The Diversity of Makeup Shades.

Jesus Mendoza

Background

- The Pudding essay Beauty Brawl published in June 2018 to discover how inclusive were beauty brands around the world
- Collected a list of beauty brands in the United States, Nigeria, India, and Japan that were considered "best sellers" by several sources
- Visited each brand's website, recorded the hex values for each of the colored swatches shown for the product(s)
- Used Adobe Photoshop to extract the lightness value of each color using the CIE Lab color

Why?

Through research, the Pudding decided what brands/products to sample based on the bestselling foundations according to sources that included...

US bestseller lists: POPSUGAR, Amazon, StyleCaster, Refinery29, Statista, BEAUTY/crew

Articles recommending beauty products to people of color: VIBE, Byrdie, The FADER, Allure, Glamour, Fast Company, THE CUT, Bustle,

HuffPost, more.com, BuzzFeed, Refinery29

Articles recommending Nigerian beauty products: <u>BeautyInLagos</u>, <u>Beauty Geek</u>, <u>Lux Afrique</u>, <u>Zikel Cosmetics</u>, <u>Pulse.ng Pulse.ng again</u>, Information Nigeria Women, Girly Essentials, Winnie The Make-Up Artist, Jumia Travel

Column Descriptions

- brand full name of the brand(s) that creates the foundation shade
- brand_short an abbreviated version of the brand(s)' name without special characters or spaces
- product full name of the foundation product that was sampled
- product_short a 2 to 3 character string that is unique to a particular product
- hex the hexadecimal color code for a particular shade
- H the hue value from the HSB or HSV color space
- S the saturation value from the HSB or HSV color space
- V the value (sometimes known as the brightness) from the HSB/HSV
- L the lightness value from the CIE Lab color space
- group each product belongs to one group, there were 7 groups used

[1]:	import pandas														print("First ten rows")								
[2]:	df =	pandas.read	_csv('shad	les.csv')									[195]		<pre>print(df.head(10))</pre>								
[5]:	df														First ten rows								
[5]:		Unnamed: 0	brand	brand_short	product	product_short	hex	н	s	v	L	group		0	Unnamed: 0	bran Maybellin		product produ Fit Me	ct_short fmf	hex f3cfb3	26.0		
	0	0	Maybelline	mb	Fit Me	fmf	f3cfb3	26.0	0.26	0.95	86	2				Maybellin Maybellin		Fit Me Fit Me	fmf fmf	ffe3c2 ffe0cd			
	1	1	Maybelline	mb	Fit Me	fmf	ffe3c2	32.0	0.24	1.00	92	2				Maybellin Maybellin		Fit Me Fit Me	fmf fmf	ffd3be bd9584			
	2	2	Maybelline	mb	Fit Me	fmf	ffe0cd	23.0	0.20	1.00	91	2		į	5 5	Maybellin	e mb	Fit Me	fmf	eabda6	20.0		
	3	3	Maybelline	mb	Fit Me	fmf	ffd3be	19.0	0.25	1.00	88	2		6 7		Maybellin Maybellin		Fit Me Fit Me	fmf fmf	fbd2ad e2b597 e4b38e b2856f	24.0 26.0		
	4	4	Maybelline	mb	Fit Me	fmf	bd9584	18.0	0.30	0.74	65	2				Maybelline Maybelline		Fit Me Fit Me					
			***	***	•••	***	•••								9 9	Maybettin	e iiib	rit me	11111	020301	20.0		
	620	620	L'Oréal		True Match	tms	eecfba					7											
	621	621	L'Oréal		True Match	tms						7	D \	~	print("La	print("Last ten rows")							
	622	622	L'Oréal		True Match	tms						7			print(df.tail(10))								
	623	623	L'Oréal		True Match		e9c4b1						[197]	7]								_	
	624	624	L'Oréal		True Match	tms						7			Last ten row Unnamed		nd brand shor	t product	product	short	hex	н	\
	024	024	LOTeal	10	True Materi	uns	eabear	24.0	0.51	0.92	80	,			615	615 L'Oré	al l	o True Match		tms	ddb496	25.0	`
	625 ro	ws × 11 colum	ns										_			616 L'Oré 617 L'Oré		o True Match o True Match		tms tms	eebfa3 e7cbb5	22.0 26.0	
D ~		(a. 1876)						_							618	618 L'Oré	al l	o True Match		tms	dcb29a	22.0	
	<pre>print("Random ten rows") print(df.sample(n=10))</pre>														619 L'Oré 620 L'Oré		o True Match o True Match		tms tms	f0c7b3 eecfba			
[199]	рі	TITC (all I Sample	.c(II=10))													621 L'Oré		o True Match		tms		19.0	
																622 L'Oré		o True Match		tms		20.0	
	Random ten rows Unnamed: 0 brand brand_short product_product_short \															623 L'Oré		o True Match		tms		20.0	/
	134	134		Oréal		falliable	ipm	`							624	624 L'Oré	at t	o True Match		tms	eabea1	24.0	
	8	8		elline	mb	Fit Me	fmf																
	174	174		Fenty		RO FILT'R	pf																
	560	560		NARS		vet Matte	vm																
	407 41	407 I 41	Make Up For bareMin		mu bm	Ultra HD barePRO	uhd																
	167	167		Fenty		RO FILT'R	pro pf																
	607	607		elline		Me Matte	fmm																
	400	400	Bobbi			Long-Wear	slw																
	518	518	Addi	ction	ad The F	oundation	tf																

