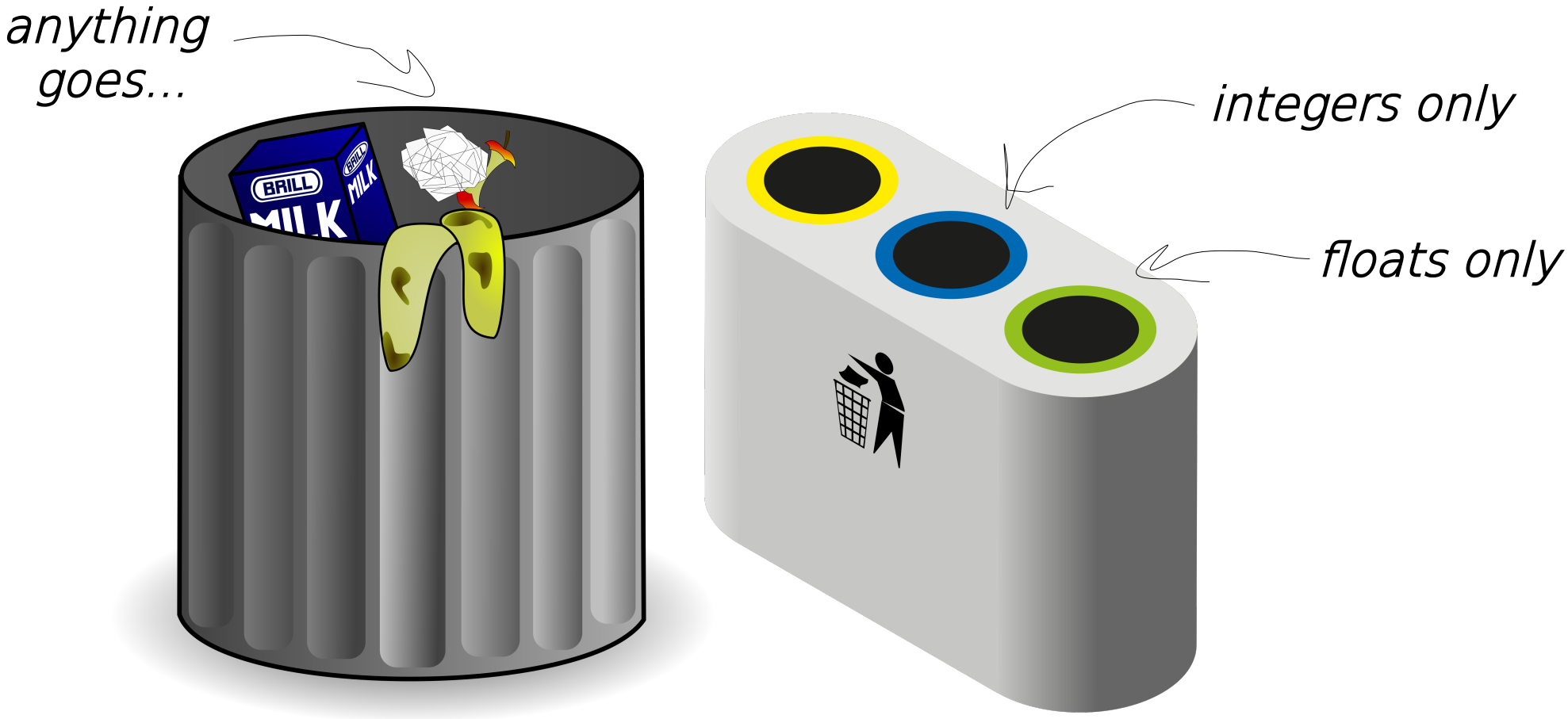
- consider alternatives to nesting e.g. **`numpy.array`** more functional than nested list & layering of keys possible in **`ChainMap`** [both outlined later]

- (shallow) nesting sometimes justified

- for *top-level* choice, note **`set`** elements & **`dict`** keys cannot be mutable, restricting nesting potential

