cisco life!

# GO BEYOND

#CiscoLiveAPJC



"Reconciliation" - Dustin Koa Art

cisco Live!

illiili CISCO

# Future-Proof Your Workplace with Gen Al

Across Security, Networking, & Collaboration

#### Richard Jang

Senior Product Manager, Generative Al & Cybersecurity Cisco Live Distinguished Speaker BRKAPP-2546



#### Cisco Webex App

#### Questions?

Use Cisco Webex App to chat with the speaker after the session

#### How

- 1 Find this session in the Cisco Live Mobile App
- 2 Click "Join the Discussion"
- 3 Install the Webex App or go directly to the Webex space
- 4 Enter messages/questions in the Webex space

Webex spaces will be moderated by the speaker until November 15, 2024.

https://ciscolive.ciscoevents.com/ciscolivebot/#BRKAPP-2546





# Agenda

#### What is Al?

· Background, key concepts, definitions

#### Al Deep Dive

· How AI, machine learning, deep learning

#### Generative Al Deep Dive

• Types of models, training, capabilities, prompt design, etc.

#### Cisco's Generative Al Strategy

- Generative Al Assistants for security, networking, collaboration
- Unifying generative across Cisco with the Unified Al Assistant

#### Conclusion

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#### Richard Jang



Expertise

Generative AI, Cybersecurity, Enterprise Networking

Personal Life

Taiwanese American, Silicon Valley Native, San Jose State University

Hobbies

Playing Drum Set and Seeing The world!

Add me on LinkedIn!



http://linkedin.com/in/jangrichard/



Artificial Intelligence What is it?



# "Al is software that mimics human intelligence and performs tasks to solve problems."

ChatGPT, November 7 2024

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#### Artificial Intelligence is Everywhere

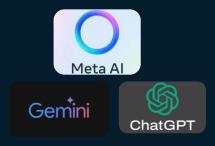


Facial Recognition

Deep Learning



Voice Activated Assistants Natural Language Processing



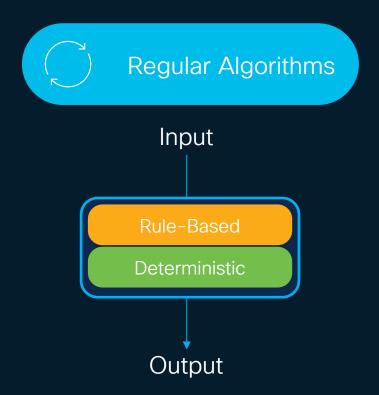
Chat Bots Generative Al

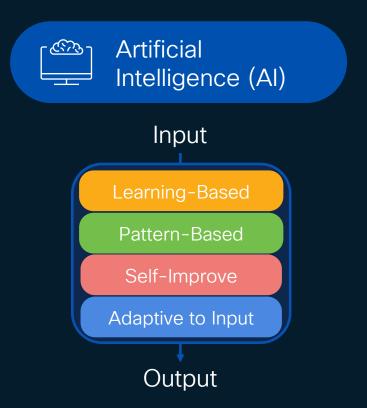


#### How Is Al Different From Regular Algorithms?

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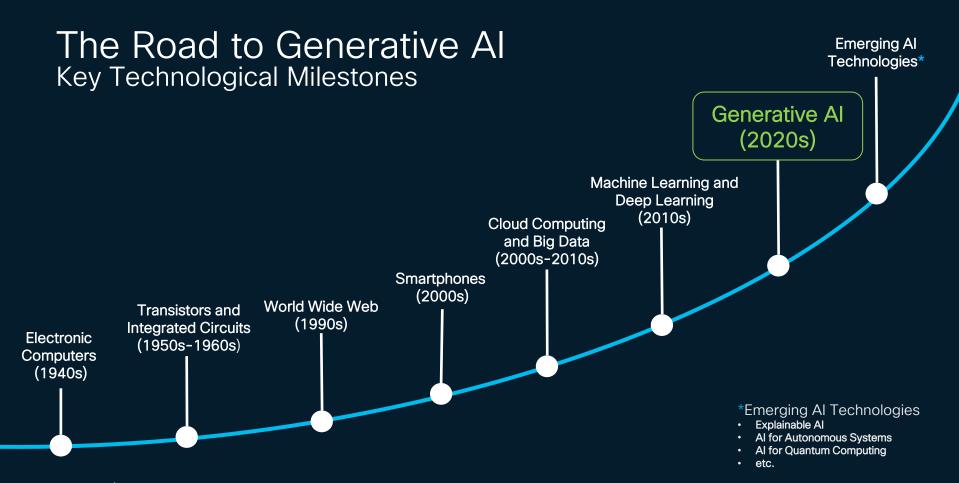


#### A Growing Phenomenon

#### ChatGPT Sprints to One Million Users







### Al Technical Deep Dive



#### The Breakdown of Artificial Intelligence

#### Artificial Intelligence

A discipline that encompasses all aspects of simulating human intelligence in machines.



#### The Breakdown of Artificial Intelligence

**Artificial Intelligence** 

#### **Machine Learning**

Uses statistical models to learn from data to perform tasks without explicit programming.



#### Supervised Learning

#### Supervised Learning

Using past "labeled" data to predict future trends

- Spam email identifier
- Stock price prediction
- Sales forecast

Note: Labeled data is data that has been tagged with the correct answer or output.

#### Scenario: Predicting if an Email is Spam Email 1 Email 2 To/From To/From Subject Subject Content Content Email 3 Email 4 To/From To/From Subject Subject Content Content Spam Labels Not Spam

#### Unsupervised Learning

**Unsupervised Learning** 

Using "Unlabeled" data to learn patterns

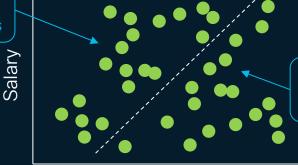
- User segmentation
- Anomaly detection
- Image/Video analysis

**Note**: Unlabeled data refers to data that does not have predefined categories or outputs.

Scenario: Predicting if an employee is going to be a top performer

Clustering Employee Data





Cluster 2
Low Performers

Years at Company

Unlabeled Data — Employee Data

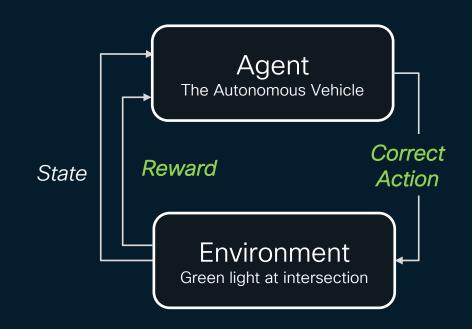


#### Reinforcement Learning

#### Reinforcement Learning

Trained on reward or penalty feedback loop based on its actions during simulations.

- Autonomous vehicles
- Robotics
- Resource management



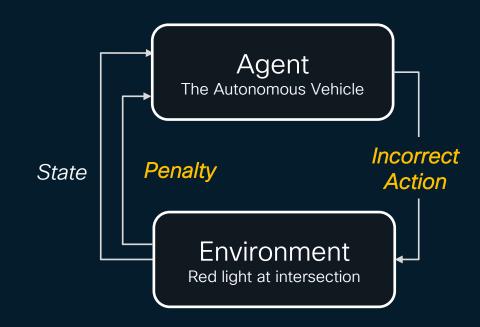


#### Reinforcement Learning

#### Reinforcement Learning

Trained on reward or penalty feedback loop based on its actions during simulations.

- Autonomous vehicles
- Robotics
- Resource management





#### The Breakdown of Artificial Intelligence

**Artificial Intelligence** 

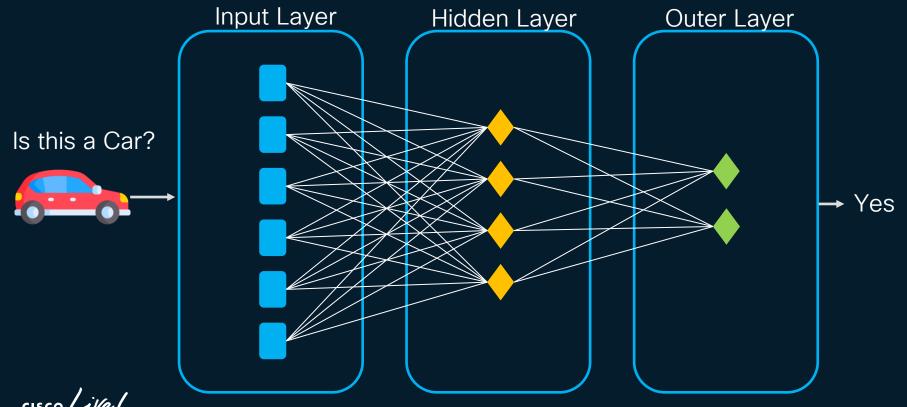
**Machine Learning** 

#### Deep Learning

Utilizes neural networks to model & interpret complex patterns in large datasets.



