Next Gen Internal Audit Analytics: Continuous Monitoring & Predictive Analytics



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Speaker Bios



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Polling Question One

Question: How many session participants have ventured into continuous monitoring?

Question: Do you plan to integrate contintuous monitoring analytics into your 2024 audit plan?



Learning Objectives

- 1. Provide a **roadmap of the internal audit analytics transformation** from historical analysis to risk sensing predictive analytics for attendees to apply within their organizations.
- 2. Offer **practical insights into key stages of the analytics lifecycle** while sharing actionable tips, strategies, and methodologies to enhance the effectiveness of internal audit analytics.
- 3. Showcase **real-world examples illustrating successful applications of** <u>*historical*</u> **analysis and** <u>*predictive*</u> **analytics** to demonstrate the impact on decision-making processes for internal audit and the business.
- 4. Explore the historic and predictive analytics **insights that arise from tapping into the collective intelligence of today's participants** to effectively leverage the diverse expertise within our field.



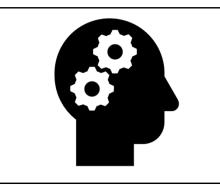
Establishing Foundations: "Crafting Vision and Purpose"

An analytics function is more than just the technical scripting and data analysis......it is important to establish 'why' and 'what' for your existence

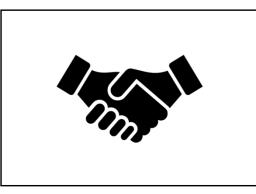
Cultivate a vision for IA analytics that emphasizes:



Cross-Functional Collaborations



Position as Trusted Advisors to Business



Aligning Vision with Business Goals

Then, establish <u>purpose</u> through a mission and corresponding objectives



Establishing Foundations: "Crafting Vision and Purpose"

Internal Audit Automation Program Mission & Objectives

Mission Statement

Internal Audit aims to further leverage technology in support of auditing tactical areas of the business to provide smarter and more comprehensive audit results while freeing resources for strategic and enterprise-risk focused auditing.

"WHY"

Key Program Objectives

Efficiency

- Engagement of audit in key tactical areas with increased frequency/ breadth (full population) and with limited impact to resources
- Free up resources from tactical audits to more strategic & top-risk areas

"WHAT"

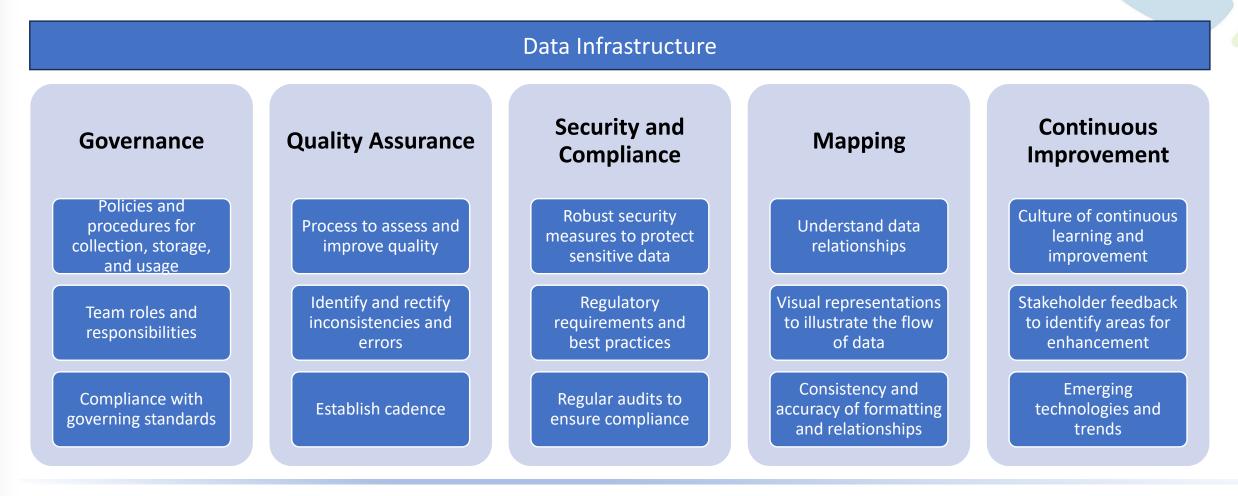
Value

- Engage real-time risk evaluation for • specific areas to drive business conversation and/or future audits
- Provide for third-line of defense based • on historic activity
- Audit full populations of data and • smarter sample selections
- Shift audit to the **predictive space** •



Establishing Foundations: "Technical Proficiency in Data Management"

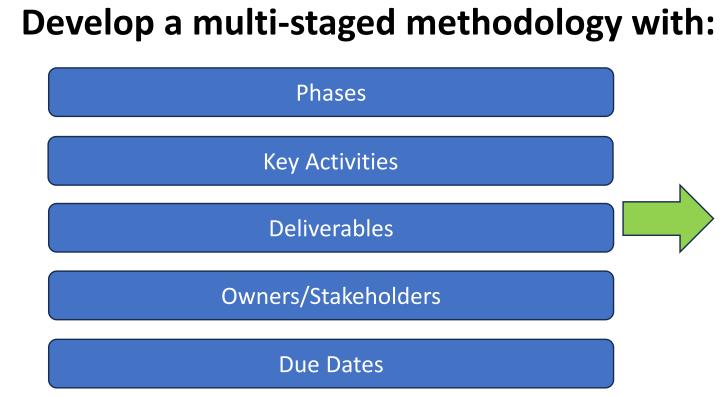
Investing in effective data management practices promotes reliability and informs decision making.





