

## LDA Topic Modelling

### My Output:

[(0, '0.344\*"pike" + 0.209\*"bass" + 0.140\*"horn" + 0.135\*"deep" + 0.121\*"catapult" + 0.052\*"tuba"')] – Corresponds to **Topic 2** of Beta Vector because probabilities for words deep, tuba and bass match closely with beta vectors of these words for topic 2.

(1, '0.304\*"deep" + 0.280\*"pike" + 0.249\*"bass" + 0.073\*"catapult" + 0.063\*"horn" + 0.031\*"tuba"')] – Corresponds to **Topic 1** of Beta Vector because tube, horn and catapult have probabilities close to 0 matching topic 1 closely.

(2, '0.230\*"bass" + 0.202\*"horn" + 0.171\*"deep" + 0.163\*"pike" + 0.129\*"tuba" + 0.106\*"catapult"')] – Corresponds to **Topic 3** of Beta Vector because bass and horn are both present with high probabilities matching topic 3 beta vectors for these words.

There is some noise but that is expected. This is the closest the above output matches to the beta vectors below.

### Beta Vectors:

Word Order: "bass", "pike", "deep", "tuba", "horn", "catapult"

[0.4, 0.4, 0.2, 0.0, 0.0, 0.0] – Topic 1

[0.0, 0.3, 0.1, 0.0, 0.3, 0.3] – Topic 2

[0.3, 0.0, 0.2, 0.3, 0.2, 0.0] – Topic 3