

DermaGlow Al

Dermatology Essentials and Product Ingredient Analyzer Bot

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Purpose of the Project:

The Dermatology Essentials and Product Ingredient Analyzer Bot helps individuals to make smart choices about skincare essentials. It provide users with a convenient tool to analyze skin concerns and recommend suitable skincare products based on their skin types, specific needs, and ingredient preferences to promote healthier and better skincare habits.

Domain

The project is in the dermatology and skincare domain, focusing on analyzing skin concerns and recommending suitable skincare products based on user input. By analyzing these concerns, this bot recommends the right products that would work best for each person, making skincare more personal and effective.

Reasoning Method

Natural Language Processing (NLP) for Intent Recognition

This chatbot understands the user's intent and extracts key features (product type, preferences). This is a combination of rule-based and machine learning methods for example, intent classification with Naive Bayes, NN and BERT).

Example:

 The chatbot detects that the user mentioned "want" and "toner," so it classifies this intent as a request for a product recommendation.

Implementation:

Combined these with a machine learning classifier trained on labeled intents.

Data Preparation

product id	product name	product, type	main ingredients	ingredient benefits	suitability	directions for use skin concerns	skin concerns addressed allergens ingredients to avoid	prices	alcohol free	cruelty free	fragrance free
- 1	delo SunExpert Whitening Sunscreen SPFSD 50mL	Moisturing	Water, Ethylhexyl Methoxycinnamate, Cyclotetrasiloxane, Cyclopentasiloxane, Niacin	F Hydratus skin, sooth	All skin types	yday after clearning Oily, Dry	Fade dark spots, even out a Ethylhexyl Methoxycinnamate (C	359.00	no	yes.	yes
2	Human Nature Hyaluronic Acid Gel Day Moisturize	Mosturiver	Water, Propanediol, Betaine, Glycorin, Stearyl Alcohol, Aloe Barbadensis Leaf Juce, Glycine S	o 99.82% Natural We	All skie types	after cleansing and t Sensitive	Anti-aging, brightening, sca. Stearyl Alcohol, Cetearyl Alcohol,	250.00	60	yes	yes
3	Ellana Mineral Cosmetics Even Better Skin Care Tir	Mosturiore	Water, Cyclopentasilosane, Titanium Dioxide, Peg-10 Dimethicone, Triethoxycaprylykilane, Si	li Ellana Skin Care Tint	All skie types	an, moisturized skin Aging, Wrinkles/Fine Lines	Hydrating, Acre Fighting, Br Silicones (e.g., Cyclopentasilosan	225.00	.00	yes	yes
- 4	Barentifes BNB Mari Mori Makerup Remover Wipe	Cleamer	Water, Sodium Lauroyl Sarcosinate, Methylpropanediol, Decyl Glucoside, Phenoxyethanol, Pf	(Mangold-Moringa er	All skin types	ities. Ensure all area Sensitive	Souther the skin. Amino Acid Surfactants: These go	199.00	yers	90%	no
5	Garnier Pink Micellar Water Pink (125ML/400ML)	Clearser	AQUA / WATER, HEXYLENE, GLYCOL, GLYCERIN, DISODIUM COCDAMPHODIACETATE, DISODIU	Y Cleames like toner, g	All skin types	and lips to remove in Combination	Sensitive, Dry, Normal, Con Heaylene Glycol: Functions as a s-	199,00	yes	yes.	901
6	Luse Organiz Aloe Vera Southing Gel 100ml Sachet	Moisturizer	Aloe Barbademis Leaf Juice, Water, Aloe Barbademis Leaf Juice Powder, Efectin, Carboner,	Arginine, Phenoxyeth.	All skie types	sep out of reach of cOily, Dry	Sensitive, Dry, Normal, Com Aloe Barbadensis Loaf Juice and I	109.00	yes	yes	ne
7	Fond's Bright Beauty Serum Day Cream SPF30	Mosturiore	Aqua, Octoorylene, Ethylhoxyl Salicylate, Stearic Acid, Niacinamide, Glyceryl Stearate, Butyl N	h Pond's Bright Beauty	All skie types	Rer 7 continuous da Sensitive	Sensitive, Dry, Normal, Com Niacinamide (Vitamin B3): Knows	160.00	yes	yes	np