# Deep Learning Uses Neural Networks to Identify Patterns Divide and conquer large amounts of complex data



#### The Breakdown of Artificial Intelligence

**Artificial Intelligence** 

**Machine Learning** 

**Deep Learning** 

Generative Al

Al that produces content.



#### LLMs vs Diffusion Models vs GANs Few key Generative Al models

# Large Language Models

Trained to create text content.

Ex: ChatGPT 4o

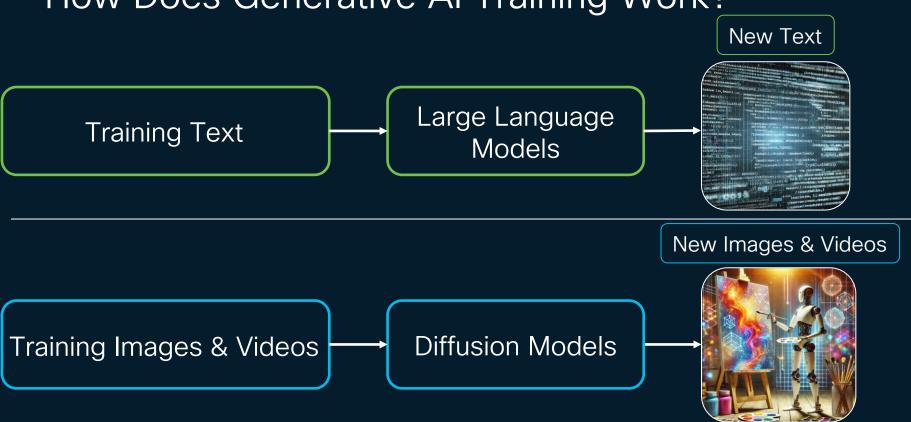
#### **Diffusion Models**

Trained to create artsy image and video content.

Ex: DALL·E 3



#### How Does Generative Al Training Work?





# Generative Al Capabilities



#### Generative Al: Text Capabilities in Creative Writing

Prompt: Write a poem of Richard's Presentation at Cisco Live Melbourne

#### Generated Text

In Melbourne's heart, where tech's alive, It's Cisco Live, 2024, we thrive!
Richard Jang takes the stage with pride, The Cisco Al Assistant by his side.

We'll laugh, we'll learn, we'll reach new heights,
With Al guiding through the bytes.
So buckle up, and take the ride,
At Cisco Live, tech's on our side!



#### Generative Al: Text Capabilities in Planning

Prompt: I'm visiting Melbourne for the first time in Nov I want some recommendations

#### Generated Text

Day 1: Explore the Heart of Melbourne

8:00 AM - 9:00 AM: Breakfast at a Laneway Café
Start your day with a coffee and breakfast at a popular laneway café like
Manchester Press or Hardware Société. Melbourne is known for its amazing coffee and brunch culture.

9:15 AM - 11:00 AM: Walk Through Hosier Lane and Laneways
Explore Hosier Lane and surrounding laneways, famous for street art. Snap some photos of the murals and wander through Flinders Lane.

11:15 AM - 12:30 PM: Federation Square & ACMI Visit Federation Square and spend some time at the ACMI (Australian Centre for the Moving Image) to explore digital culture and film history.



#### Generative Al: Text Capabilities at Work Prompt: Enhance my email to become more professional

#### Before

Hi [Coworker's Name]

I'm not happy with how things went today. You missed important deliverables, and it looked bad in front of the client. You need to need to fix this.

Let's talk soon, Richard

#### After

Hi [Coworker's Name]

I wanted to bring up the recent client meeting where a few deliverables were overlooked. To ensure we stay on track and maintain client confidence, could we arrange a time to discuss how we can prevent this moving forward?

I value your input, and I believe we can address this effectively together.

Let's talk soon. Richard



# Generative AI: Code Capabilities for Games Prompt: Provide me code to build the Flappy Bird game.

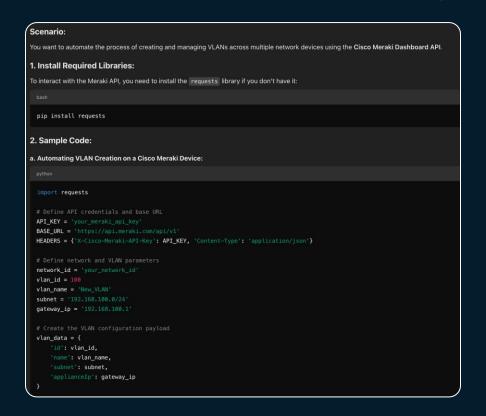
```
import pygame, random
# Initialize the game
pygame.init()
# Game variables
WIDTH, HEIGHT = 400, 600
BIRD_X, BIRD_Y, GRAVITY = 50, HEIGHT // 2, 0.5
bird_movement = 0
screen = pygame.display.set_mode((WIDTH, HEIGHT))
# Load images
bird_img = pygame.image.load('bird.png')
bg_img = pygame.image.load('background.png')
pipe_img = pygame.image.load('pipe.png')
   pipe_height = random.randint(150, 450)
   return pipe_img.get_rect(midtop=(WIDTH, pipe_height)), pipe_img.get_rect(midbottom=(WIDTH, pipe_height - 150))
# Main game loop
pipes = [create_pipe()]
clock = pygame.time.Clock()
```



Generated with DALL-E 3



## Generative AI: Code Capabilities for IT Admins Prompt: Automate VLAN Creation on a Cisco Catalyst Switch





#### Generative AI: Image Capabilities

Prompt: Generate an image representing Melbourne Australia





# Generative AI: Image Capabilities Prompt: Have a dog on a rocket fly over





# Generative AI: Image Capabilities Prompt: Make it intense and cyberpunk.





#### If Use Correctly, The Possibilities are Endless









Large Language Models (LLMs) Deep Dive

#### How are LLMs Trained for Text and Code?

Step 1: Data Collection (Feeding Knowledge)

**Step 2: Tokenization** (Breaking It Down)

Step 3: Parameter Learning (Storing Knowledge)

Step 4: Fine-Tuning (Specialized Learning)



# Step 1: Data Collection (Feeding Knowledge)

#### **What Happens?**

• LLMs are trained on massive amounts of text data—books, articles, websites, and more.

#### **Analogy:**

• Giving a child access to a library of books, the more they read, the more they learn.



Fun Fact: GPT-4 was trained on terabytes of text, equivalent to hundreds of millions of books.



# Step 2: Tokenization and Vectorization Breaking it Down

#### **How It Works:**

- The text is split into tokens
   (words, subwords, or characters)
   so the model can process it.
- Tokens are further split into vectors (numerical values)

#### **Analogy**:

 Teaching a child to break down sentences into words & letters.

```
Raw Text
          "My name is Richard"
Tokenized Text
 ["My", "name", "is", "Richard"]
Vectorized Tokens
"My" \rightarrow [0.12, -0.43, 0.33, 0.85, -0.17]
"name"-> [0.52, 0.10, -0.21, 0.44, -0.09]
"is" -> [0.09, -0.15, 0.47, 0.13, 0.56]
"Richard" \rightarrow [0.67, -0.25, -0.33, 0.78, 0.45]
```

# Step 3: Parameters Learning (Storing Knowledge)

#### What Happens?

Vectors flow through neural networks; parameters learn token relationships.

#### Analogy:

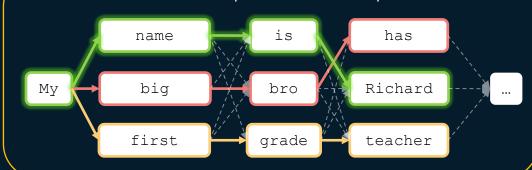
A child learns how words fit together to form sentences.

#### Vectorized Text

```
"My" \rightarrow [0.12, -0.43, 0.33, 0.85, -0.17]
"name"-> [0.52, 0.10, -0.21, 0.44, -0.09]
"is" \rightarrow [0.09, -0.15, 0.47, 0.13, 0.56]
"Richard" \rightarrow [0.67, -0.25, -0.33, 0.78, 0.45]
```

#### Neural Network

Parameters store relationships between tokens to predict next words.