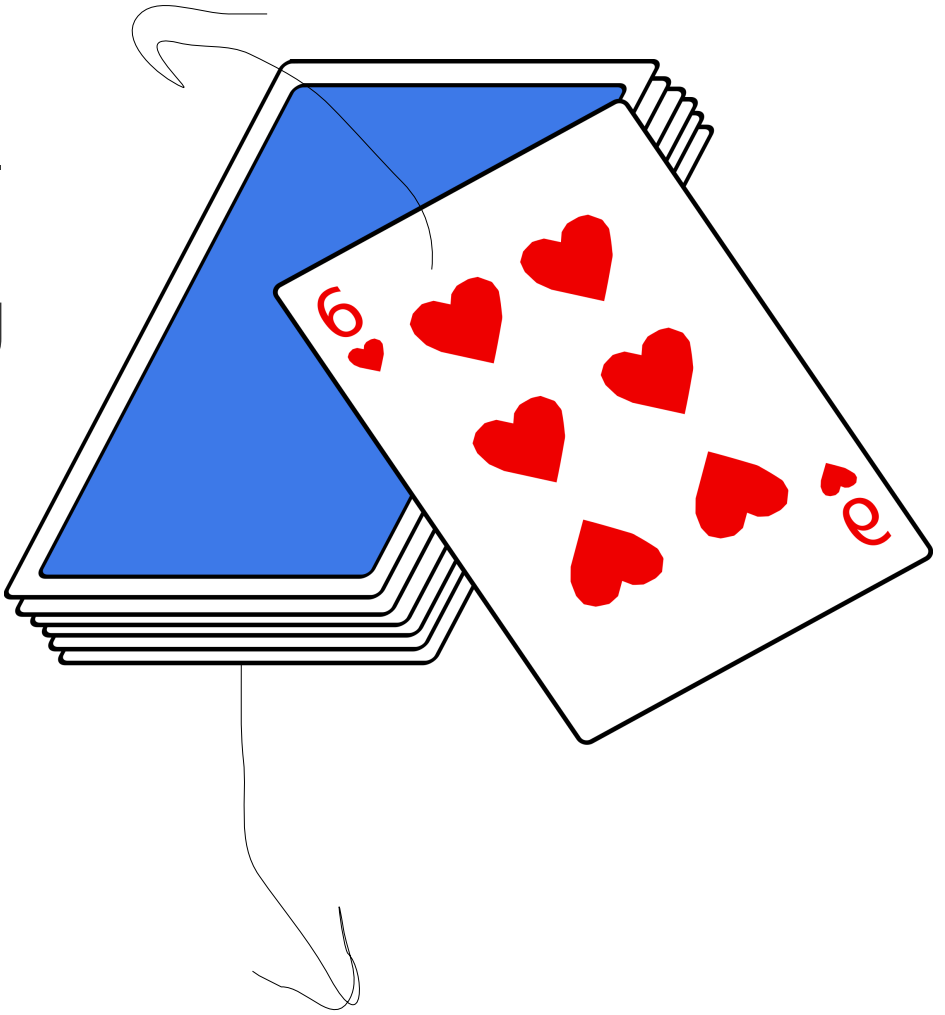
Considerations: Efficiency → Type

- are the data types to be collected uniform?
- `array.array` is *more* efficient by *restricting* element type, e.g. to character, **`int`** or **`float`**



Considerations: Efficiency → Access

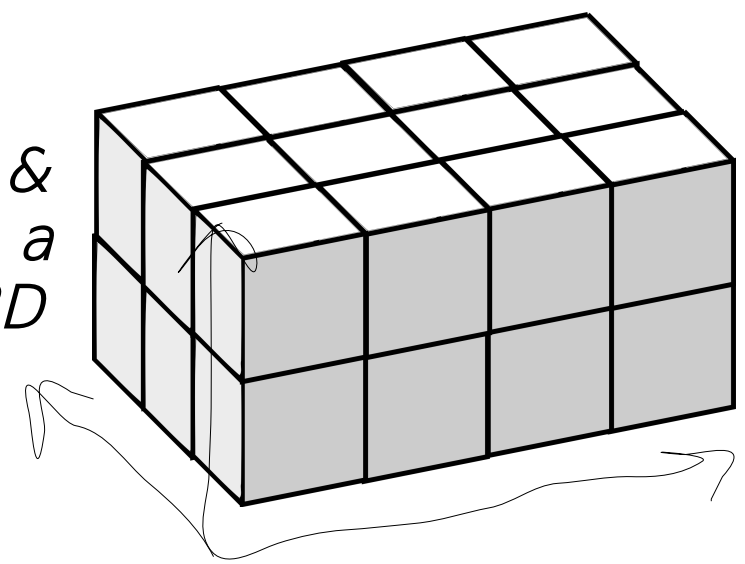
- `list` is fully flexible
- different operations e.g. copying, iterating over, element getting, setting & removing, have different $\mathcal{O}(n)$, so consider what is important contextually
- deque is *more* efficient by being *less* flexible: can only add, extend & return from either end