В	b.fresh, Hydrating Lip Serum, Cherry Bomb, 0.5 ft c	Serum	Caprylic/capric triglyceride, glycerin, water (aqua, eau), quillaja saponaria (scapbark) wood e	s It's Lip Care Done Di	All skin types	, Reapply if necessar Dry	Sensitive, Dry, Normal, Com Jojoba Seed Oil (Simmondsia Che	250.00	995	y03	no
. 5	PAPA FEEL 3X Retinol Cream 30g anti-wrinkles a	Moisturizer	3X Retinol, soluble collegen (collegen), tocopherol (vitamin E), and ceremide NP	Moisturizing texture	All skin types	ikles or melasma. Ur Agerg, Dry, Semilive	Sensitive, Dry, Normal, Con Retinol and Retinol Derivatives: T	300,00	yes	yes.	961
10	Green Herb Hydrating Gel 50ml	Mosturiver	Guava Inaf, Hyaluronic acid, Centella Asiatica, Alos Vera, Distilled Water, Menthol	Hydrates dry wound	All skie types	nassage until fully at Dry	Sensitive, Dry, Normal, Con: Menthol: Can cause initation or a	400.00	yes	yes	ne
11	SNDE MORE AWESOME PORESOME PORE PERFEC	Serum	Aqua (Water), Glycerin, Hamamelis Virginiana (Witch Hazel) Extract, Carica Papaya (Green Pa	p. pore-minimizing, re-	All skie types	s with visible pores a Dullness/timeven Texture	Sensitive, Dry, Normal, Corr Witch Hazet: May cause initation	359,00	60	yes	no
12	Garnier Vitamir C Day Cream with SPF36 SOML	Mosturier	AQUA/WATER, GLYCERIN, ALCOHOL, STEARIC ACID, OCTOCRYLENE, PALMITIC ACID, BUTYL MI	lighten and protec	All skin types	torning. Avoid the e Sensitive	Sensitive, Dry, Normal, Corr Fragrance (Parlum): Can cause all	399.00	60	90%	no
13	Luse Organia Power Glow Vita Glow C Whip Clean	Cleanser	Water, Glycerin, Palmitic Acid, Stearic Acid, Myristic Acid, Potassium Hydroxide, Lauric Acid, S	cleanses face with	All skin types	ly massage onto a da Agerg, Dry, Sensitive	Sensitive, Dry, Normal, Con Cocamidopropyl Betaine: Can car	5	467	yes.	901
14	GLAD2GLOW Yuja Symwhite 377 Dark Spot Serum	Serum	AQUA, BUTYLENE GLYCOL, GLYCERIN, CITRUS JUNOS FRUIT EXTRACT, 3-0-ETHYL ASCORBIC A	Clades dark spots & a	All skie types	L Cently pat until ab Dullness/Uneven Texture	Semilive, Dry, Normal, Com Citrus Junis Fruit Extract (Yuja): 1	277.00	80	yes	no
15	Mary Elizabeth R: Rosehip & Fruits Serum Rosehip	Serum	Water, Rosehip Oil, Fruit Acid Estract (Pineapple, Orange, Lemon, Strawberry, Tomato, Calam	Acne Care, Sensitive,	All skie types	patch text to determ Combination	Sensitive, Dry, Normal, Corr Fruit Acid Extract (Pineapple; Ora	275.00	yes	yes	yes
16	Soul Apotherary: Snail Repair Serum (with Centell	Serum	Snail Secretion Filtrate, Nacinamide, Centella Asiatica Extract, Allantoin, Papain, Tocopherol (V Snad Secretion Filtre	All skin types	massage until fully al Aging, Dry, Sensitive	Sensitive, Dry, Normal, Com Snail Secretion Filtrate: While sna	300.00	yes	90%	yes
-17	Human Nature Vitamin C + Hya Calamansi Radiano	Serum	aqua (water)*, propanediol*, glycerin*, ascorbyl glucovide*, sinc lactate*, Citrus microcarpa	chelps being out natu	All skin types	issage until absorber Combination	Sensitive, Dry, Normal, Con Citrus Microcarpa (Calamansi) Fr	495.00	985	yes.	no
18	Galactomyces Serum Magic Liquid Water Essence	Serum	Galactomyces Ferment Filtrate, Butylene Glycol, Niacinamide, Panthenol, Hydrolyzed Hyeluro	r Anti Aging, Skin Now	All skie types	pat until absorbed. Acre/Blemishes, Dark spots/ Hyperp	i Semitive, Dry, Normal, Con Galactomyces Ferment Filtrate: F	350.00	yes	yes	yes
19	Organic Skin Japan Antiacrer Whitening Collection	Moisturiore	Salicylic acid, Tea Tree Estract, Nacinamide	Helps reduce allmin	All skie types	one areas. Gently ma Acne/Blemishes, Dark spots/ Hyper;	p Sensitive, Dry, Normal, Corr Salicylic Acid: As an exfoliating ag	179.00	yes	yes	yes
20	GLADZGLOW Milk Amino Acids Gentle Cleanser	Cleamer	AQUA, POTASSIUM PALM KERNELATE, POTASSIUM COCOATE, POTASSIUM COCOY	L deeply cleans the sk	All skin types	sage in circular moti Dry, Oily	Sensitive, Dry, Normal, Corr Milk Protein Extract: People with	175.00	yes	90%	no