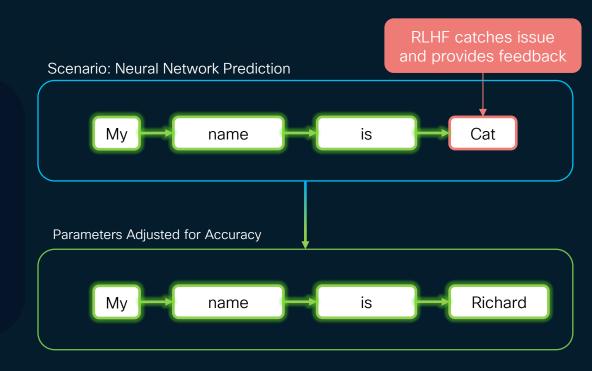
#### Step 4: Fine-Tuning the Model (Optimizing Predictions)

#### What happens?

- Parameters are adjusted to minimize prediction errors.
- The model improves by learning from its mistakes

#### Analogy:

 A child practices speaking by receiving feedback & adjusting.



# Diffusion Model Deep Dive





# How are Diffusion Models Trained for Images?

Step 1: Data Collection (Feeding Knowledge)

Step 2: Noise Addition (Disrupting the Image)

Step 3: **Denoising** (Restoring the Image)

# Step 1: Data Collection (Feeding Knowledge)

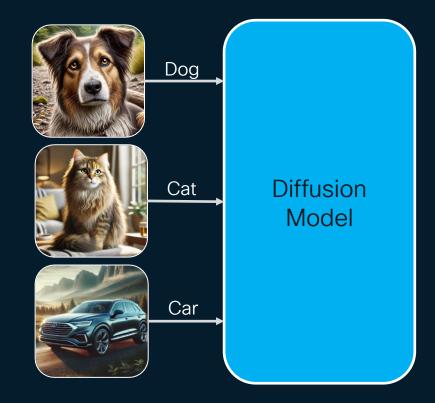
#### **What Happens?**

 Feel model large datasets of images paired with descriptive text to learn.

#### **Analogy:**

 Given a person photos with captions they learn to associate certain text with specific visual patterns.

Fun Fact: DALL·E is trained on billions of images with text descriptions.





# Step 2: Add Noise (Disrupting the Image)

#### What Happens?

The model adds noise to the image until it's unrecognizable.

#### **Analogy:**

Crumpling a piece of so it's harder to recognize the original content.



Fun Fact: The noise used in diffusion models isn't random—it's carefully structured to ensure that when the model learns to denoise, it's learning how to generate fine details.



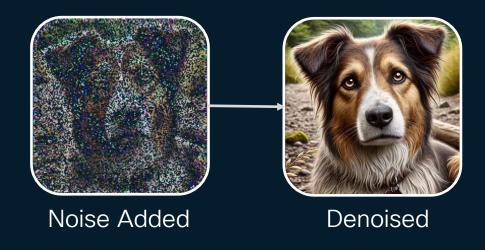
# Step 3: **Denoising** (Restoring the Image)

#### **What Happens?**

The model attempts to reconstruct the noisy image to its original form while learning in the process.

#### **Analogy:**

Learning to draw the artwork from that crumpled paper.



## How Diffusion Model Converts Text to Image

Step 1: Understanding the Prompt

Step 2: Noise Initialization

Step 3: Denoising with Text

Step 4: Image Refinement



# Step 1: Understanding the Prompt

Prompt: Generate an image of a dog on a flying carpet!

# Tokenized Text ["Generate", "an", "image", "of", "a", "dog", "on", ...] Vectorized Tokens [0.23, -0.56, 1.04, 0.78, 0.99, 1.21, 0.89, 0.32, ...] Neural Network Layers Analyzes and understands the meaning and context



## Step 2: Noise Initialization

Prompt: Generate an image of a dog on a flying carpet!

#### How It Works

- Start with a random noisy image.
- Text encoding guides the process using learned text-image pairs.



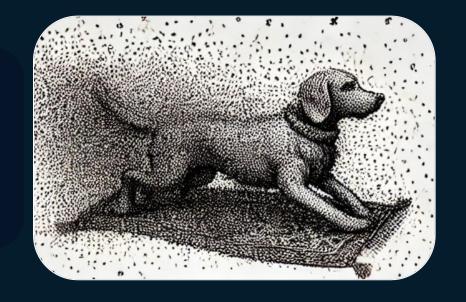


## Step 3: Denoising with Text Guidance

Prompt: Generate an image of a dog on a flying carpet!

#### How It Works

- The model removes noise using patterns learned from training.
- The shapes of the dog and carpet become clearer.





# Step 4: Refining the Image

Prompt: Generate an image of a dog on a flying carpet!

#### How It Works

- Noise is entirely removed.
- Final image of the dog flying on the magic carpet is complete.





What's a model trained on everything called?

#### A Foundational Generative Al Model!

#### **Jack of All Trades Model:**

- Pre-trained on vast datasets including text, images, code, etc.
- Can handle a broad array of questions across domains.





# However, foundational models do have limitations...



# Lack of

Real-Time Data

Domain Specific Data

Up-to-Date Data

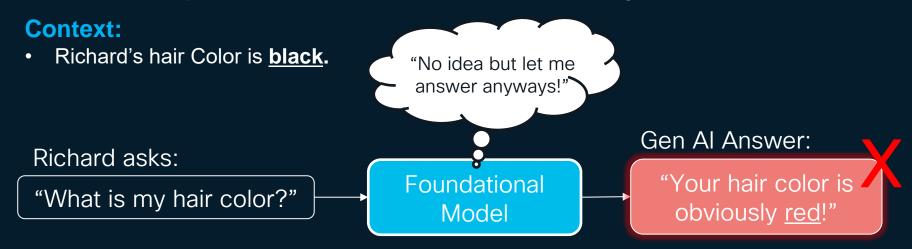
# can cause <u>Hallucinations</u>



#### What is a Gen Al Hallucination?

#### **Definition:**

 When AI models generate information that is plausible but incorrect or completely made up often due to insufficient training data.



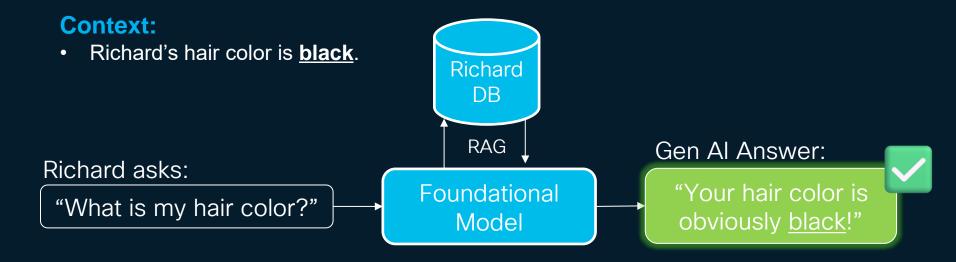
# How Are Hallucinations Mitigated?



# Retrieval Augmented Generation (RAG)

#### **Definition:**

Allows the Al Model to query external sources for data.





# How to use Gen Al effectively?



# Prompt Design and Engineering!

The strategy of crafting specific input to produce the best response



# Strong Search Engine Prompt



# Strong Gen Al Prompt



# With Gen Al, Context and Details Matter!



# Prompt Design Example Scenario

#### User's Intent:

Wants to plan a healthy meal for a family on a budget.

#### Likely Search Engine Prompt:

"Healthy budget meal plan for family"

Not a good prompt for Gen Al





# Al Response Using Search Engine Prompts

Original Prompt: "Healthy budget meal plan for family"

#### Al Response:

- Meal 1: Spaghetti with Marinara Sauce
   Simple, affordable dish with pasta and tomato sauce.
- Meal 2: Chicken Stir-Fry
   Stir-fry chicken with mixed vegetables and rice.



# Guideline 1: Be Specific

# Improved Prompt:

"What are some healthy meals I can prepare for a <u>family of four</u> on a budget?"



#### Guideline 2: Give Context

# Improved Prompt:

"What are some healthy meals I can prepare for a family of four on a \$50 weekly grocery budget?"



# Guideline 3: Use Examples

## Improved Prompt:

"What are some healthy meals I can prepare for a family of four on a \$50 weekly grocery budget?

Include specific ingredients like pasta and chicken."



# Guideline 4: Breakdown Complex Tasks

## Improved Prompt:

"First, list out healthy meals I can prepare for a family of four on a \$50 weekly budget including specific ingredients such as pasta and chicken.

Then, provide a simple recipe for each meal."



