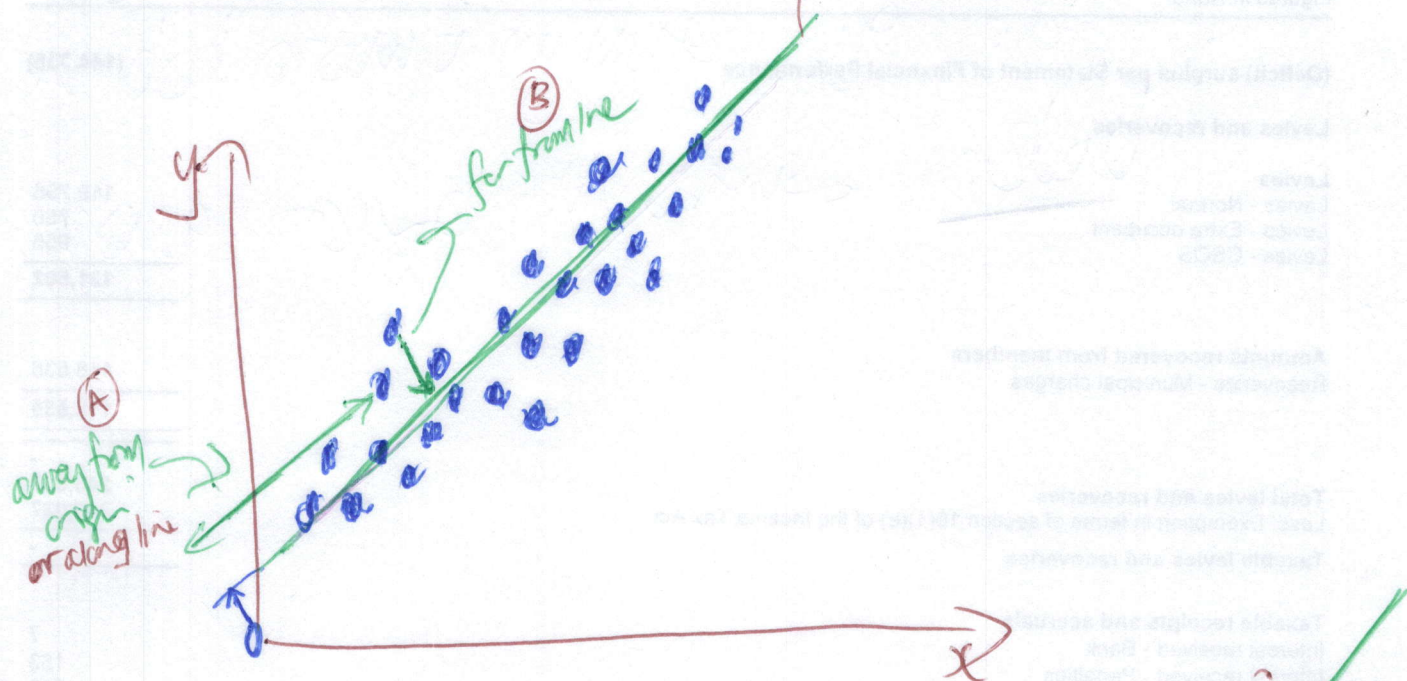


Vectors Basis - Application - Example

①

- we have 2D data points



and they may less lie on straight line ✓

- so we can describe the data by mapping it onto that line

- and say how 'far' ^(fit) they are on the line

How

2 Dimensions

① → how far along the line

② → and how from the line

and all the other points are slightly
different distances from line

②

how far from line (B) - really how
noisy the data stand is (from line) (if measure)

⇒ spread (big distance away)

⇒ narrow (small distance away)

Distance from line ⇒ NOISE

But that NOISE term tells how
good the line fit is

So 2 deductions:

⇒ along and away from line

is that they are orthogonal from
each other, so we can use

the dot product to do the projection.

③

to map the data from x, y space
onto space of features on the line and
away from line

→ If we apply this to a neural network in ML. Not
recognise face

— we want to make some transformations
of all pixels in face for new
Basis that describe nose shape,
skin hue, distance between ~~eye~~ eyes,
and then discard the actual pixel data.

Goal of Learning process of NN is to
somehow derive a set set of basis
vectors that extract the most
information rich features of face

Recap: [+ period Module]

(4)

- we talked about the dimensionality of vector space i.e. the number of independent vector space that it has
- we found a test for independence
- and what that means i.e. 'mapping' from one space to another.
- and how it's going to be useful in Data Science