We collected approved skin care products and other related data from **FDA** website with entries that are timely, significant and complete. To keep things relevant, we selected key fields like product ID, name, type, main ingredients, their benefits, suitability, directions for use, skin concerns, allergens to avoid, pricing, and whether the products are alcohol-free, cruelty-free, or fragrance-free. This way, our dataset stays relevant and meaningful for the analysis.

FDA Website: https://verification.fda.gov.ph/COSMETIC ALL INDUSTRYlist.php

Data Preparation

product id	product name	product, type	main_ingredients	ingredient benefits	suitability	directions for use skin_concerns	skin_concurrs_addressed_aflergens_ingredients_to_evoid_	prices	alcohol free	cruelty free	fragrance free
1	delo SunExpert Whitening Sunscreen SPFSD 50mL	Moisturion	Water, Ethylhexyl Methoxycinnamate, Cyclotetrasiloxane, Cyclopentasiloxane, Niacina	Hydrates skin, sooth	All skin types	yday after clearning Oily, Dry	Fade dark spots, even out a Ethylhexyl Methoxycinnamate (C)	359.00	no	yes.	961
2	Human Nature Hyaluronic Acid Gel Day Moisturize	Moisturiver	Water, Propanediol, Belaine, Glycorin, Steanyl Alcohol, Aloe Barbadensis Leaf Juce, Glycine So	99.82% Natural We	All skie types	after cleaning and I Sensitive	Anti-aging, brightening, sea Stearyl Alcohol, Cetearyl Alcohol,	250.00	60	yes	995
3	Ellana Mineral Cosmetics Even Better Skin Care Tir	Mosturier	Water, Cyclopentasilosane, Titanium Dioxide, Peg-10 Dimethicone, Triethoxycaprylykilane, Sili	Ellana Skin Care Tint	All skie types	an, motisturized skin. Aging, Wrinkles/Fine Lines.	Hydrating, Acre Fighting, Br Silicones (e.g., Cyclopentasilosan	225.00	.00	yes	yes
- 4	Barentifes BNB Mari Mori Makerup Remover Wipe	Clearner	Water, Sodium Lauroyl Sarcosinate, Methylpropanediol, Decyl Glucoside, Phenoxyethanol, PEC	Mangold-Moringa in	All skin types	ities. Ensure all area Sensitive	Souther the skin. Amino Acid Surfactants: These go	199.00	yes	90%	no
5	Garnier Pink Micellar Water Pink (125ML/400ML)	Clearser	AQUA / WATER, HEXYLENE, GLYCOL, GLYCERIN, DISODIUM COCCIAMPHODIACETATE, DISODIUM	Cleames like toner, g	All skin types	and lips to remove in Combination	Sensitive, Dry, Normal, Con Heaylene Glycol: Functions as a s-	199,00	yes	yes.	361
6	Luse Organiz Aloe Vera Southing Gel 100ml Sachet	Mosturizer	Aloe Barbadensis Leaf Juice, Water, Aloe Barbadensis Leaf Juice Powder, Glycerin, Carbonier, A	Arginine, Phenoxyeth.	All skie types	eep out of reach of cOily, Dry	Sensitive, Drg, Normal, Con Aloe Barbadensis Loaf Juice and I	109.00	yes	yes	no
7	Fond's Bright Beauty Serum Day Cream SPF-80	Mosturier	Aqua, Octoorylene, Ethylhoxyl Salicylate, Stearic Acid, Niacinamide, Glyceryl Stearate, Butyl Mi	Pond's Bright Beauty	All skie types	fter 7 continuous de Sensitive	Sensitive, Dry, Normal, Com Niacinamide (Vitamin B3): Knows	160.00	yirk	yes	np
В	b.fresh, Hydrating Lip Serum, Cherry Bomb, 0.5 ft c	Serum	Caprylic/capric triglyceride, glycerin, water (aqua, eau), quillaja saponaria (scapbark) wood ex	It's Lip Care Done Di	All skin types	, Reapply if necessar Dry	Sensitive, Dry, Normal, Com Jojoba Seed Oil (Simmondsia Chic	250.00	yes	90%	no
. 5	PAPA FEEL 3X Retinol Cream 30g anti-wrinkles a	Moisturion	3X Retinol, soluble collegen (collegen), tocopherol (vitamin E), and ceremide NP	Moisturizing texture	All skin types	ikles or melasma. Ur Agerg, Dry, Sensitive	Sensitive, Dry, Normal, Com Retinol and Retinol Derivatives: T	300,00	yes	yes.	361