#### Guideline 5: Refine - Add a Persona

#### Improved Prompt:

"Imagine you're a professional chef creating a meal plan for a family of four on a \$50 weekly grocery budget including ingredients such as pasta and chicken.

First, list the meals you would suggest, then provide a simple recipe for each meal with affordable ingredients."



## After Prompt Design - Detailed & Precise

Improved Prompt: "Imagine you're a professional chef creating a meal plan for a family of four on a \$50 weekly grocery budget...."

#### Al Response:

#### Meal 1: Lemon Garlic Chicken Pasta

- Ingredients: Chicken breast, pasta, garlic, lemon, olive oil, parmesan cheese.
- Recipe:
  - Cook pasta and set aside.
  - Sauté garlic in olive oil, then add chicken and lemon juice.
  - Toss with pasta and top with parmesan cheese.
     Cost Estimate: Around \$10 for 4 servings.

#### Meal 2: Chicken and Vegetable Stir-Fry

- Ingredients: Chicken thighs, frozen mixed vegetables, soy sauce, garlic, rice.
- · Recipe:
  - Stir-fry chicken in a pan with garlic and soy sauce.
  - Add mixed vegetables and cook until tender.
  - Serve over rice.
     Cost Estimate: Around \$8 for 4 servings.



# You've Mastered the the Basics!





What is Cisco's Al Strategy?



# Our Al strategy... To connect and protect the Al era.



## Bolstering our Al capabilities with acquisitions, investments and strategic partnerships





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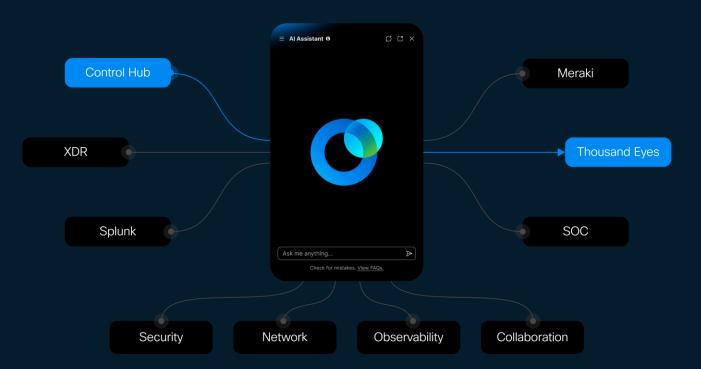
## Al Across Cisco's Product Portfolio

	Networking	Security	Collaboration	Observability	CX
AI Generative AI	<ul> <li>Enabled by Silicon One Scheduled Fabric Ethernet Solution</li> <li>Next-gen AlOps &amp; Assurance</li> <li>Al Assistant for Networking</li> </ul>	<ul> <li>Al Assistant Experience Purpose-built LLMs</li> <li>Autonomous Actions</li> </ul>	<ul> <li>Language Intelligence</li> <li>New Webex Al Assistant</li> <li>Intelligent Contact Center</li> <li>Realtime Media Models</li> </ul>	<ul><li>Model observability</li><li>Prompt interface</li><li>Al assistant for summarization</li></ul>	<ul> <li>Engineer copilots</li> <li>Automated code creation</li> <li>Customer Q&amp;A &amp; Sentiment analysis</li> <li>Root cause analysis</li> <li>Self-remediation</li> <li>Adoption advisory</li> </ul>
Predictive AI	<ul> <li>Network Management</li> <li>Network Assurance</li> <li>Anomaly Detection</li> <li>Change Automation</li> <li>Predictive Internet</li> </ul>	<ul> <li>Email Compromise Analysis</li> <li>Threat Detection</li> <li>Endpoint Trust Monitor</li> <li>Malware Protection</li> <li>Data Protection</li> <li>Alert Prioritization</li> </ul>	<ul> <li>Background noise removal</li> <li>Voice focus</li> <li>Bandwidth extension</li> <li>People focus / frames</li> <li>Personal/Team insights</li> <li>Cinematic Meetings</li> <li>Meeting Zones</li> </ul>	<ul> <li>Statistical Modeling</li> <li>Baselining</li> <li>Anomaly Detection</li> <li>Intelligent Automation</li> <li>Data Tagging</li> </ul>	<ul> <li>Virtual Engineers</li> <li>Analytics Engine</li> <li>Anomaly Detection</li> <li>Sentiment Prediction</li> <li>Managed Insights</li> <li>CX Cloud - Insights &amp; Analytics</li> </ul>



## Cisco Al Assistant

Bringing together "Skills" across Cisco to 10x customer experience





## Al Assistants Have "Skills", Not Features

#### Al Skills

- Definition: Any action that a Cisco Al Assistant can performance.
- Skills: Troubleshooting, configuration, recommendations, etc.





## Let's Deep Dive into Cisco's Generative Al Portfolios!



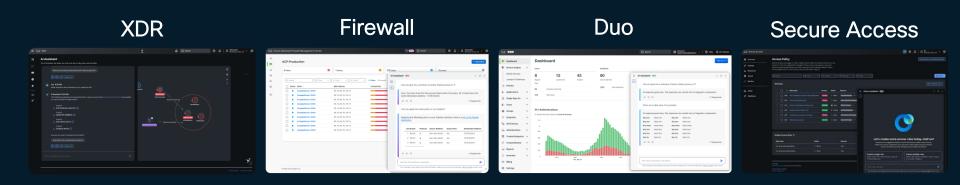


for Collaboration

Generative Al for for Security



## Cisco Security's Suite of Al Assistants



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BRKAPP-2546

Detect a phishing attack that has setup a C&C and is exfiltrating data outside the network.

Block any outbound exfiltration to the IP address identified from the C&C

Lock affected user out of critical applications

Ensure users access only resources they need securely



## Al Assistant for

Firewall Management Center



## Cisco Firewall Al Assistant Skills

#### **Documentation Summarization**

Search Cisco's product docs for clear, actionable, and simplified insights

#### Policy Rule Creation

Create policies with natural language

Ticketing Integration with CX

Enables TAC case creation and modification



#### Policy Insights

Enables quick discovery of policies with fast, rich data responses on-demand

#### Policy Analysis & Optimization

Proactively discover and remediate policy misconfigurations and anomalies

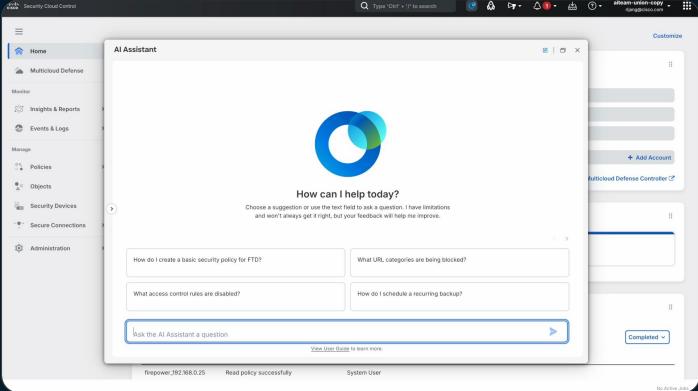
#### AlOps Integration

Insights into firewall traffic, configuration, capacity to enhance security posture.





## CX Ticketing Skill





Al Assistant for Cisco Secure Access



## Modernize your defense with Cisco Secure Access

Converged cloud-native security grounded in zero trust Users Web Everywhere Managed and Public From anything Secure Access To anywhere unmanaged devices SaaS apps Private apps loT devices SAFER FOR EVERYONE



### Cisco Secure Access Al Assistant Skills

#### **Documentation Summarization**

Search Cisco's product docs for clear, actionable, and simplified insights

#### Policy Rule Creation

Create policies with natural language

#### Troubleshooting User Access

Automates triaging for access issues



#### Policy Insights

Enables quick discovery of policies with fast, rich data responses on-demand

#### Policy Analysis & Optimization

Proactively discover and remediate policy misconfigurations and anomalies

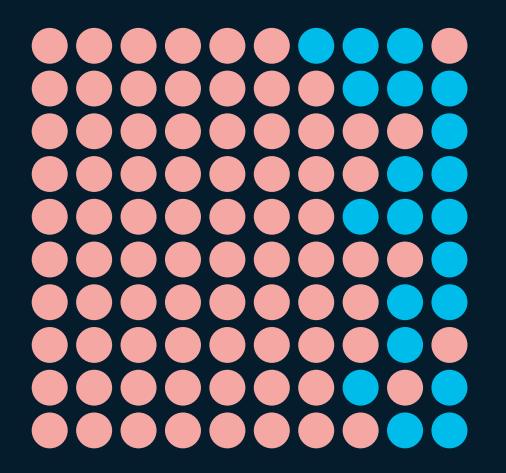
#### Digital Experience Monitoring

Quick issue research, trend analysis, insights



# Al Assistant for Cisco Duo







of breaches leveraged identity as a key component

Cisco Talos Incident Response Data 2023-2024



## Protect Against Identity Threats with Cisco Duo



Identify

• Identity posture evaluation

Cross-platform identity visibility

Baseline of identity behavior



- Enforce most effective MFA
- Establish device trust

Protect

Build dynamic, risk-based access policy



Correlate activity across IAM platforms

Detect identity-based attack patterns

Monitor threats & changes in posture



- Step-up or block risky logins
- Powerful remediation actions
- Automate quarantine actions