Developing Organizational Excellence: "How We Will Execute"

Long term, sustainable results depend on having a methodology that is **transparent, all-encompassing and repeatable**

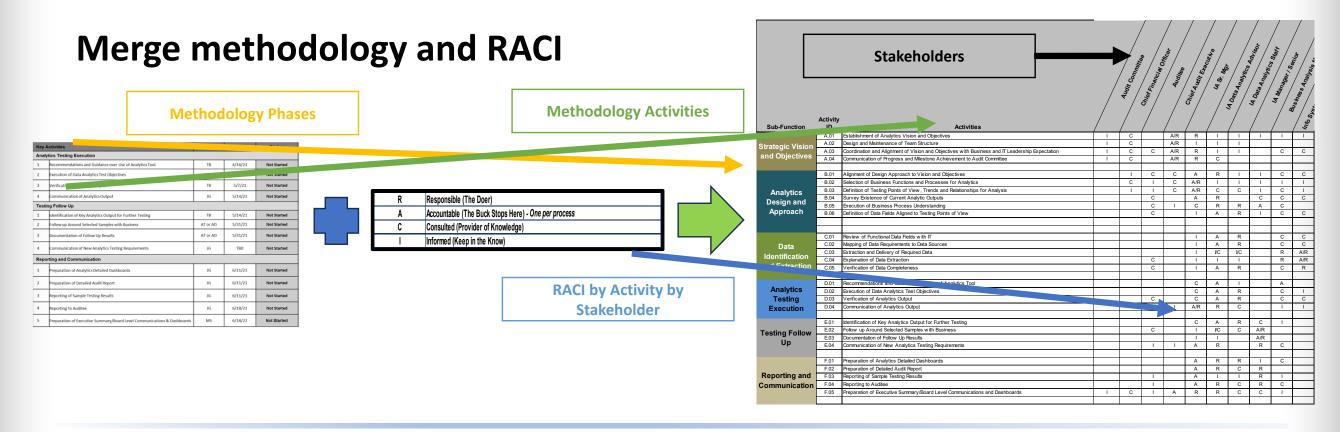


| Kov | Activities | Owner | Due Date | Status |
|------|--|----------|----------|-------------|
| | lytics Testing Execution | Owner | Due Date | Status |
| 1 | Recommendations and Guidance over Use of Analytics Tool | ТВ | 4/16/21 | Not Started |
| 2 | Execution of Data Analytics Test Objectives | ТВ | 4/30/21 | Not Started |
| 3 | Verification of Analytics Output | ТВ | 5/7/21 | Not Started |
| 4 | Communication of Analytics Output | JG | 5/14/21 | Not Started |
| Test | ing Follow Up | | l | |
| 1 | Identification of Key Analytics Output for Further Testing | ТВ | 5/14/21 | Not Started |
| 2 | Follow up Around Selected Samples with Business | AT or AD | 5/31/21 | Not Started |
| 3 | Documentation of Follow Up Results | AT or AD | 5/31/21 | Not Started |
| 4 | Communication of New Analytics Testing Requirements | JG | TBD | Not Started |
| Rep | orting and Communication | | | |
| 1 | Preparation of Analytics Detailed Dashboards | JG | 6/11/21 | Not Started |
| 2 | Preparation of Detailed Audit Report | JG | 6/11/21 | Not Started |
| 3 | Reporting of Sample Testing Results | JG | 6/11/21 | Not Started |
| 4 | Reporting to Auditee | JG | 6/18/21 | Not Started |
| 5 | Preparation of Executive Summary/Board Level Communications & Dashboards | MS | 6/18/21 | Not Started |



Developing Organizational Excellence: "Who Does What"

Realization of mission and objectives is dependent on identifying, defining and communicating stakeholder engagement and responsibility





Developing Organizational Excellence: "Who Does What"