10	Green Herb Hydrating Gel 50ml	Mosturizer	Guava leaf, Hyaluronic acid, Centella Asiatica, Aloe Vera, Distilled Water, Menthol	Hydrates dry wound	All skie types	nassage until fully at Dvy	Sensitive, Drg, Normal, Con: Menthol: Can cause initation or a	400.00	yes	yes	00
11	SNDE MORE AWESOME PORESOME PORE PERFEC	Serum	Aqua (Water), Glycerin, Hamamelis Virginiana (Witch Hazel) Extract, Carica Papaya (Green Pap	pore-minimizing, or	All skie types	s with visible pores a Dullness/Unieven Texture	Sensitive, Dry, Normal, Con Witch Hazel: May cause initation	359,00	.00	yes	np
12	Garnier Vitamir C Day Cream with SPF36 SOML	Mosturieer	AQUA/WATER, GLYCERIN, ALCOHOL, STEARIC ACID, OCTOCRYLENE, PALMITIC ACID, BUTYL ME	lighten and protec	All skin types	corning. Avoid the e Sensitive	Sensitive, Dry, Normal, Corr Fragrance (Parfum): Can cause all	399.00	en en	90%	no
13	Luse Organis Power Glow Vita Glow C Whip Clean	Clearset	Water, Glycerin, Palmitic Acid, Stearic Acid, Myrtelic Acid, Potassium Hydroxide, Lauric Acid, Sc	cleanses face with	All skin types	ly massage onto a di Agerg, Dry, Sensitive	Sensitive, Dry, Normal, Con Cocamidopropyl Betaine: Can car	5	yes	yes.	901
14	GLAD2GLOW Yuja Symwhite 377 Dark Spot Serum	Serum	AQUA, BUTYLENE GLYCOL, GLYCERIN, CITRUS JUNOS FRUIT EXTRACT, 3-D-ETHYL ASCORBIC AC	fades dark spots & a	All skie types	. Gently pat until alz Dullness/Uneven Testure	Semilive, Drg. Normal, Com Citrus Junes Fruit Extract (Yuja): 1	277.00	60	yes	00
15	Many Elizabeth R: Rosehip & Fruits Serum Rosehip	Serum	Water, Rosehip Cill, Fruit Acid Estract (Pineapple, Orange, Lemon, Strawberry, Tomato, Calama	Acne Care, Semilitive,	All skie types	patch text to determ Combination	Sensitive, Dry, Normal, Corr Fruit Acid Extract (Pineapple, Ora	275.00	yirk	yes	yes
16	Soul Apotherary: Snail Repair Serum (with Centell	Serum	Snail Secretion Filtrate, Nacinamide, Centella Asiatica Extract, Allantoin, Papain, Tocopherol (V	Snaë Secretion Filtre	All skin types	massage until fully af Aging, Ory, Sensitive	Sensitive, Dry, Normal, Com Snail Secretion Filtrate: While sna	300.00	yes	90%	yes
17	Human Nature Vitamin C + Hya Calamansi Radiano	Serum.	aqua (water)*, propanediol*, glycerin*, ascorbyl glucoside*, zinc lactate*, Citrus microcarpa (c	helps bring out natu	All skin types	issage until absorber Combination	Sensitive, Dry, Normal, Con Citrus Microcarpa (Calamansi) Fr	495.00	yes	yes.	no
18	Galactomyces Serum Magic Liquid Water Essence	Serum	Galactomyces Ferment Filtrate, Butylene Glycol, Niacinamide, Partherol, Hydrolysed Hyeluror	Anti Aging, Skin Now	All skie types	pat until absorbed. Acree/Blemishes, Dark spots/ Hy	perpi Servitive, Dry, Normal, Corr Galactomyces Ferment Filtrate: F	350.00	yes	yes	yes
19	Organic Skin Japan Antiacrer Whitening Collection	Moisturiore	Salicylic acid, Tea Tree Estract, Nacinamide	Helps reduce cilinary	All skie types	one areas. Gently ma Acne/Blemishes, Dark spots/H	yperp Sensitive, Dry, Normal, Corr Salicylic Acid: As an exfoliating ag	179.00	yes	yes	yes
20	GLAD2GLOW Milk Amino Acids Gentle Cleanses	Clearner	AQUA, POTASSIUM PALM KERNELATE, POTASSIUM COCOATE, POTASSIUM COCOYL	deoply cleans the sk	All skin types	sage in circular moti Dry, Oily	Sensitive, Dry, Normal, Con Milk Protein Extract: People with	175.00	yes	90%	no