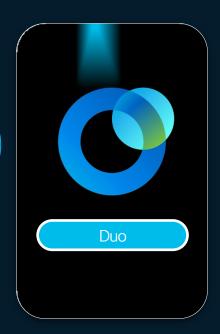




### Cisco Duo Al Assistant Skills

#### **Documentation Summarization**

Search Cisco's product docs for clear, actionable, and simplified insights



#### Troubleshooting User Access

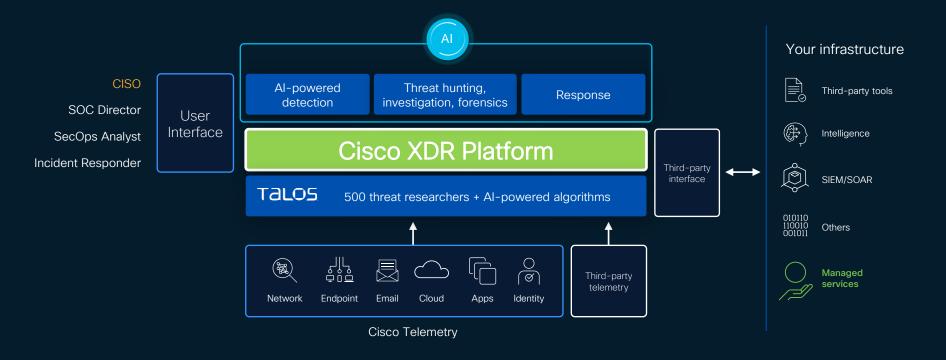
Troubleshoots auth issues by looking at auth sequences, policy breakdown, group status, etc.



# Al Assistant for Cisco XDR



## Complexity, simplified with an Al-first XDR



#CiscoLiveAPJC



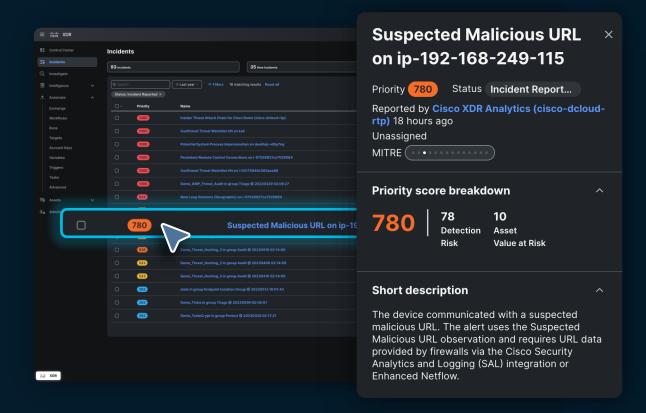
## Improve alert fidelity | XDR + Al Assistant

Deliver risk- and impact-based prioritizations with threat correlation

Investigate, prioritize, respond, and recover from one console

Command every response and action with Al-driven automation

Get confident verdict with cross-domain telemetry and threat intelligence enrichment





## Cisco XDR Al Assistant Skills

#### **Documentation Summarization**

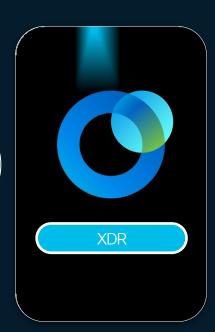
Provides concise incident summary to enable quick understanding of an ongoing incident.

#### Respond and Remediate

Automate response actions, create Webex War rooms, and generate incident reports

#### Al-Generated Notes

Automatic generation of incident investigation notes of manual actions for XDR audit logs



#### **Explain and Recommend**

Provides recommendations to remediate incidents, thereby reducing response times

#### Interpret Incidents

Ask natural language questions to rapidly understand an ongoing incident

#### Preemptive Detection & Response

Augment attack chain detection in XDR with recommended response actions



Al Assistant for Security Cloud Control



## Gain end-to-end visibility from a single screen



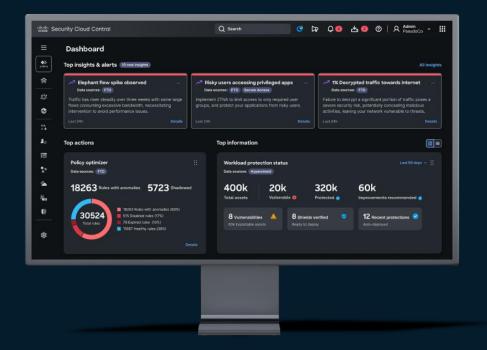
Access comprehensive insights across all security platforms.



Enable prompt issue resolution through a live view of network traffic and security events



Empower informed decisions with operational insights from network data





## First Glimpse of the Unified Al Assistant

#### Single Al Assistant

 One assistant on SCC routes user questions to the appropriate platform.

#### **Skills Integration**

Includes skills from Firewall, Secure Access, and Hypershield Al Assistants

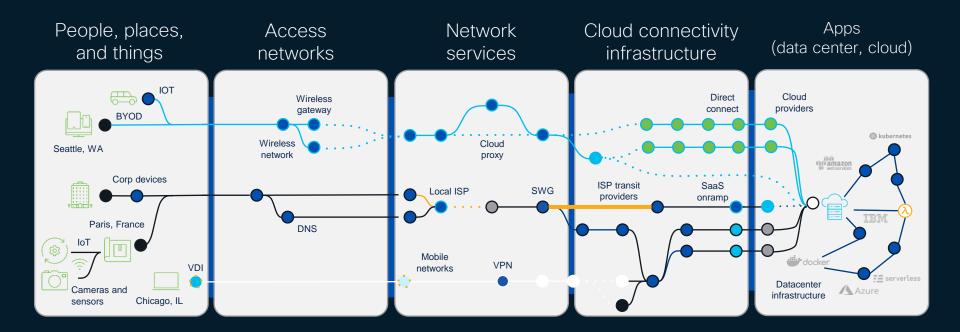
Note: For first phase.



Generative Al for Networking



## Networks Are Complex, Al Simplifies It!





## Cisco Networking's Suite of Al Assistants

#### Meraki



#### **Catalyst Center**



#### **Identity Service Engine**



#### SD-WAN



Cloud-managed networking with security, visibility, and device control.

On-prem network management for automation, policy, security & assurance.

Enforces identity-based access policies, ensuring secure network access and compliance

Optimizes WAN traffic and security across remote sites.



## Cisco Networking Al Assistant Skills

#### **Documentation Summarization**

Summarize key documentation and best practices for quick guidance.

#### Troubleshooting

Diagnose issues and recommend solutions based on network data.

#### Impact Analysis & Optimization

Assess impacts of changes and optimize for efficiency.



#### **Network Visibility**

Monitor real-time and historical network health and performance.

#### Configuration & Migration

Assist with feature setup and migration readiness

#### Support & Case Management

Streamline TAC support and automate case management tasks.