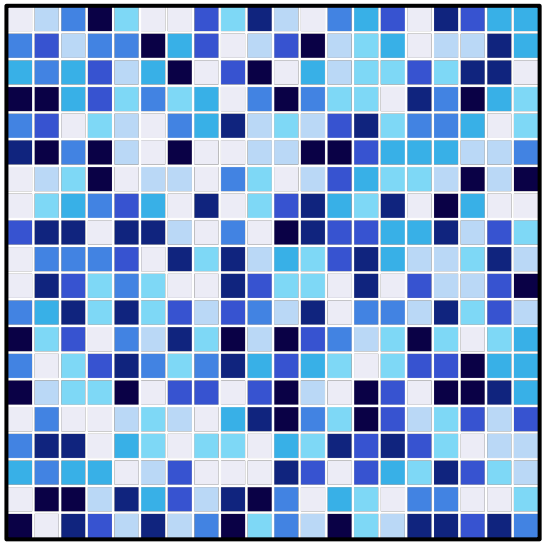


Considerations: Functionality

- external collections possess unique functionality

flexible dimension & shape, e.g. a 2 x 3 x 4 3D array



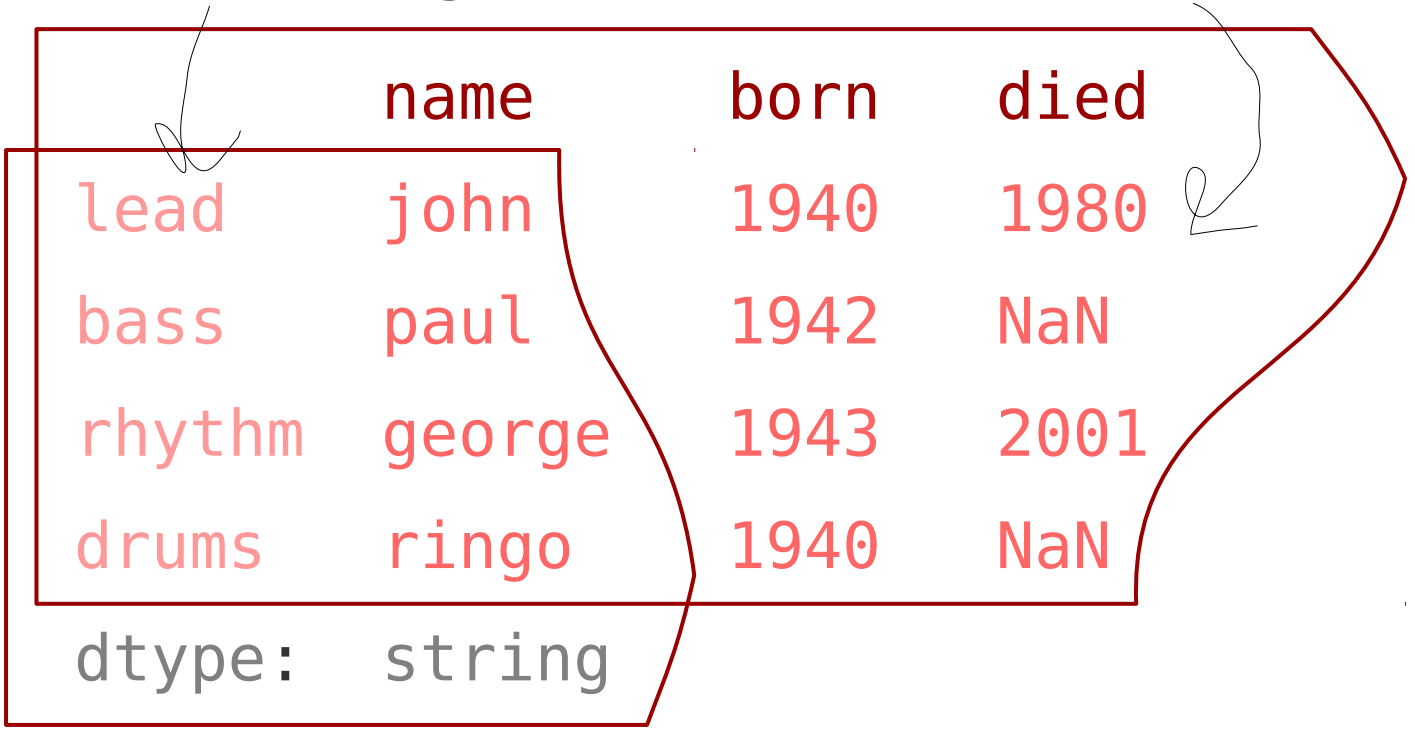


trivial to render adaptably e.g. for a 2D square array

- `numpy` library: `numpy.array` is a multi-dimensional array of identical-type data
- efficient high-level (e.g. vector) maths for `array` manipulation, enabling numerical computing
- intuitive for visualisation; `matplotlib` utilises

Considerations: Functionality

- `pandas` library extends NumPy & is based on two data objects that are tabular like statistical tables
- data analysis basis: efficient management of data sets e.g. **NaN** for missing data
- `Series` single-column; `DataFrame` multi-column



Considerations: Control

- need specific behaviour or functionality?

subclass an existing collection?

```
class myCollection<(otherCollection)>:
    """ My collection that does
    exactly what I want it to.
    """

    # create your own custom collection
```