Columns in the Dataset:

- product_id: Unique ID for each product.
- product_name: Name of the skincare product.
- **product_type:** Category (e.g., Moisturizer, Cleanser).
- main_ingredients: Key ingredients in the product.
- ingredient_benefits: Benefits of the ingredients.

- suitability: Suitable skin types (e.g., Oily, Dry).
- directions_for_use: Instructions on how to use the product.
- **skin_concerns:** Common skin issues the product addresses.
- allergens_ingredients_to_avoid: Allergens or irritants to avoid.
- prices: Product cost.
- Attributes: alcohol_free, cruelty_free, fragrance_free.

Data Preparation

```
    Data Collection and Handling

    data = pd.read_csv("/content/derma_essentials-fda.csv")
     data.fillna({
         'product_name': 'Unknown',
         'suitability': 'Unknown',
         'skin_concerns': 'None',
         'main_ingredients': 'None',
         'allergens_ingredients_to_avoid': 'None',
         'prices': 0,
         'cruelty_free': False,
         'fragrance free': False
     }, inplace=True)
     data = data.drop_duplicates(subset='product_name', keep='first')
     data = data.drop(columns=['Unnamed: 14'])
     print(data.isnull().sum())
```

In this step, we imported the derma_essentials-fda.csv dataset, which contains the necessary data for our analysis.



Training the Model

Display training code and explain key parameters (e.g., epochs, learning rate).

Show part of the training output and discuss performance.