## Al Assistant for Meraki



## Built on an Intelligent cloud platform

**Security and WAN** 

WAN

access

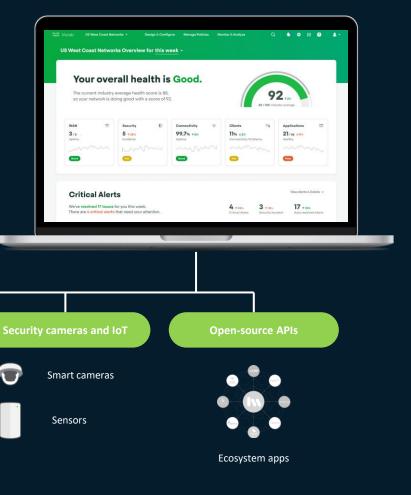
Endpoint

security

Security and SD-

Fixed wireless

94% of customers save time with a single intuitive dashboard





**Switching** 

Wireless



Secure Connect

Network-wide

Assurance New

Security & SD-WAN

Switching

Wireless

Systems Manager

Cameras

Sensors

||| Insight

. Organization

Q Find in Menu







# Al Assistant for Catalyst Center



## Catalyst Center use cases and personas

The 5 use cases



NetOps

Automation and workflows to streamline and simplify complex or repetitive tasks

**AlOps** 

AI/ML to ensure consistent performance and **optimal user experience** 

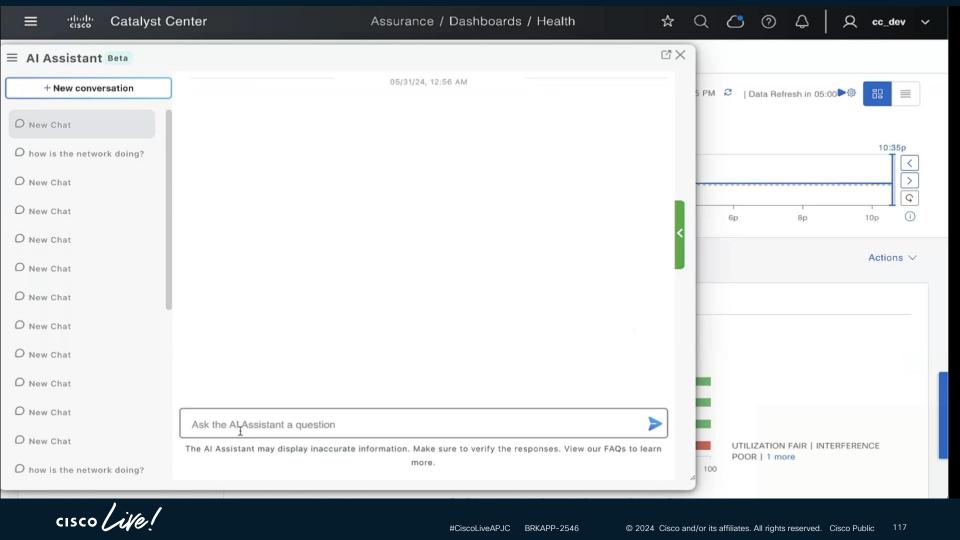
SecOps

Complete zero trust solution for flexible segmentation and policy enforcement including Al-driven device identification

DevOps

APIs combined with playbooks for managing networks at scale and numerous out-of-the-box integrations





# Generative Al for Collaboration





## Cisco Webex's Suite of Al Assistants



Collaboration workspace with chat, file sharing, and video meetings

High-quality video conferencing for virtual meetings and webinars.

Centralized management for Webex services, users, and security

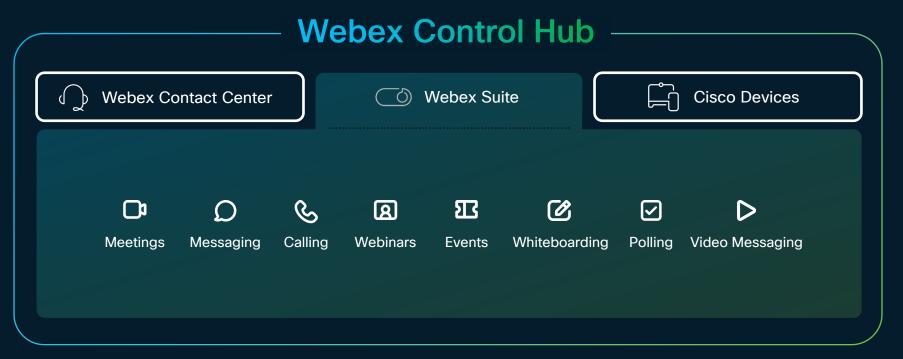


# Al Assistant for Control Hub



### Control Hub

A single-pane-of-glass management





### Cisco Control Hub Al Assistant Skills

#### Setup and Configure

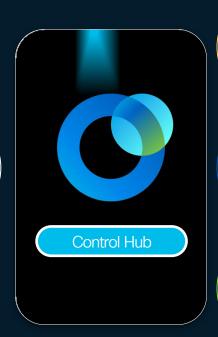
Easily configure settings and manage setups. Ask "How do I...?" or request, "Do it for me."

#### Troubleshooting

Receive proactive notifications of issues, along with root cause analysis.

#### Contextual Q&A with Recall

Get answers with context-based memory for accurate, relevant responses.



#### Analytics and Insights

Query your data for valuable insights and metrics to enhance decision-making.

#### Converse with Reports

Engage with your reports seamlessly for enhanced, conversational insights.

#### Smart Search with Al

Leverage Al-driven search for faster, smarter information retrieval.





#### Hi there.

I'm Cisco Al Assistant. I can guide you through Control Hub and help with admin tasks. I can make mistakes, so consider double-checking my responses. I'll always respect your privacy and won't use your data to train Al models without your permission.

How do I add new users?

What are the benefits of SSO?

How do I enable transcription?

Write a message to the Qisco Al Assistant



Al assistant may provide inaccurate information. Please verify answers and check sources.

Unifying Cisco's Al Assistants across security, networking, collaboration

## Individual Al Assistants Are Integrated Across Cisco



Security

Firewall, Secure Access, XDR, Hypershield, Duo



Networking

Meraki, Catalyst Center, ISE, Catalyst SD-WAN



Observability

Splunk Enterprise Security, Splunk Observability, Splunk Enterprise



Collaboration

Webex Meeting, Webex Teams, Webex Control Hub



## Al Assistants Increase Intra-Product Productivity

Documentation Summarization Troubleshooting & Visibility

Configuration

Recommendation & Optimization



## Cross-Platform Troubleshooting is Complex



## Integrations Are Seldom Comprehensive

- Cisco platforms often only have basic integrations with each other.
- Users must manually correlate data across platforms for troubleshooting.

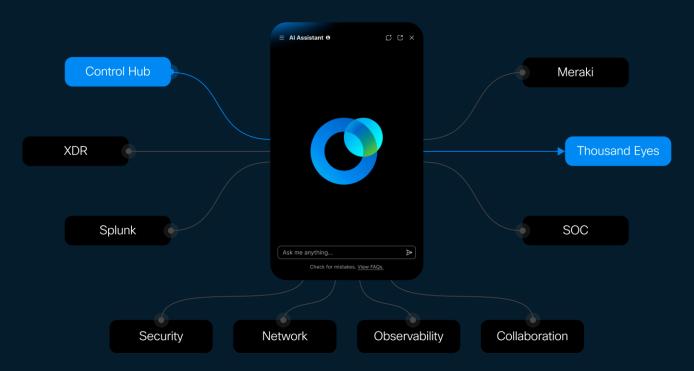


## Troubleshooting Experience Can be Time-consuming

- It can require multiple teams & tickets, hours to days to resolve.
- Example: Poor Webex experience could require Meraki,
   ThousandEyes, & Firewall teams.

## Unifying Cisco Al Assistants

Bringing together "Skills" across Cisco to 10x customer experience





## Unified Al Assistant introduces "Composite" Skills

#### Simple Skills

- Definition: Insights and actions taken for a single platform at a time.
- Example: Checking the client's health on Meraki.



Individual Al Assistant

Skills for One Platform

#### Composite Skills

- Definition: Combines intelligence across multiple products for enriched insights to accelerate RCA action.
- Example: Correlating a poor-performing Meraki client with XDR, Firewall, & ThousandEyes data for cross-domain context.

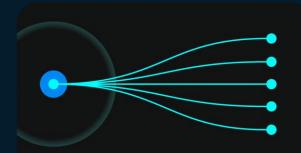


Unified Al Assistant

Skills for Many Platforms

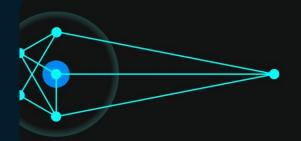


### Benefits of the Unified Al Assistant



#### One assistant, many skills

Each Cisco product enhances the Unified AI Assistant with additional "simple" skills to troubleshoot issues.



#### Compounding value

Combines cross-platform 'simple' skills into 'composite' skills—more Cisco products mean exponentially richer context and smarter recommendations.



#### Accelerated resolution

Troubleshooting is consuming, but the Al Assistant enables RCA in minutes by correlating cross-domain insights!