| | | | Audit Committee | Chief Financial Officer | Auditee | Chief Audit Executive | IA Sr. Mgr | IA Data Analytics Advisor | IA Data Analytics Staff | IA Manager / Senior | Business Analysis Manager | Info Systems Financial Systems Leade |
|-------------------------|---------------------|---|--------------------|-------------------------------|---------------------------------------|--|---------------|---------------------------------|-------------------------------|---------------------------------------|---------------------------------|--|
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| Sub-Function | Activity ID A.01 | Activities Establishment of Analytics Vision and Objectives | 1 | с | | A/R | R | 1 | 1 | Í | 1 | 1 |
| | A.02 | Design and Maintenance of Team Structure | | c | | A/R | 1 | | | | | + <u>·</u> |
| Strategic Vision and | A.02 | Coordination and Alignment of Vision and Objectives with Business and IT Leadership Expectation | | c | с | A/R | R | | | | с | c |
| Objectives | A.04 | Communication of Progress and Milestone Achievement to Audit Committee | | c | | A/R | R | c | | | - | |
| | A.04 | Communication of Progress and Innestone Admevement to Audit Committee | | 0 | | ~~ | N | 0 | | | | + |
| | B.01 | Alignment of Design Approach to Vision and Objectives | | 1 | с | с | A | R | 1 | 1 | с | c |
| | B.01 B.02 | Selection of Business Functions and Processes for Analytics | + | c | 1 | c | A/R | 1 | | 1 | 1 | 1 |
| | B.02 B.03 | Definition of Testing Points of View, Trends and Relationships for Analysis | - | 1 | | c | A/R | c | c | | c | · · |
| Analytics Design and | B.04 | Survey Existence of Current AnalyticOutputs | | | i i i i i i i i i i i i i i i i i i i | , in the second se | | | | i i i i i i i i i i i i i i i i i i i | | |
| Approach | B.05 | Execution of Business Process Understanding | | | c | 1 | c | R | R | A | c | |
| Approach | B.05 | Definition of Data Fields Aligned to Testing Points of View | | | c | 1 | 1 | A | R | <u> </u> | c | с |
| | | | | | | | | | | | - | |
| | C.01 | Review of Functional Data Fields with IT | | | | | Ι | А | R | | С | с |
| | C.02 | Mapping of Data Requirements to Data Sources | | | | | I | A | R | | С | С |
| Data Identification and | C.03 | Extraction and Delivery of Required Data | | | | | I | I/C | I/C | | R | A/R |
| Extraction | C.04 | Explanation of Data Extraction | | | С | | Ι | I | I | | R | A/R |
| | C.05 | Verification of Data Completeness | | | С | | I | A | R | | С | R |
| | D.01 | Recommendations and Guidance over Use of Analytics Tool | | | | | С | A | 1 | | А | - |
| | D.02 | Execution of Data Analytics Test Objectives | | | | | С | A | R | | С | 1 |
| Analytics Testing | D.03 | Verification of Analytics Output | | | с | | С | A | R | | С | с |
| Execution | D.04 | Communication of Analytics Output | | | | 1 | A/R | R | с | | 1 | 1 |
| | | | | | | | | | | | | - |
| | E.01 | Identification of Key Analytics Output for Further Testing | | | | | С | A | R | С | I | - |
| | E.02 | Follow up Around Selected Samples with Business | | | С | | Ι | I/C | С | A/R | | - |
| Testing Follow Up | E.03 | Documentation of Follow Up Results | | | | | I | I | | A/R | | - |
| g. enen ep | E.04 | Communication of New Analytics Testing Requirements | | | I | I | A | R | | R | С | |
| | | | | | | | | | | | | - |
| | F.01 | Preparation of Analytics Detailed Dashboards | | | | | А | R | R | I | С | |
| | F.02 | Preparation of Detailed Audit Report | | | | | А | R | с | R | | |
| Reporting and | F.03 | Reporting of Sample Testing Results | | | I | | А | I | I | R | I | 1 |
| Communication | F.04 | Reporting to Auditee | | | I | | А | R | С | R | С | 1 |
| | F.05 | Preparation of Executive Summary/Board Level Communications and Dashboards | 1 | С | I | A | R | R | С | С | I | 1 |
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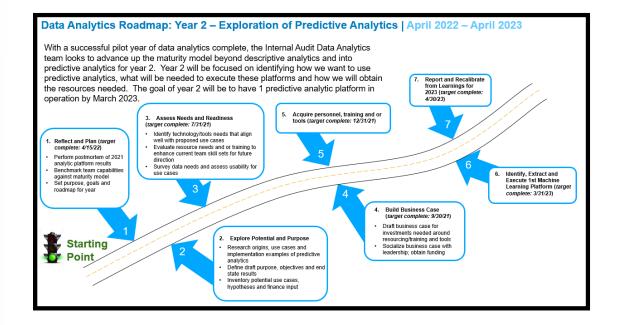


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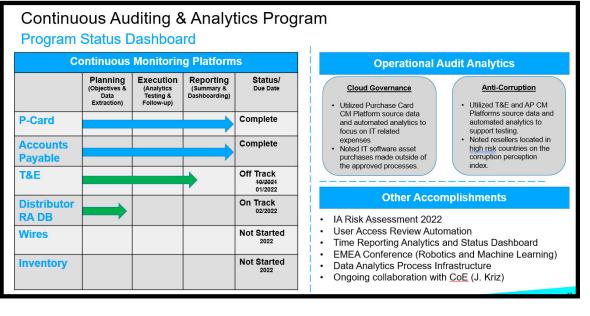
Developing Organizational Excellence: "Set Milestones; Keep Accountability"

Ensure success by charting an incremental path and schedule regular touchpoints with management (i.e. CFO, CAE, VPs, etc.) to report status of program progress

Timebound tasks on a roadmap



Dashboard status and broadcast accomplishments





Embed Continuous Improvement: "Lookback, Assess and Adjust"

Drive year over year progress through postmortem analysis; have a vision for your analytics' program maturity and set goals and action plans accordingly

Leverage a maturity model to help frame analytics program current state vs. desired state on multiple dimensions

Internal Audit Analytics Maturity Model

Use the Maturity Model below to assess Internal Audit Analytics performance and identify priorities for improvement opportunities. Overall, ZIA data analytics team scores at or slightly above industry practices with strong capabilities in strategy, structure, people and technology with opportunities to improve in risk assessment, audit planning and continuous monitoring.