- search the *full* Python catalogue first: don't reinvent the wheel
- `collections.UserList` & `.UserDict` may help

Loose ends

- final named collections yet to be covered:

  - `collections.defaultdict`: like **`dict`** but with a default value assigned on lookup of non-existent keys instead of giving a **`KeyError`**
  - `collections.ChainMap`: collects & allows processing of multiple **`dict`**

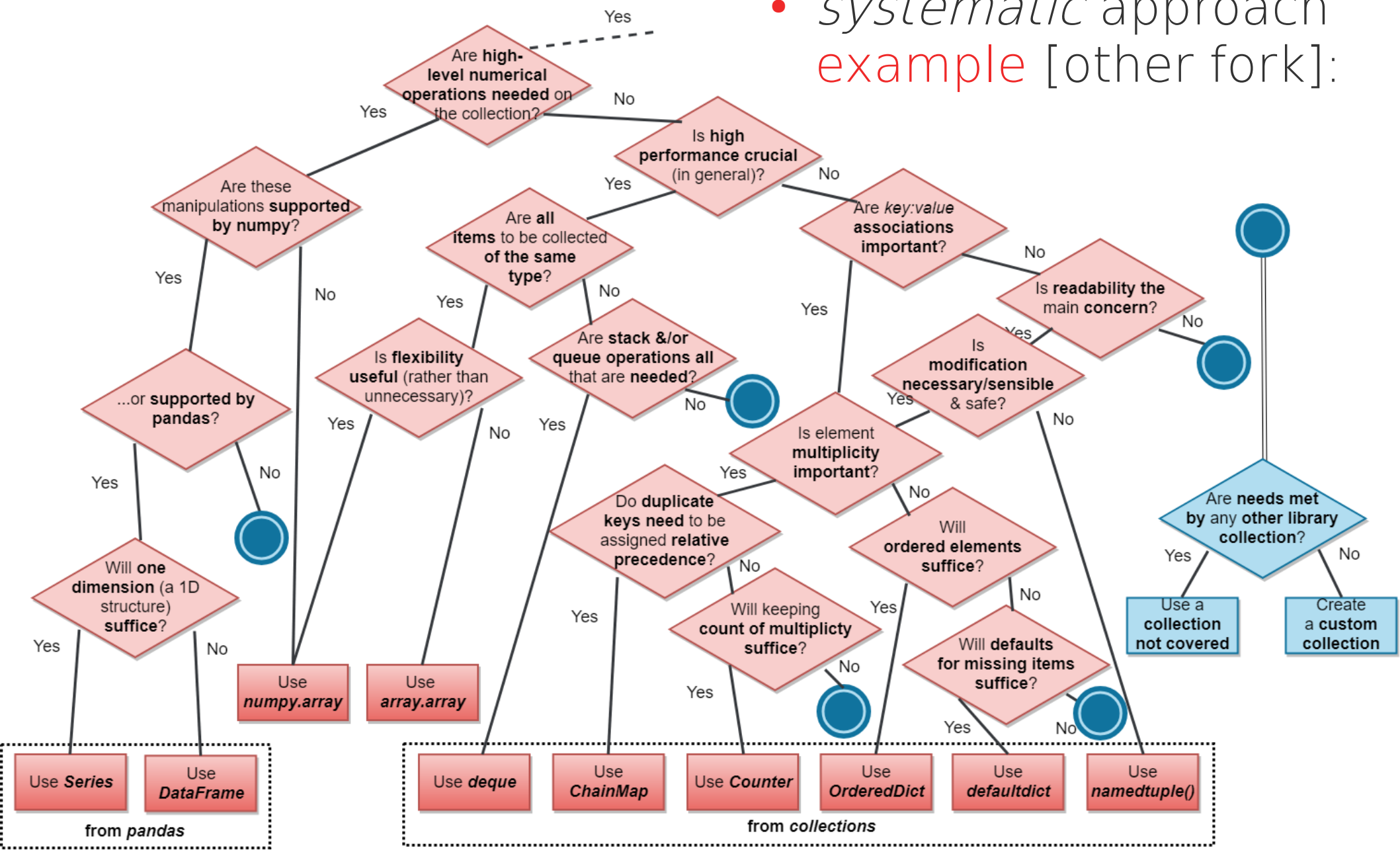
Choice by context: mnemonic

- regarding collections, we should all aim to be…

C ontrol (custom class?)
R eadability
A ssociations (keys?)
F unctionality
T ally & order
S implicity (a built-in?)
[**WO**]
M utability
E fficiency (time & space complexity)
N esting capability

Choice by context: flow chart

- *systematic* approach example [other fork]:



EOF

- any questions?

- for reference, slides located on GitHub at:
  - ↖ *sadielbartholomew/talks/python-collections.pdf*
- clip art sourced from:
  - ↖ *openclipart.org*