## Troubleshooting with the Unified Al Assistant

#### Issue Identified



#### **Network Assessed**



#### **WAN Analyzed**



#### Issue Resolved!





Meraki





## Key Composite Skill Use Cases

Multi-Source Docs
Summarization

Cross-Platform Correlation for Threat Investigation

XDR + Any Other Platform



Cross-Platform Correlation for Troubleshooting

Any Platform + Any Other Platform

Cross-Platform Correlation for Threat Mitigation

XDR + Any Other Platform

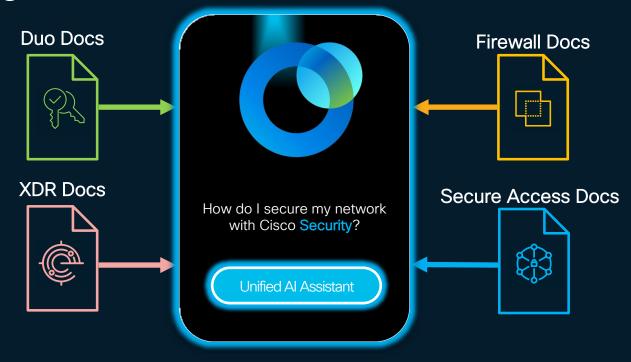


### Connect Cisco's Al Assistant to All Platforms





## Seamless Knowledge Access Across Cisco Products



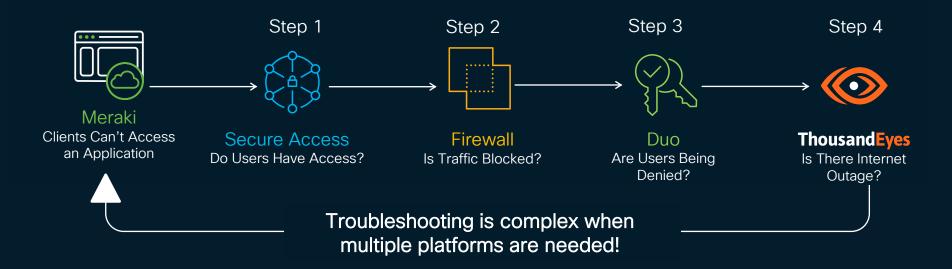


Any Platform + Any Platform



#### Cross-Platform Troubleshooting Pain Points

## Troubleshooting Clients, Devices, & Apps Often Require Data From Multiple Platforms

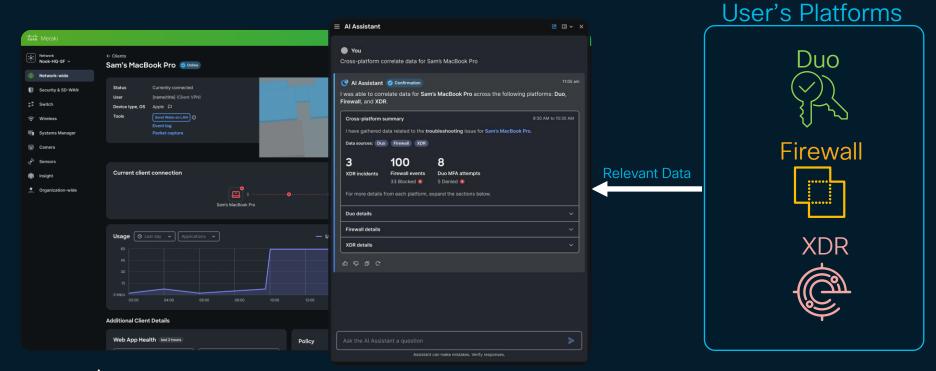




#### **Cross-Platform Troubleshooting Pain Points**

### Accelerate Troubleshooting with Cross-Platform Insights

Client with Performance Issue Discovered



#CiscoLiveAPJC

BRKAPP-2546

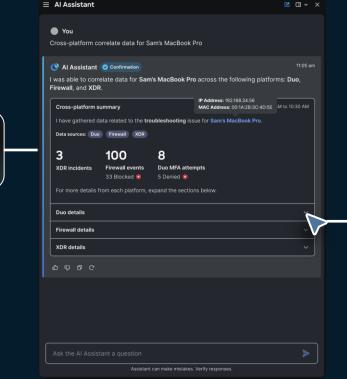


**Cross-Platform Summary** 

Key troubleshooting insights across

all correlated platforms

## Cross-Platform Data Summary for Troubleshooting



#### **Individual Platform Details**

Expandable sections for additional cross-platform

BRKAPP-2546

## Correlate Any Data From Anywhere for Full Context

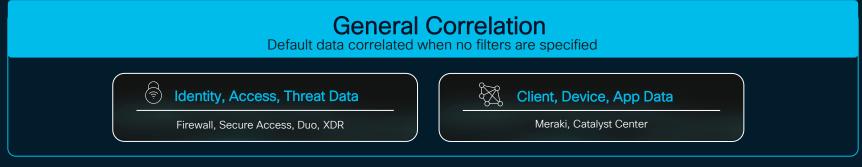
#### All Platforms That Supplement the Issue Being Troubleshooted

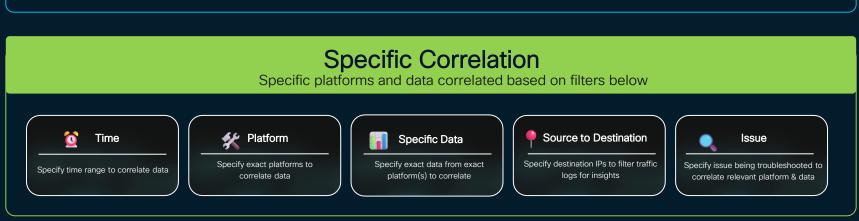


Note: List of platforms above isn't comprehensive to all that is supported.



## Cross-Platform Correlation Types

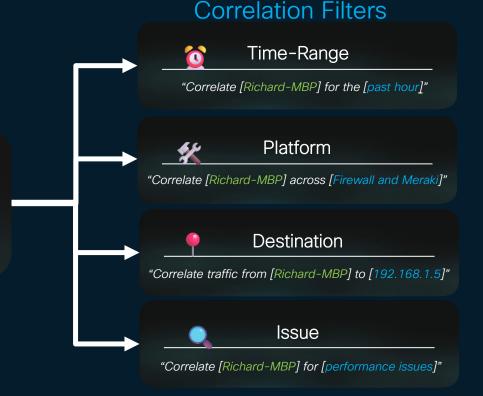




Cross-Platform Correlation Entry Point: Natural Language

#### **General Correlation**

"Cross-platform troubleshoot [client, device, app, user]"