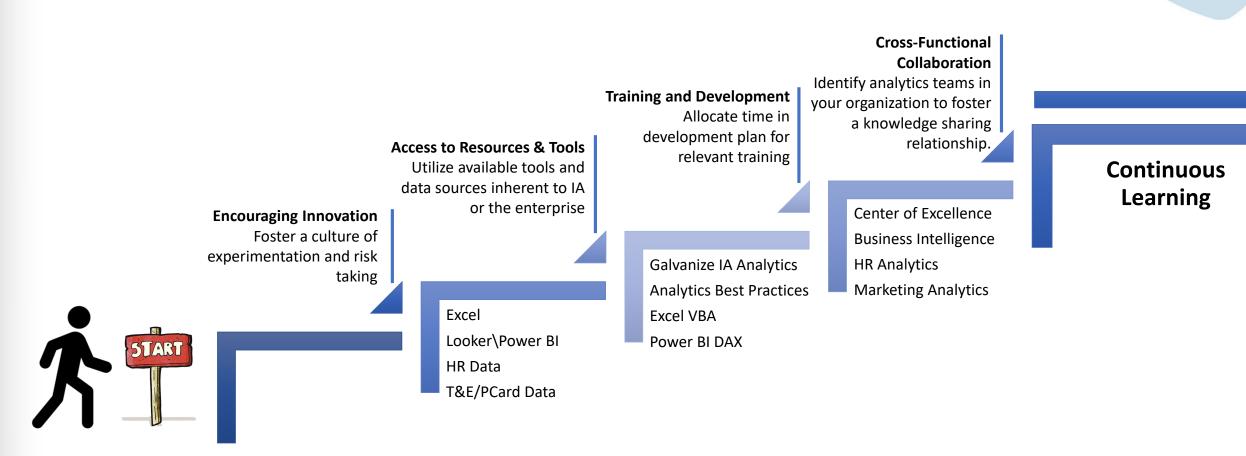
| Process inefficiency & duplication Business criticality drives process evolution Culality.control processes Process inefficiency and duplication Minimal use of workflow and mapping tools Standardized process flows and leverage of work mapping tools Risk Assessment Ratings defined through interviews Limited on-going analysis Metrics from business leveraged Some on-going analysis Independent risk metrics Orgoing risk analysis Audit Planning Scope defined through interviews Data not leveraged to make scope decisions Some data utilized to understand business Data used for all applicable areas of business Data used or utilized ashboards Controls Testing Analytics use driven by requests Most analytics used for reconciliations No historical testing libraries Analytics planned quarter by quarter Mustice all analytics aread of multiple enalytics techniques utilized Some historical workpapers Independent analysis and metrics Shared dashboards Continuous Monitoring defined through interviews Limited on -going analysis Metrics from business leveraged Some on-going analysis Independent analysis and metrics Shared dashboards | | Leading Practices | | At Industry Practices | Below Industry Practices | |
|--|---------------------------------|--|-------|--|---|--------------------------|
| Organizational Structure Unclear roles and responsibilities Small span of control Roles and responsibilities understood but not documented Integrated coordination & management Roles and responsibilities clearly defined People Narrow skill base Limited development, ad-hoc learning Limited analytics understanding Narrow skill base Functional training and development Uneven expertise across functional audit areas Extensive excel-based Access ACL analytics Strong analytics mindset Staff rolations & leadership programs Deep technical skills & understanding Process Limited integration / no data governance Manual data collection processes Extensive excel-based Access ACL analytics Multiple and redundant systems and data feeds Limited advanced tools and information sharing Minimal data governance Automatic data capture Strong data governance Analytic specific tools and data warehouses Process Fragmented & non-standardized Process inefficiency & duplication Business criticality drives process evolution Business criticality drives process evolution Optimized by function and audit area Optimized by roution and augit area Process inefficiency & duplication Business criticality drives and yaugity audit universe only qualitatively analyzed Metrics from business leveraged Some on-going analysis Data ware to guide some discussion Some data utilized to understand business Data used for reconciliations No historical testing libraries Data used for al applicable areas of business Outside metrics leveraged for marke scope decisions Controls Testing Analytics used for reconciliations No historical testing libraries An | ♦ IA Assessed Maturity Level | Metrics in place to track performance |] < | Moderate anticipation of analytics needs | Reactive to business needs | Strategy |
| Technology Limited development, ad-hoc learning Limited analytics understanding Functional training and development Uneven expertise across functional audit areas Staff rotations & leadership programs Deep technical skills & understanding Technology Limited integration / no data governance Manual data collection processes Extensive excel-based Access ACL analytics Multiple and redundant systems and data feeds Limited data acquire Automatic data capture Process Fragmented & non-standardized Process inefficiency and dupication Business critically drives process evolution Optimized by function and audit area Ouality control processes inefficiency and dupication Minimal use of vondfwor and mapping tools End-to-end process ownership and documentatio Standardzed process flows and leverage of work mapping tools Risk Assessment Ratings defined through interviews Audit universe only qualitatively analyzed Optimized by function and audit area Oraging analysis Audit universe only qualitatively analyzed Independent risk metrics Ongoing risk analysis Some on-poing analysis Audit universe only qualitatively analyzed Data used to profit and cost centers Data used for all applicable areas of business Data used to guide some discussions Excel and ACL modeling Data used for all applicable areas of business Data used to makers data business Data used to reconciliations No historical testing libraries Analytics telenders data business Data used to guide some discussion so part of annual planning Some on-poing analysis Analytics discussion as part of annual planning Shared dashboards Independent analysis and metrics Shared dashboards <th>Current IA Skill Set</th> <th>Integrated coordination & management</th> <th>it 🚽</th> <th>Roles and responsibilities understood but not documented</th> <th>Unclear roles and responsibilities</th> <th></th> | Current IA Skill Set | Integrated coordination & management | it 🚽 | Roles and responsibilities understood but not documented | Unclear roles and responsibilities | |
| Manual data collection processes Extensive excel-based Access ACL analytics Limited data acuigation automation Minimal data governance Strong data governance Analytic specific tools and data warehouses Process Fragmented & non-standardized Process inefficiency & duplication Business criticality drives process evolution Optimized by function and audit area Quality control process gess Process inefficiency & duplication Minimal use of vorkflow and mapping tools End-to-end process ownership and documentatic Standardized process flows and leverage of work mapping tools Risk Assessment Ratings defined through interviews Limited on-going analysis Audit universe only qualitatively analyzed Metrics from business leveraged Some on-going analysis Quives under to profit and cost centers Independent risk metrics Ongoing risk analysis Shared dashboards Audit Planning Scope defined through interviews Data not leveraged to make scope decisions Data used to guide some discussions Data used to reconciliations Notistoric at leveraged for scoring Excel and ACL modeling Data used for all applicable areas of business Data used to guide analytics used for reconciliations No historical testing libraries Analytics planned quarter by quarter Multipe analytics techniques utilized Some historical workpapers Analytics discussion as part of annual planning Shared dashboards Contrinuous Monitoring effined through interviews No historical testing libraries Metrics from business leveraged Some on-going analysis Independent analysis and metrics Shared dashboards | | Staff rotations & leadership programs | eas 🔷 | Functional training and development | Limited development; ad-hoc learning | People |
| Process inefficiency & duplication Quality control processes Standardzeid process forws and leverage of work mapping tools Risk Assessment Ratings defined through interviews Metrics from business leveraged Independent risk metrics Audit Planning Scope defined through interviews Some data utilized to understand business Data used for all applicable areas of business Controls Testing Analytics use driven by requests Analytics planned quarter by quarter Analytics discussion as part of annual planning Continuous Monitoring defined through interviews Analytics from business leveraged Some data utilized to understand business Data used for all applicable areas of business Some data utilized to understand business Data used for all applicable areas of business Controls Testing Analytics used for reconciliations Analytics planned quarter by quarter Analytics discussion as part of annual planning Spread of multiple techniques Monitoring defined through interviews Some no-going analysis Some historical workpapers Shared dashboards | | Strong data governance | ···· | Limited data acquisition automation Limited advanced tools and information sharin | Manual data collection processes | Technology |
| Audit Planning Limited on-going analysis Audit universe only qualitatively analyzed Some on-going analysis Universe linked to profit and cost certers Ongoing risk analysis Shared dashboards Audit Planning Scope defined through interviews Data not leveraged to make scope decisions Some data utilized to understand business Data used to guide some discussions Data used for all applicable areas of business Outside metrics leveraged for scoring Controls Testing Analytics used for reconciliations No historical testing libraries Analytics planned quarter by quarter Multiple analytics techniques utilized Some historical workpapers Analytics discussion as part of annual planning Spread of multiple techniques Continuous Monitoring defined through interviews [Limited on-going analysis Metrics from business leveraged Some n-going analysis Independent analysis and metrics Shared dashboards | | End-to-end process ownership and documentation Standardized process flows and leverage of workflow and mapping tools | | Quality control processes Process inefficiency and duplication | Process inefficiency & duplication | Process |
| Controls Testing Maniptics used for make scope decisions Data used to guide some discussions Outside metrics leveraged for scoring Controls Testing Analytics use driven by requests Most analytics used for reconciliations No historical testing libraries Analytics planned quarter by quarter Multiple analytics techniques utilized Some historical workpapers Analytics discussion as part of annual planning Spread of multiple techniques Continuous Monitoring defined through interviews Uninted on-poing analysis Metrics from business leveraged Some n-poing analysis Independent analysis and metrics Shared dashboards | | Ongoing risk analysis | | Some on-going analysis | Limited on-going analysis | Risk Assessment |
| Continuous Monitoring defined through interviews Metrics from business leveraged Independent analysis and metrics | | Outside metrics leveraged for scoring | | Data used to guide some discussions | | Audit Planning |
| Monitoring Limited on-going analysis Some on-going analysis Shared dashboards | | | | Multiple analytics techniques utilized | Most analytics used for reconciliations | Controls Testing |
| Manual Processes Excel trending of metrics Use of external data | | | | | | Continuous Monitoring |