Machine Learning Module and Classification Prediction using BERT, Naive Bayes, and Neural Network with TensorFlow

Tokenization Process (BERT)

```
tokenizer = BertTokenizer.from_pretrained('bert-base-uncased')
bert_model = BertForSequenceClassification.from_pretrained('bert-base-uncased', num_labels=len(label_mapping))

encoded_labels = [label_mapping[label] for label in labels]

inputs = tokenizer(intents, padding=True, truncation=True, return_tensors="pt")

train_intents, eval_intents, train_labels, eval_labels = train_test_split(
    intents, encoded_labels, test_size=0.2, random_state=42
)

train_inputs = tokenizer(train_intents, padding=True, truncation=True, return_tensors="pt")
eval_inputs = tokenizer(eval_intents, padding=True, truncation=True, return_tensors="pt")
train_labels_tensor = torch.tensor(train_labels)
eval_labels_tensor = torch.tensor(eval_labels)
```

Bert Setup for NLP

We load a pre-trained **BERT model** and **tokenizer** to classify dermatology-related texts, convert dermatology labels like "acne" or "dry" into numbers, tokenize the texts for BERT while making sure they're padded and truncated, split the data into training and test sets, tokenize both sets separately and convert the labels into tensors for model input.

Naive Bayes

```
[ ] vectorizer = CountVectorizer()
    X = vectorizer.fit_transform(intents)
    intent_model = MultinomialNB()
    intent_model.fit(X, labels)

label_encoder = LabelEncoder()
    encoded_labels = label_encoder.fit_transform(labels)

X_train, X_test, y_train, y_test = train_test_split(X, encoded_labels, test_size=0.2, random_state=42)
```

Naive Bayes Setup

We convert the text data into numerical features with `CountVectorizer`, train the Naive Bayes model using labels like "acne" or "pores," encode the labels numerically and split the data into training and test sets for performance evaluation.

Neural Networks using Tensorflow

```
nn_model = models.Sequential()
nn_model.add(layers.Dense(64, activation='relu', input_dim=X_train.shape[1]))
nn_model.add(layers.Dense(32, activation='relu'))
nn_model.add(layers.Dense(len(set(labels)), activation='softmax'))
nn_model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])
X_train_dense = X_train.toarray()
X_test_dense = X_test.toarray()
history = nn_model.fit(X_train_dense, y_train, epochs=50, validation_data=(X_test_dense, y_test))
results = nn_model.evaluate(X_test_dense, y_test)
print("Test Loss:", results[0])
print("Test Accuracy:", results[1])
y pred = nn model.predict(X test dense)
y_pred_classes = tf.argmax(y_pred, axis=1).numpy()
precision = precision_score(y_test, y_pred_classes, average='weighted')
recall = recall_score(y_test, y_pred_classes, average='weighted')
print("Test Precision:", precision)
print("Test Recall:", recall)
results = nn_model.evaluate(X_test_dense, y_test)
```