Observations

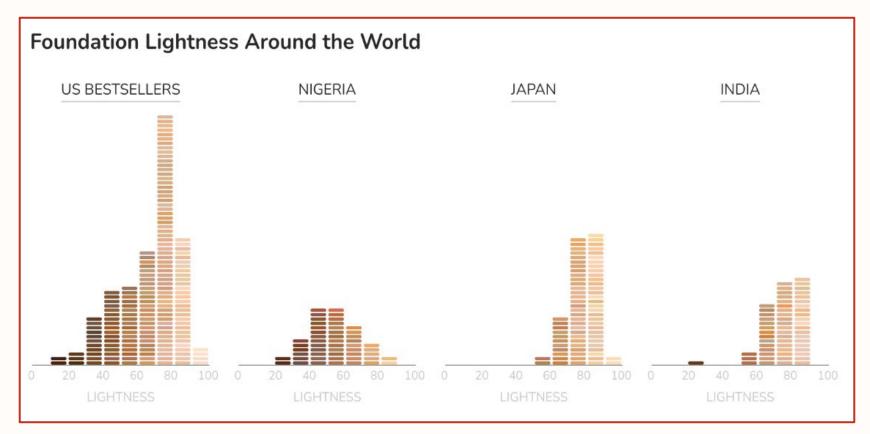
- It contains 625 rows and 11 columns.
- In the dataset, the mean (average) is 3.472 and the median is 3.0
- For the column **brand**, the most occurring name is Maybelline which appears 54 times.
- For the column **product**, the most occurring product is the Fit Me Matte foundation which appears 54 times.





Findings

- The dataset contains 625 products, 7 groups
- Brands include:
 - Maybelline (brand) → mb (brand_short)
 - Laws of Nature (brand) → In (brand_short)
 - NARS (brand) → na (brand_short)
 - Shiseido (brand) → sh (brand_short)
- Products include:
 - Fit Me (brand) → fmf (brand_short)
 - Foxy Finish (brand) → ff (brand_short)
 - Velvet Matte (brand) → vm (brand_short)
 - Synchro Skin (brand) → ss (brand_short)



Visual Representation of the Dataset Findings

Conclusion

- Overall, the U.S. has the largest number of shades despite being more catered to lighter shades when it comes to foundation.
- Nigeria focuses more on darker foundation shades.
- Both India and Japan cater more to the lighter shades either having none or few darker foundation shades.
- The dataset helps identify what beauty brands around the world are inclusive or just claim to be inclusive.

Sources

- https://pudding.cool/2018/06/makeup-shades/
- https://github.com/the-pudding/data/blob/master/makeup-shades/shades.csv
 des.csv