* "Internal Audit Analytics Maturity Model", 2022, Author Unknown



Establishing Foundations: "Upscaling Team Members"

Unlocking the potential of your team's analytical skills is within reach through...





Polling Question Two

Question: How many session participants have ventured into *predictive* analytics?



Strategic Analytics Partnerships

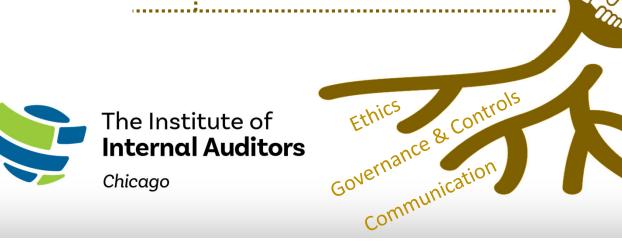
Becoming a trusted advisor through internal audit continuous monitoring analytics involves demonstrating expertise, providing valuable insights, and building strong relationships with stakeholders.

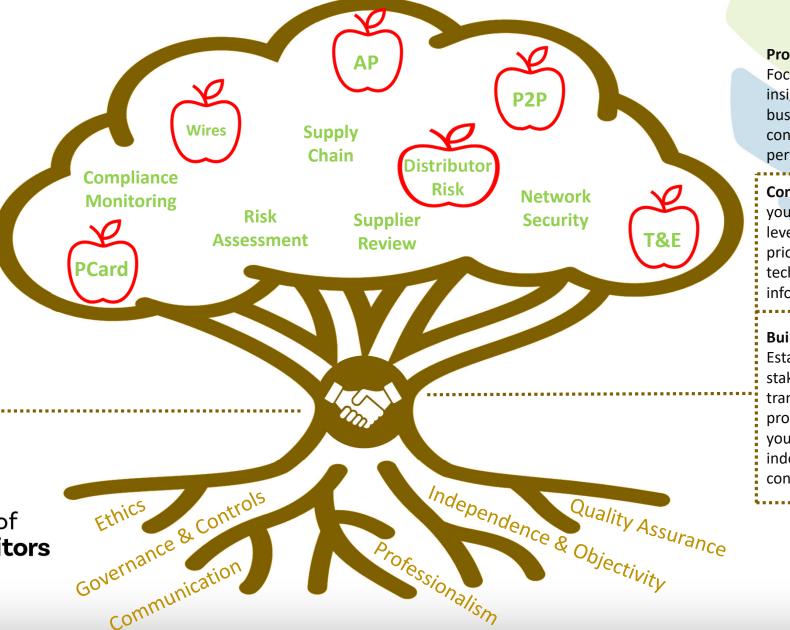
Develop Expertise: Stay updated with industry trends, best practices, and emerging tools in data analytics and audit methodologies.

.........

Understand Business Objectives: Understand the key risks, challenges, and opportunities faced by the business to tailor your analytics initiatives accordingly.

...... Collaborate with Stakeholders: Engage stakeholders in the audit process, solicit their input and feedback, and address their concerns proactively.





Provide Actionable Insights: Focus on delivering value-added insights that help improve business processes, enhance controls, and optimize performance.