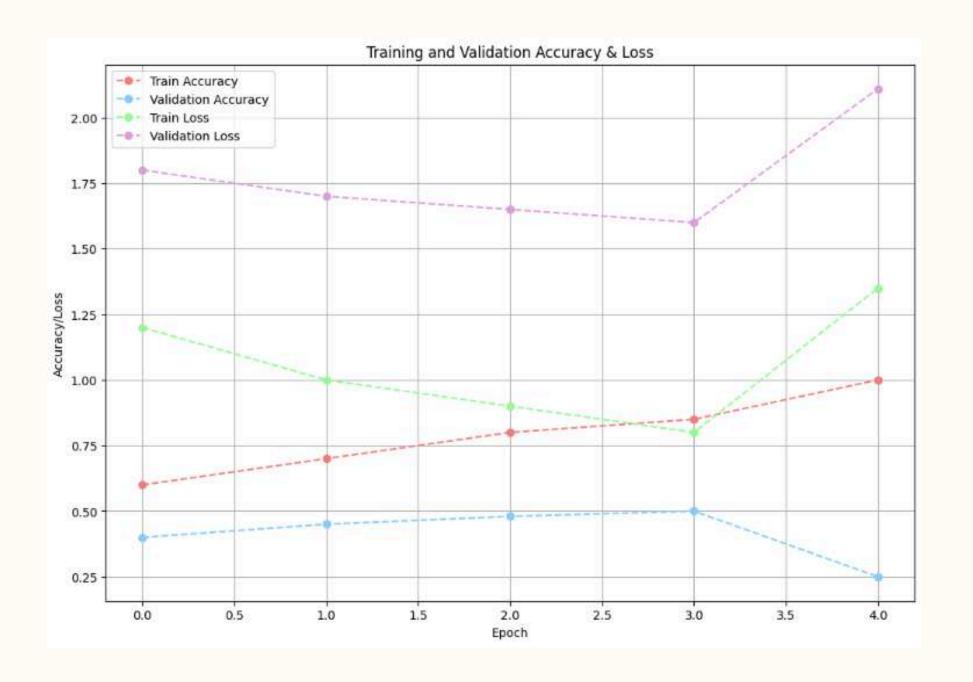
Neural Network Setup with TensorFlow

The code sets up a neural network with two hidden layers and uses softmax for multi-class classification. After training, it checks the model's performance, makes predictions, and calculates precision and recall.

Testing and Evaluation

- Show the chatbot in action with test cases.
- Discuss performance metrics like accuracy, precision, and recall

Back to Overview



Results

Evaluation Results

• The training process over 50 epochs showed improvements in training performance, with accuracy rising from 86.67% to 93.33% and loss decreasing from 0.91 to 0.78. By the end of the 50th epoch, this increased to 93.33%, showing that the model became better at recognizing patterns and making correct predictions during training.

Evaluation Results

50 Epochs Training Cycles

```
Epoch 20/50
                        0s 103ms/step - accuracy: 0.8667 - loss: 1.4723 - val_accuracy: 0.0000e+00 - val_loss: 1.9661
Epoch 21/50
                        0s 112ms/step - accuracy: 0.8667 - loss: 1.4458 - val accuracy: 0.0000e+00 - val_loss: 1.9605
Epoch 22/50
1/1 -
                        0s 122ms/step - accuracy: 0.8667 - loss: 1.4198 - val accuracy: 0.0000e+00 - val loss: 1.9548
Epoch 23/50
                        0s 284ms/step - accuracy: 0.8667 - loss: 1.3937 - val_accuracy: 0.0000e+00 - val_loss: 1.9491
Epoch 24/50
                         0s 85ms/step - accuracy: 0.8667 - loss: 1.3675 - val_accuracy: 0.0000e+00 - val_loss: 1.9431
Epoch 25/50
1/1 -
                         0s 149ms/step - accuracy: 0.8667 - loss: 1.3415 - val_accuracy: 0.0000e+00 - val_loss: 1.9365
Epoch 26/50
1/1 -
                        0s 120ms/step - accuracy: 0.8667 - loss: 1.3151 - val_accuracy: 0.2500 - val_loss: 1.9301
Epoch 27/50
                        0s 95ms/step - accuracy: 0.9333 - loss: 1.2887 - val_accuracy: 0.2500 - val_loss: 1.9239
Epoch 28/50
1/1 -
                        Os 96ms/step - accuracy: 0.9333 - loss: 1.2625 - val_accuracy: 0.2500 - val_loss: 1.9179
Epoch 29/50
1/1 -
                         0s 167ms/step - accuracy: 0.9333 - loss: 1.2365 - val_accuracy: 0.5000 - val_loss: 1.9125
Epoch 30/50
                        0s 201ms/step - accuracy: 0.9333 - loss: 1.2105 - val_accuracy: 0.5000 - val_loss: 1.9079
```

```
1/1 ---
                         0s 147ms/step - accuracy: 0.9333 - loss: 0.9564 - val_accuracy: 0.5000 - val_loss: 1.8722
Epoch 41/50
1/1 -
                         Os 109ms/step - accuracy: 0.9333 - loss: 0.9319 - val_accuracy: 0.5000 - val_loss: 1.8692
Epoch 42/50
1/1 ---
                         0s 150ms/step - accuracy: 0.9333 - loss: 0.9077 - val_accuracy: 0.5000 - val_loss: 1.8664
Epoch 43/50
1/1 -
                         0s 315ms/step - accuracy: 0.9333 - loss: 0.8840 - val accuracy: 0.5000 - val loss: 1.8638
Epoch 44/50
                         0s 273ms/step - accuracy: 0.9333 - loss: 0.8606 - val_accuracy: 0.5000 - val_loss: 1.8618
1/1 -
Epoch 45/50
1/1 -
                         0s 133ms/step - accuracy: 0.9333 - loss: 0.8378 - val_accuracy: 0.5000 - val_loss: 1.8598
Epoch 46/50
1/1 -
                         0s 182ms/step - accuracy: 0.9333 - loss: 0.8154 - val_accuracy: 0.5000 - val_loss: 1.8581
Epoch 47/50
                         0s 154ms/step - accuracy: 0.9333 - loss: 0.7937 - val_accuracy: 0.5000 - val_loss: 1.8563
Epoch 48/50
1/1 --
                         0s 131ms/step - accuracy: 0.9333 - loss: 0.7724 - val_accuracy: 0.5000 - val_loss: 1.8546
Epoch 49/50
1/1 —
                         0s 126ms/step - accuracy: 0.9333 - loss: 0.7516 - val accuracy: 0.5000 - val loss: 1.8529
Epoch 50/50
                         0s 196ms/step - accuracy: 0.9333 - loss: 0.7313 - val_accuracy: 0.5000 - val_loss: 1.8511
                         0s 37ms/step - accuracy: 0.5000 - loss: 1.8511
Test Loss: 1.8511312007904053
Test Accuracy: 0.5
1/1 -
                         0s 233ms/step
Test Precision: 0.625
```