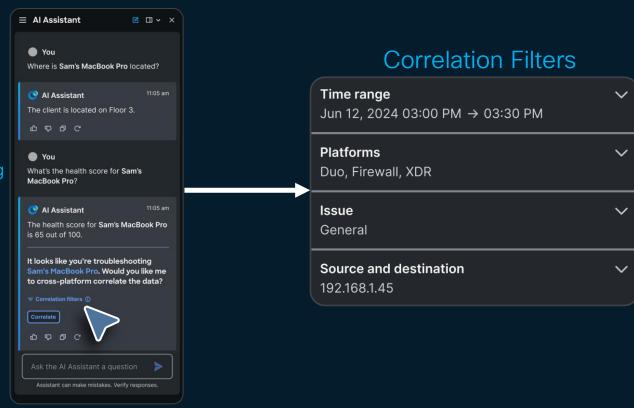




Cross-Platform Correlation Entry Point: Suggested

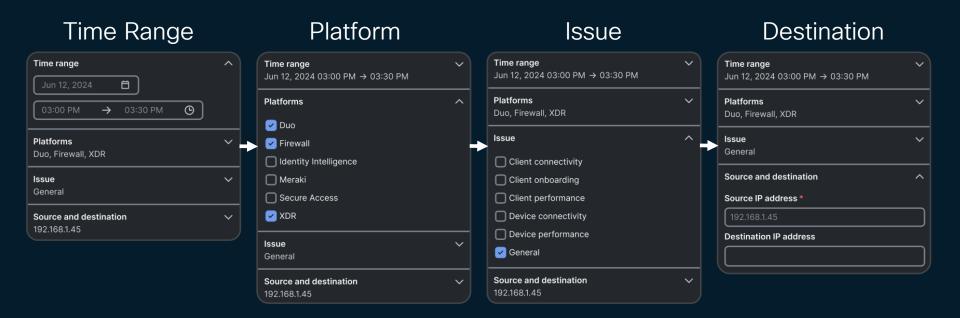
Actions

User Troubleshooting Intent Detected





## Cross-Platform Correlation Entry Point: Suggested Actions

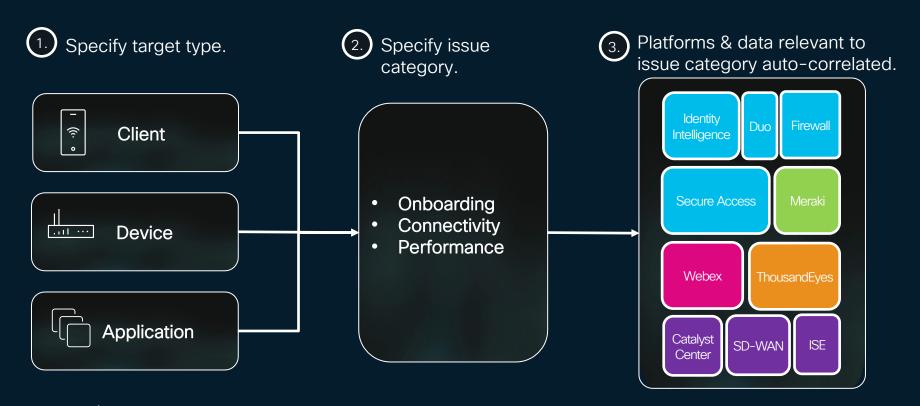




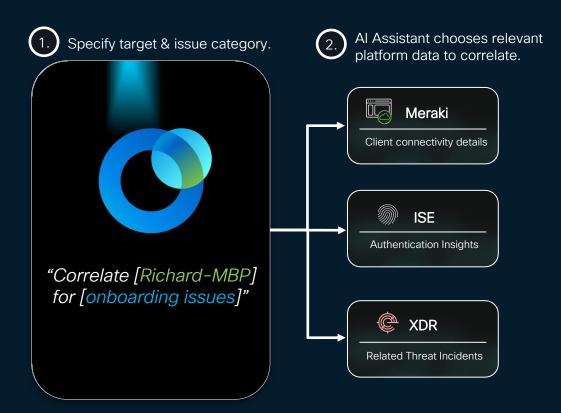
How Source to Destination, Specific Correlation Works



## How Issue-Based, Specific Correlation Works



## Issue-Based, Specific Correlation Example







Benefits of Cross-Platform Correlation for Troubleshooting

#### **Comprehensive Context**

Leverage insights across platforms for full operational clarity





#### Instant Issue Identification

Immediately discern what's causing—and not causing—issues

#### Benefits

#### **Expedited RCA**

Speed up root cause analysis for faster problem resolution





#### Eliminates Context Switching

Seamlessly integrate data without toggling between platforms

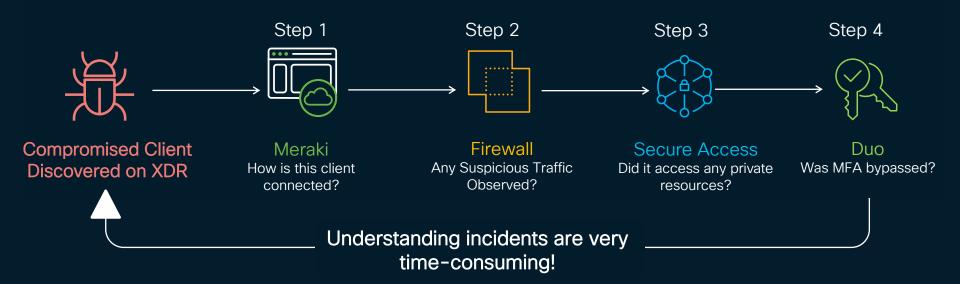


Cross-Platform Correlation for Threat Investigation

XDR + Any Platform

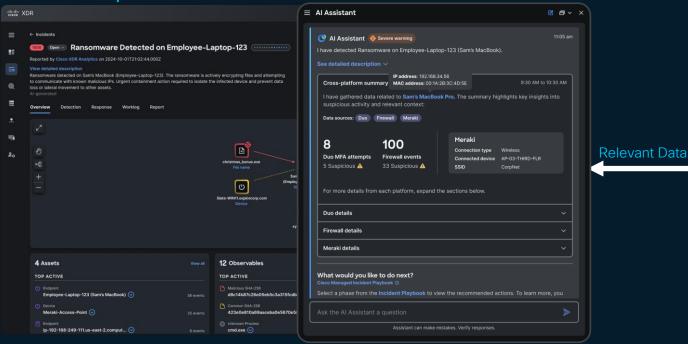


## Manual Threat Investigations Can Be Complex





Accelerate Threat Investigations with Cross-Platform Insights
Compromised Client Discovered on XDR



User's Platforms

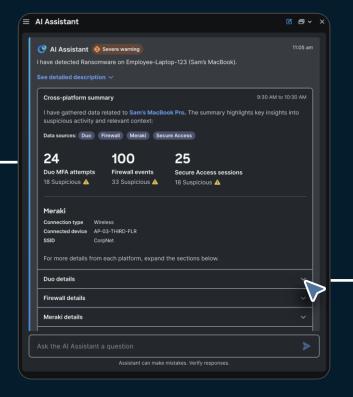




## Cross-Platform Data Summary for Investigations

#### Cross-Platform Summary

Key investigation insights across all correlated platforms



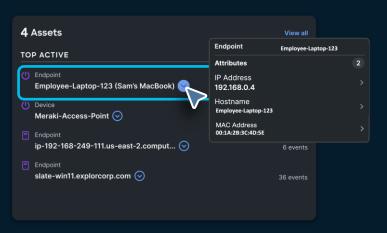
#### Individual Platform Details

Expandable sections for additional cross-platform



## Al Assistant Auto-Correlates Data Using XDR Assets

Upon launch, Al Assistant identifies the top asset & its attributes.

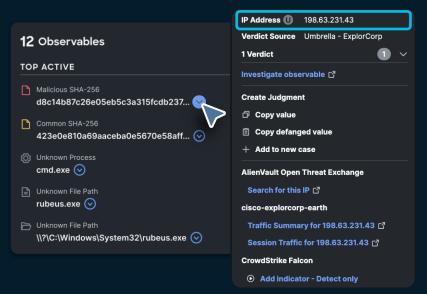


Attributes are used to auto-correlate with relevant platform for data. relevant data.



## Al Assistant Highlights Suspicious Cross-Platform Data

Al Assistant identifies potentially malicious observables that interacted with the asset.



Interactions with observables are flagged as "Suspicious" insights.





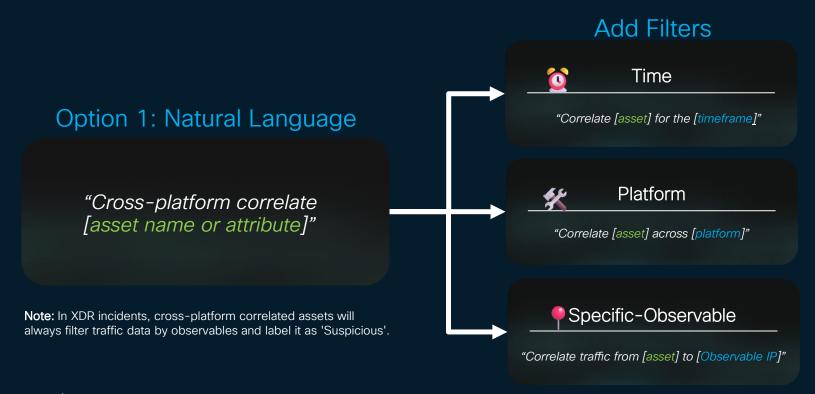
### Relevant Cross-Platform Data is Auto-Correlated



Note: List of platforms above isn't comprehensive to all that is supported.



### Cross-Platform Correlate Additional Assets





## Cross-Platform Correlation for Threat Investigation Demo



### Benefits of Cross-Platform Correlation for Threats

#### **Rapid Threat Confirmation**

Instantly Validate Threats with Full Context





#### Impact Understanding

Identify Critical Resources at Risk

#### Benefits

#### **Expedited Response**

**Enables Swift & Accurate Mitigation** 





#### Impact-Based Prioritization

Focus on High-Impact Threats First



## Al Assistant Simplifies User Experience Across Cisco







Generative Al for Collaboration

## cisco Live!

## Keynote Deep Dives



Experiences
Amplified:
How Al Can Fuel
Better Employee
and Customer
Experiences

Level 1 Room 106



Smart, Secure, Seamless: Transforming Experiences with Next-Generation Networking

Level 2 Room 204



Harness a Bold New Era: Transform Data Centre and Service Provider Connectivity

Level 2 Room 203

### Wednesday 10:30am -11:30am



Securing User to Application and Everything in Between

Level 2 Melbourne Room 2



Unlocking
Digital Resilience
through Unified
Observability

The HUB Centre Stage

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## Thank you



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