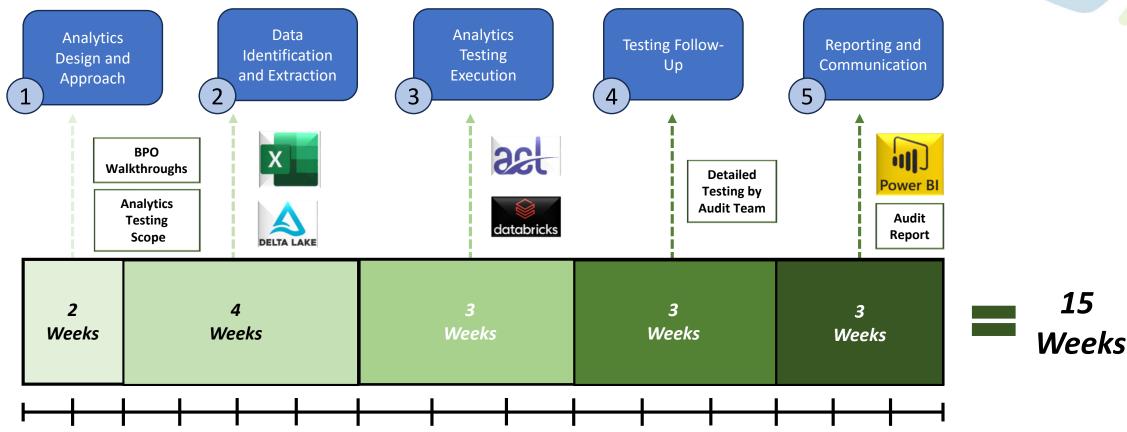
Communicate Effectively: Tailor your message to the audience's level of understanding and priorities. Use data visualization techniques to convey complex information effectively.

Build Trust and Credibility:

Establish trust and credibility with stakeholders through transparency, integrity, and professionalism. Demonstrate your commitment to objectivity, independence, and ethical conduct in all audit activities.

Becoming a Trusted Advisor through Continuous Monitoring: "Continuous Monitoring Platform Lifecycle"

Building the bedrock of our continuous monitoring program in support of internal audit and the business.



Our Methodology



Becoming a Trusted Advisor through Continuous Monitoring: "Continuous Monitoring Platform Scoping: Purchase Card Program"

Policy Compliance

- 1. Spend over Credit Limits
- 2. Split Transactions
- 3. Missed Capitalization
- 4. Significant Credit Limit Changes
- 5. Inclusion & Diversity Statistics

Fraud, Waste, and Abuse

- 6. Higher Risk Merchant Classification Codes (MCCs)
- 7. Expenses by Terminated Employees
- 8. Expenses by class of transaction, function
- 9. Expenses by merchant, MCC (possible lost volume discount opportunity)
- 10. Duplicate Transactions (including w/ T&E)
- 11. Weekend / Holiday Transactions,
- 12. Unusual P-Card Expense Locations
- 13. Average Monthly Spend Compared to Credit Limit
- 14. Dormant and/or Seldom Used Cards
- 6. Significant Spend Increases (MoM and QoQ)
- 7. Ghost Cards

Cardholder Administration

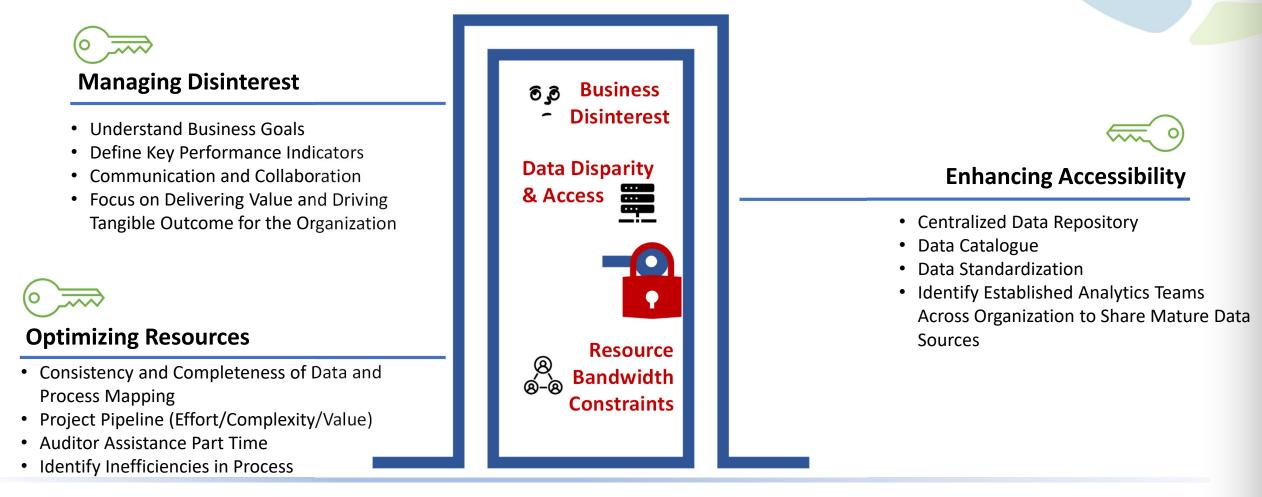
- 15. Terminated Employees w/ Active PCards
- 16. Credit Limits Not in Line w/ Employee Role
- 17. Duplicate Cards, Multiple Cards per Employee
- 18. Employees Who Also Hold a T&E card.
- 19. Shared Cards
- 20. Missed PCard Rebate Opportunities

*Beneficial Analytics per the Business Group



Becoming a Trusted Advisor through Continuous Monitoring: "Navigating Limitations"

Your team's adaptability and agility hold the keys to overcoming inevitable limitations.





Becoming a Trusted Advisor through Continuous Monitoring: "Key Learnings in Continuous Improvement"

Annual post-mortems of methodology are vital for insightful forward planning, ensuring alignment with the evolving maturity of capabilities, and optimizing future strategies.