We trained the model for 50 epochs due to the limited size of our dataset.

Evaluation Results

Training vs Validation/Test Accuracy:

 The training accuracy is relatively high (93.33%), and the test accuracy is 50%. This indicates that the model is performing better in terms of generalizing to unseen data compared to previous runs where the test accuracy was lower. There is still a bit of overfitting where training accuracy is higher than test accuracy.

Precision and Recall:

 Test precision is 0.625, which means that when the model predicts a positive class, it is correct 62.5% of the time. Test recall is 0.5, meaning the model correctly identifies 50% of all true positive instances. These metrics suggest that the model is making reasonably accurate predictions, with for improvement in both precision and recall.

Challenges and Future Suggestions

Challenges

- High training accuracy (93.33%) but low test accuracy (50%), indicating overfitting.
- Test precision (0.625) and recall (0.5) show the model can improve in identifying true positives.

Suggestions for Improvement:

- Increase the dataset size to help our model generalize better.
- Apply regularization techniques to reduce overfitting.
- Experiment with different algorithms to improve precision and recall.

Conclusion

This Dermatology Essential and Product Analyzer Bot performed well during operation. It can answer some skin care questions questions accurately including all the features that we added. The model has a high training accuracy of 93.33%, but its test accuracy drops to 50%, likely due to small dataset, leading to overfitting. It correctly predicts 62.5% of positive cases (precision) but only identifies 50% of all true positives (recall). This means the model is fairly accurate but could still improve in identifying more positive cases. Future suggestions include increasing the dataset size, trying regularization techniques to reduce overfitting, as well as experimenting with different algorithms to improve both precision and recall.

References

DATASET SOURCE

FDA Website: https://verification.fda.gov.ph/COSMETIC_ALL_INDUSTRYlist.php

LEARNING PACKET TOPICS 8-10

https://classroom.google.com/c/NzA4NTE1Mjk2MjMz/m/NzM5MzAyNjQ1OTk5/details

SUMMARY TOPIC 1-6

https://classroom.google.com/c/NzA4NTE1Mjk2MjMz/m/NzM5MzAyNjQ1OTk5/details



Thanks for watching...