

# Python ↔ R Handy Mental Mapping Cheat Sheet

Goal: quick “how do I think about this in the other language?” mappings. R idioms ↔ Python (NumPy/pandas).

## Assignment + printing

R mindset / idiom	Python analogue
<code>x &lt;- 1 x # or x = 1 print(x)</code>	<code>x = 1 x # in scripts print(x)</code>

## 1-indexing vs 0-indexing

R mindset / idiom	Python analogue
<code>x &lt;- c(10,20,30) x[1] # 10 x[3] # 30</code>	<code>x = [10,20,30] x[0] # 10 x[2] # 30</code>

## Slicing (inclusive vs exclusive end)

R mindset / idiom	Python analogue
<code>x &lt;- 1:10 x[2:5] # 2,3,4,5 x[-1] # drop first</code>	<code>import numpy as np x = np.arange(1, 11) x[1:5] # 2,3,4,5 x[1:] # drop first</code>

## Vectorization is default

R mindset / idiom	Python analogue
<code>x &lt;- c(1,2,3) x^2 x + 10</code>	<code>import numpy as np x = np.array([1,2,3]) x**2 x + 10</code>

## Recycling (R) ↔ broadcasting (NumPy)

R mindset / idiom	Python analogue
<code>c(1,2,3) + c(10,20) # 11,22,13 (recycles)</code>	<code>import numpy as np np.array([1,2,3]) + np.array([10,20]) # ValueError unless shapes broadcast. # Use np.resize / explicit broadcasting.</code>

## Missing values

R mindset / idiom	Python analogue
<code>x &lt;- c(1, NA, 3) is.na(x) mean(x, na.rm=TRUE)</code>	<code>import numpy as np x = np.array([1, np.nan, 3.0]) np.isnan(x) np.nanmean(x)</code>

## Logical filtering

R mindset / idiom	Python analogue
<code>x &lt;- 1:10 x[x %% 2 == 0]</code>	<code>import numpy as np x = np.arange(1, 11) x[x % 2 == 0]</code>

## Apply family ↔ map/apply

R mindset / idiom	Python analogue
<code>lapply(list(1:3, 4:6), mean) sapply(1:5, function(i) i^2)</code>	<code>list(map(np.mean, [np.arange(1,4), np.arange(4,7)])) [i**2 for i in range(1,6)]</code>

## Pipes (%>%) ↔ method chaining

R mindset / idiom	Python analogue
<code>library(dplyr) df %&gt;% filter(x &gt; 0) %&gt;% group_by(g) %&gt;% summarise(m=mean(y))</code>	<code>df[df['x'] &gt; 0] .groupby('g', as_index=False) .agg(m=('y', 'mean'))</code>

## Data frame column access

R mindset / idiom	Python analogue
<code>df\$x # or subset(df, x &gt; 0)</code>	<code>df['x'] # boolean mask df[df['x'] &gt; 0]</code>

## Factor ↔ categorical

R mindset / idiom	Python analogue
<code>g &lt;- factor(c('A','B','A')) levels(g)</code>	<code>import pandas as pd g = pd.Categorical(['A','B','A']) g.categories</code>

## Grouping: dplyr verbs ↔ pandas

R mindset / idiom	Python analogue
<code>df %&gt;% group_by(g) %&gt;% summarise(n=n(), m=mean(y))</code>	<code>df.groupby('g').agg(n=('g','size'), m=('y','mean')).reset_index()</code>

## Reshaping: pivot\_longer/wider ↔ melt/pivot

R mindset / idiom	Python analogue
<code>library(tidyr) pivot_longer(wide, cols=c(a,b), names_to='k', values_to='v') pivot_wider(long, names_from=k, values_from=v)</code>	<code>wide.melt(id_vars=['id'], value_vars=['a','b'], var_name='k', value_name='v') long.pivot(index='id', columns='k', values='v').reset_index()</code>

## Formula modeling syntax (lm/glm) ↔ statsmodels

R mindset / idiom	Python analogue
<code>fit &lt;- lm(y ~ x1 + x2 + g, data=df) summary(fit)</code>	<code>import statsmodels.formula.api as smf fit = smf.ols('y ~ x1 + x2 + C(g)', data=df).fit() fit.summary()</code>

## Randomness: set seed

R mindset / idiom	Python analogue
<code>set.seed(42) rnorm(3)</code>	<code>import numpy as np rng = np.random.default_rng(42) rng.normal(size=3)</code>

## In-place vs copy (gotcha)

R mindset / idiom	Python analogue
<code>df2 &lt;- df # usually copy-on-modify behavior</code>	<code>df2 = df.copy() # make an explicit copy # beware chained assignment; prefer .loc</code>

## Common “where is this function?” mental map

R mindset / idiom	Python analogue
<code>mean(), sd(), cor()</code> # stats package (base) # dplyr verbs in dplyr	<code>np.mean(), np.std()</code> scipy.stats.* for tests pandas for wrangling seaborn/matplotlib for plots