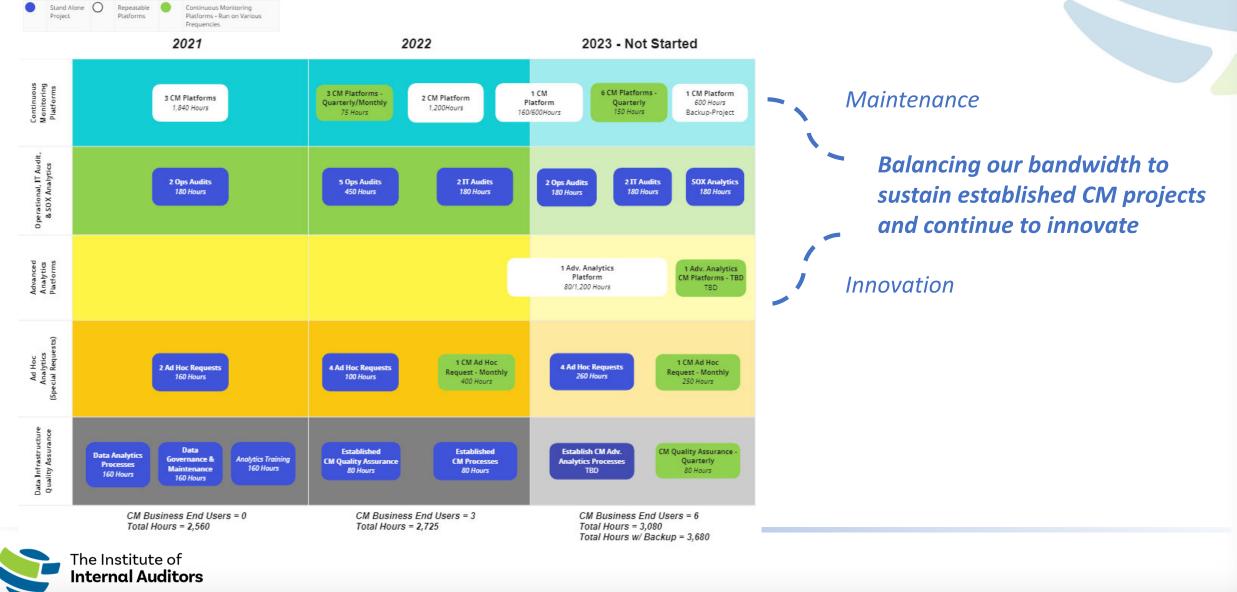


The Institute of I**nternal Auditors**

Chicaao

The Evolution of Our Methodology

Becoming a Trusted Advisor through Continuous Monitoring: "Key Learnings in Continuous Improvement"



Chicago

Shift auditing perspectives from being artifact and historical driven to forward looking insights that help a company anticipate and plan for risk

What is Grey Market?

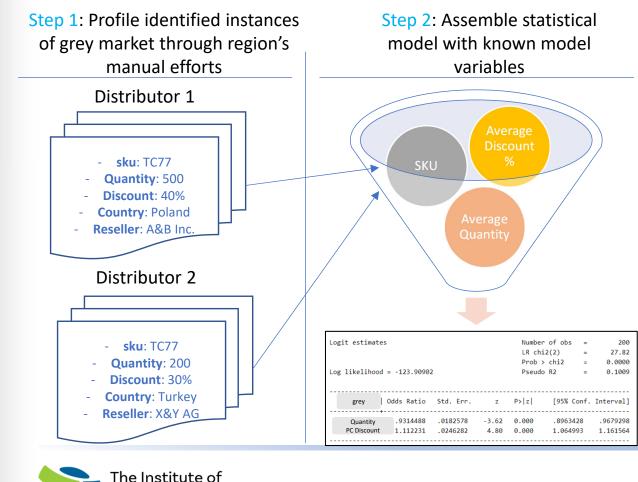
Authentic product sold out of a low-cost territory to a high-cost territory thereby disrupting distributor business and profits What is current solution?

Manual "secretshopper" purchasing that detects (and reacts to) grey market product <u>already</u> in the market. What are advantages of predictive analytics?

- 1. Shift the control from being manual and detective to preventive and automated.
- 2. Insight is now forward looking instead of historical.
- Tedious manual efforts are eliminated; investigations can be targeted based on analytic output.



Areas that lend well to predictive analytics have a) <u>suspected multiple variables impacting</u> <u>another variable</u>, b) <u>adequate data capture</u>, and c) <u>governance around data input</u>



