Shine Skincare App - Dev Sprint Instructions

Sprint Overview

This development sprint addresses critical user experience improvements and bug fixes for the Shine Skincare app. The sprint focuses on four main areas: image thumbnail implementation with face isolation, confidence/transparency loading screens, product recommendation engine fixes, and missing product image resolution.

Sprint Duration: 5-7 days

Priority Level: High

Team Size: 2-3 developers (Frontend, Backend, DevOps)

Sprint Objectives

Primary Goals

- 1. **Image Thumbnail with Face Isolation**: Implement a thumbnail display on the results page showing the analyzed image with the detected face area circled or highlighted
- 2. **Confidence and Transparency Loading Screen**: Create an informative loading screen that shows analysis progress and confidence levels during AI processing
- 3. **Product Recommendations Engine Fix**: Debug and resolve the issue where the same products are recommended regardless of skin analysis results
- 4. **Missing Product Images**: Address the missing product images in the /public/products/ directory that are referenced in the product catalog

Success Criteria

- Users can see a thumbnail of their uploaded image with face detection visualization on results page
- Loading screen provides real-time feedback on analysis progress with confidence metrics
- Product recommendations are dynamically generated based on actual skin analysis results
- All product images display correctly in the catalog and recommendations

Task Breakdown

Task 1: Image Thumbnail with Face Isolation

Assignee: Frontend Developer

Estimated Time: 1-2 days

Priority: High

Implementation Details

The new ImageThumbnail component has been created at /components/image-thumbnail.tsx. This component needs to be integrated into the analysis results page.

Key Features: - Canvas-based rendering for precise face boundary visualization - Responsive design that maintains aspect ratio - Blue circle overlay indicating detected face region - Loading state with spinner animation - "Face Detected" badge when face bounds are available

Integration Steps: 1. Import the ImageThumbnail component into analysis-results.tsx 2. Pass the original image data and face bounds from the analysis result 3. Position the thumbnail prominently in the results layout 4. Ensure proper styling consistency with existing design system

Code Integration Example:

```
import { ImageThumbnail } from './image-thumbnail'

// In the analysis results component

<ImageThumbnail
  imageData={originalImageData}
  faceBounds={result.face_detection?.face_bounds}
    className="mb-4"
/>
```

Testing Requirements: - Test with various image sizes and aspect ratios - Verify face circle positioning accuracy - Ensure responsive behavior on mobile devices - Test loading states and error handling

Task 2: Confidence and Transparency Loading Screen

Assignee: Frontend Developer

Estimated Time: 1-2 days

Priority: High

Implementation Details

The ConfidenceLoading component has been created at /components/confidence-loading.tsx. This component provides a transparent, step-by-step view of the AI analysis process.

Key Features: - Four-stage analysis process visualization - Real-time progress indicators for each stage - Confidence percentages for each analysis step - Overall progress tracking - Transparency messaging to build user trust

Analysis Stages: 1. **Image Preprocessing** (95% confidence) - 2 seconds 2. **Face Detection** (92% confidence) - 3 seconds

3. **Skin Analysis** (88% confidence) - 4 seconds 4. **Generating Recommendations** (94% confidence) - 2.5 seconds

Integration Steps: 1. Import the component into the main analysis flow 2. Show the loading screen when analysis begins 3. Hide the loading screen when analysis completes 4. Ensure proper state management between loading and results

Code Integration Example:

```
import { ConfidenceLoading } from './confidence-loading'

const [isAnalyzing, setIsAnalyzing] = useState(false)

// Show loading screen during analysis

<ConfidenceLoading
  isVisible={isAnalyzing}
  onComplete={() => setIsAnalyzing(false)}

/>
```

Testing Requirements: - Verify timing accuracy for each stage - Test progress bar animations - Ensure proper cleanup when component unmounts - Test accessibility features for screen readers

Task 3: Product Recommendations Engine Fix

Assignee: Backend Developer **Estimated Time:** 2-3 days

Priority: Critical

Problem Analysis

The current product recommendation system in /app/api/recommendations/route.ts returns static mock data regardless of the skin analysis results. This creates a poor user experience where all users receive identical product suggestions.

Root Cause: - The recommendations API endpoint returns hardcoded product data - No integration between skin analysis results and product matching logic - Missing dynamic filtering based on skin type, concerns, and conditions

Implementation Requirements

Step 1: Create Dynamic Product Matching Logic Create a new service file /lib/product-recommendation-engine.ts:

```
interface SkinAnalysisResult {
 skin_type: string
  concerns: string[]
  conditions: string[]
 severity_levels: { [key: string]: number }
}
interface ProductRecommendation {
 product: Product
  relevance_score: number
 reasoning: string
}
export class ProductRecommendationEngine {
  static generateRecommendations(
   analysisResult: SkinAnalysisResult
  ): ProductRecommendation[] {
   // Implementation logic here
}
```

Step 2: Update Recommendations API Modify /app/api/recommendations/route.ts to: - Accept skin analysis results as input parameters - Use the new recommendation engine - Return personalized product suggestions - Include reasoning for each recommendation

Step 3: Integrate with Analysis Results Update the analysis results component to: - Call the recommendations API with analysis data - Display personalized product suggestions - Show reasoning for each recommendation

Algorithm Requirements: - Weight products based on skin type compatibility - Prioritize products that address detected concerns - Consider severity levels for condition-specific recommendations - Implement fallback logic for edge cases

Testing Requirements: - Test with various skin analysis scenarios - Verify recommendation diversity across different skin types - Ensure API performance under load - Test error handling for invalid input data

Task 4: Missing Product Images Resolution

Assignee: DevOps/Frontend Developer

Estimated Time: 1 day

Priority: Medium

Problem Analysis

The product catalog references images in /public/products/ directory, but this directory doesn't exist. All product images are currently broken, showing placeholder or missing image icons.

Missing Images List: Based on /lib/products.ts, the following images need to be sourced: - iS Clinical Cleansing Complex.jpg - Dermalogica UltraCalming Cleanser.webp - SkinCeuticals C E Ferulic.webp - TNS® Advanced+ Serum.jpg - PCA SKIN Pigment Gel Pro.jpg - First Aid Beauty Ultra Repair Cream.webp - EltaMD UV Clear Broad-Spectrum SPF 46.webp - Allies of Skin Molecular Silk Amino Hydrating Cleanser.webp - Naturopathica Calendula Essential Hydrating Cream.webp - Obagi CLENZIderm M.D. System (Therapeutic Lotion).webp

Implementation Steps

Step 1: Create Products Directory

```
mkdir -p public/products
```

Step 2: Source Product Images For each product, obtain high-quality images through: - Official brand websites - Authorized retailer websites (Sephora, Ulta, Amazon) - Stock photo services with proper licensing - Brand press kits or media resources

Step 3: Image Optimization - Resize images to consistent dimensions (300x300px recommended) - Optimize file sizes for web delivery - Ensure proper format selection (WebP for modern browsers, JPG fallback) - Implement responsive image loading

Step 4: Fallback Implementation Create a fallback system for missing images:

```
const ProductImage = ({ src, alt, ...props }) => {
  const [imageSrc, setImageSrc] = useState(src)

  const handleError = () => {
    setImageSrc('/placeholder-product.jpg')
  }

  return <img src={imageSrc} alt={alt} onError={handleError} {...props} />
}
```

Legal Considerations: - Ensure all images have proper usage rights - Document image sources and licensing - Consider creating custom product photography if licensing is

unclear - Implement proper attribution where required

Testing Requirements: - Verify all product images load correctly - Test fallback behavior for missing images - Ensure images display properly across different devices - Test loading performance and optimization

Technical Specifications

Frontend Architecture Updates

Component Integration Map

State Management Requirements

The application will need enhanced state management to handle: - Image data persistence between upload and results - Loading states for different analysis phases - Analysis results caching for thumbnail display - Product recommendation state management

Recommended State Structure:

```
interface AppState {
    analysis: {
        isLoading: boolean
        currentStep: string
        progress: number
        originalImage: string | null
        results: AnalysisResult | null
}
recommendations: {
        isLoading: boolean
        products: ProductRecommendation[]
        error: string | null
}
```

Performance Considerations

- Implement image compression for uploaded photos
- Use lazy loading for product images
- Optimize canvas rendering for face detection visualization
- Implement proper cleanup for canvas elements

Backend Architecture Updates

API Endpoint Modifications

New Endpoint: Enhanced Recommendations

```
POST /api/v4/recommendations/personalized
Content-Type: application/json
Request Body:
  "skin_analysis": {
    "skin_type": "combination",
    "concerns": ["acne", "hyperpigmentation"], "conditions": ["mild_acne"],
    "severity_levels": {
      "acne": 0.6,
      "hyperpigmentation": 0.4
    }
  },
  "user_preferences": {
    "budget_range": "mid",
    "ingredient_preferences": ["natural", "fragrance-free"],
    "routine_complexity": "simple"
}
Response:
  "recommendations": [
      "product_id": "is-clinical-cleansing",
      "relevance_score": 0.92,
      "reasoning": "Salicylic acid content addresses mild acne concerns",
      "category_priority": 1
    }
  1,
  "total_recommendations": 6,
  "confidence score": 0.88
}
```

Updated Endpoint: Analysis Results The existing analysis endpoints need to return additional data for thumbnail generation:

```
{
    "analysis_results": {
        // existing fields...
        "original_image_data": "base64_string",
        "face_detection": {
            "detected": true,
            "confidence": 0.92,
            "face_bounds": {
                 "x": 150,
                 "y": 100,
                 "width": 200,
                 "height": 250
            }
        }
    }
}
```

Database Schema Updates

If using a database for product recommendations, consider these schema additions:

Products Table Enhancements:

```
ALTER TABLE products ADD COLUMN skin_type_compatibility JSON;
ALTER TABLE products ADD COLUMN concern_targeting JSON;
ALTER TABLE products ADD COLUMN ingredient_profile JSON;
ALTER TABLE products ADD COLUMN effectiveness_rating DECIMAL(3,2);
```

New Recommendation Logs Table:

```
CREATE TABLE recommendation_logs (
  id SERIAL PRIMARY KEY,
  user_id VARCHAR(255),
  analysis_id VARCHAR(255),
  recommended_products JSON,
  confidence_score DECIMAL(3,2),
  created_at TIMESTAMP DEFAULT CURRENT_TIMESTAMP
);
```

Infrastructure Considerations

AWS S3 Integration

Since the user hosts websites on Amazon S3, consider: - Storing product images in S3 with CloudFront CDN - Implementing proper CORS configuration for image access - Using S3 for analysis result caching - Implementing image optimization pipeline

S3 Bucket Structure:

Performance Monitoring

Implement monitoring for: - Analysis processing time - Image upload and processing latency - Recommendation generation speed - User engagement with recommendations

Security Considerations

Image Upload Security

- Implement file type validation
- Add file size limits (max 10MB)
- Scan uploaded images for malicious content
- Use temporary storage with automatic cleanup

API Security

- Implement rate limiting for analysis endpoints
- Add request validation and sanitization
- Use HTTPS for all API communications
- Implement proper error handling without information leakage

Data Privacy

- Ensure uploaded images are not permanently stored
- Implement proper user consent mechanisms
- Add data retention policies
- Consider GDPR compliance for EU users

Testing Strategy

Unit Testing Requirements

Frontend Component Tests

Each new component requires comprehensive unit tests:

ImageThumbnail Component Tests:

```
describe('ImageThumbnail', () => {
  test('renders canvas with correct dimensions', () => {
     // Test canvas sizing logic
  })

test('draws face detection circle when bounds provided', () => {
     // Test face circle rendering
  })

test('handles missing face bounds gracefully', () => {
     // Test fallback behavior
  })

test('shows loading state during image processing', () => {
     // Test loading state management
  })
})
```

ConfidenceLoading Component Tests:

```
describe('ConfidenceLoading', () => {
  test('progresses through all analysis stages', () => {
    // Test stage progression logic
  })

test('displays correct confidence percentages', () => {
    // Test confidence display
  })

test('calls onComplete when analysis finishes', () => {
    // Test completion callback
  })

test('handles component unmounting during analysis', () => {
    // Test cleanup logic
  })
})
```

Backend API Tests

Product Recommendations Engine Tests:

```
def test_recommendation_engine():
   # Test various skin analysis scenarios
    analysis_result = {
        "skin_type": "oily",
"concerns": ["acne", "large_pores"],
        "conditions": ["moderate_acne"]
    }
    recommendations =
ProductRecommendationEngine.generate_recommendations(analysis_result)
    assert len(recommendations) > 0
    assert all(rec.relevance_score > 0.5 for rec in recommendations)
    assert any("acne" in rec.reasoning.lower() for rec in recommendations)
def test_recommendation_api_endpoint():
    # Test API endpoint with various inputs
    response = client.post('/api/v4/recommendations/personalized', json={
        "skin_analysis": test_analysis_data
    })
    assert response.status_code == 200
    assert "recommendations" in response.json()
```

Integration Testing

End-to-End User Flow Tests

- 1. Complete Analysis Flow:
- 2. Upload image → Show loading screen → Display results with thumbnail → Show recommendations
- 3. Verify each step transitions correctly
- 4. Test error handling at each stage
- 5. Product Recommendation Flow:
- 6. Analyze different skin types → Verify different recommendations
- 7. Test recommendation reasoning accuracy
- 8. Verify product image loading
- 9. Cross-Browser Compatibility:

- 10. Test on Chrome, Firefox, Safari, Edge
- 11. Verify canvas rendering across browsers
- 12. Test responsive design on various screen sizes

Performance Testing

- Load testing for analysis endpoints
- Image processing performance benchmarks
- Recommendation generation speed tests
- Frontend rendering performance analysis

Quality Assurance Checklist

Pre-Deployment Verification

- [] All product images load correctly
- [] Face detection visualization is accurate
- [] Loading screen shows appropriate progress
- [] Recommendations vary based on analysis results
- [] Error states are handled gracefully
- [] Mobile responsiveness is maintained
- [] Accessibility standards are met
- [] Performance benchmarks are satisfied

User Acceptance Testing

- [] Upload various image types and sizes
- [] Test with different lighting conditions
- [] Verify recommendations make sense for different skin types
- [] Test user flow from start to finish
- [] Validate loading screen provides value to users

Deployment Instructions

Pre-Deployment Setup

Environment Configuration

Update environment variables for production:

```
# Frontend (.env.local)
NEXT_PUBLIC_API_BASE_URL=https://api.shineskincollective.com
NEXT_PUBLIC_ENABLE_ANALYTICS=true

# Backend (.env.production)
S3_BUCKET=shine-skincare-production
REDIS_URL=redis://production-redis:6379
DATABASE_URL=postgresql://production-db
```

Database Migrations

If using database for recommendations:

```
# Run migration scripts
python manage.py migrate
python manage.py seed_product_data
```

Deployment Process

Step 1: Backend Deployment

```
# Build and deploy backend
cd backend/
docker build -t shine-backend:latest .
docker push your-registry/shine-backend:latest

# Update production deployment
kubectl apply -f k8s/production/
```

Step 2: Frontend Deployment

```
# Build optimized frontend
npm run build
npm run export

# Deploy to S3 (user's preferred hosting)
aws s3 sync out/ s3://shine-skincare-production --delete
aws cloudfront create-invalidation --distribution-id YOUR_DISTRIBUTION_ID --
paths "/*"
```

Step 3: Asset Deployment

```
# Upload product images to S3
aws s3 sync public/products/ s3://shine-skincare-assets/products/ --acl public-
read

# Update CDN configuration
aws cloudfront create-invalidation --distribution-id ASSETS_DISTRIBUTION_ID --
paths "/products/*"
```

Post-Deployment Verification

Health Checks

1. API Endpoints: bash curl https://api.shineskincollective.com/health
 curl
 https://api.shineskincollective.com/api/v4/recommendations/health

2. Frontend Functionality:

- 3. Test image upload and analysis
- 4. Verify loading screen appears
- 5. Check thumbnail generation
- 6. Validate product recommendations

7. Asset Loading:

- 8. Verify all product images load
- 9. Test CDN performance
- 10. Check image optimization

Monitoring Setup

- Configure application performance monitoring
- Set up error tracking and alerting
- Monitor recommendation engine performance
- Track user engagement metrics

Rollback Plan

Emergency Rollback Procedure

- 1. Frontend Rollback: bash # Restore previous S3 deployment aws s3 sync
 s3://shine-skincare-backups/previous-version/ s3://shine-skincareproduction/
- 2. **Backend Rollback:** bash # Revert to previous container version kubectl rollout undo deployment/shine-backend
- 3. **Database Rollback:** bash # Restore database if schema changes were made pg_restore --clean --if-exists -d production_db backup_pre_sprint.sql

Rollback Triggers

- Analysis success rate drops below 90%
- Recommendation engine errors exceed 5%
- Frontend loading time increases by more than 50%
- User-reported critical bugs

Success Metrics

Key Performance Indicators

- Analysis Completion Rate: Target >95%
- User Engagement with Recommendations: Target >60% click-through
- Loading Screen User Satisfaction: Target >4.0/5.0 rating
- Image Thumbnail Accuracy: Target >90% user approval

• Page Load Time: Target <3 seconds

• **Error Rate:** Target <2%

Monitoring Dashboard

Set up real-time monitoring for: - Analysis processing times - Recommendation generation speed - User interaction patterns - Error rates and types - Performance metrics

This comprehensive sprint plan addresses all identified issues while maintaining high code quality and user experience standards. The modular approach allows for parallel development and reduces deployment risks.