Internal Auditors

Chicago

Step 3: Load quarterly sales out and PC data, etc.; execute statistical model

Point of Sale (PoS) Report

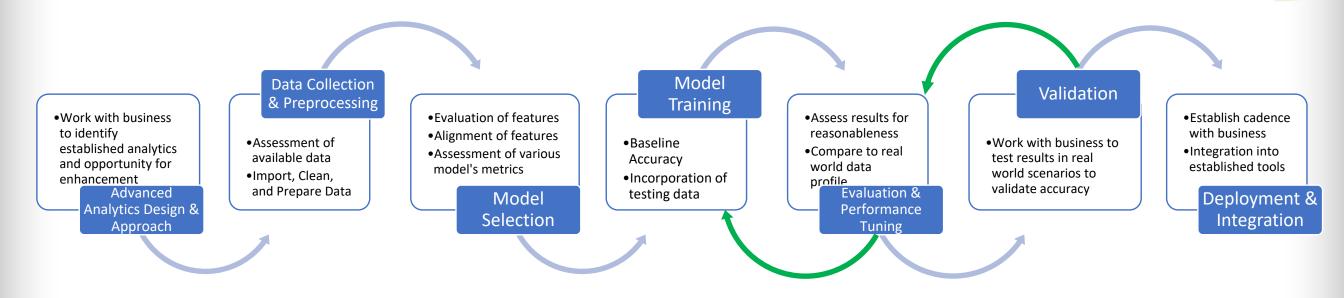
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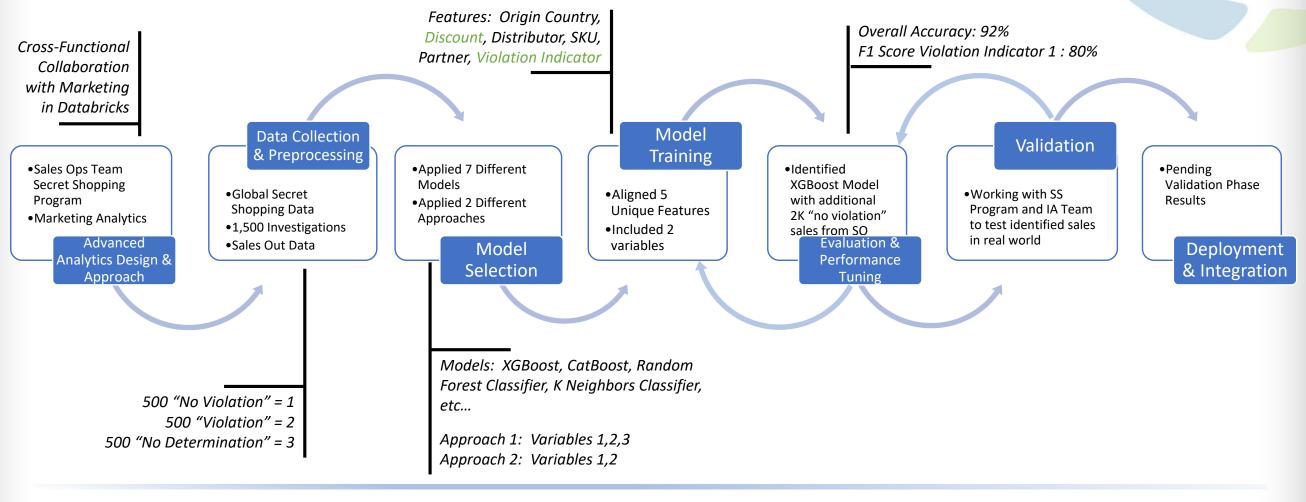


The Evolution of Our Methodology Continues to Adapt : Machine Learning Lifecycle



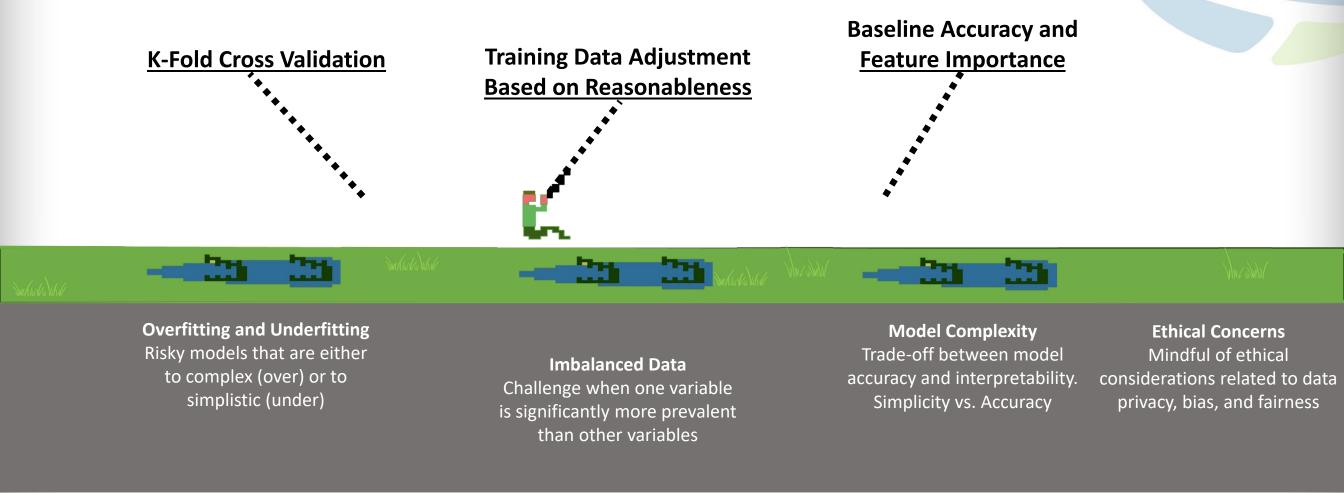


The Evolution of Our Methodology Continues to Adapt : Grey Market Predictive Platform Lifecycle





Lessons Learned on How to Avoid Pitfalls





Introduction to Predictive Analytics: "Grey Market Sales Prediction: Distributor Risk Assessment"

- Near Term: Support Zebra's Internal Audit team in selection of distributor audits and subsequent sample transactions for testing
- Medium-Term: Support Zebra's secret shopping team by enabling smarter sample selections
- Long-Term: Enable Zebra sales team to identify grey market transactions prior to close of sale/shipment Shift efforts from detective to preventive



Polling Question Three

Question:

Within different industries, what examples would lend themselves well to predictive analytics?

Ideas for Inspiration:

Telecommunications: "What predictive analytics solutions could enhance network performance and customer satisfaction?

How can we predict network outages or anticipate bandwidth requirements?" **Retail**: "In what ways can predictive analytics enhance customer experience and increase sales?

How can we predict consumer trends or forecast demand more accurately?"

Manufacturing: "How can predictive analytics optimize production processes and minimize downtime?

What predictive maintenance models could help anticipate equipment failures?"

Transportation: "What predictive analytics solutions could enhance logistics and supply chain efficiency?

How can we predict traffic patterns or optimize route planning?"

Healthcare: "How can predictive analytics improve patient outcomes and optimize healthcare delivery?

What predictive models could help in early disease detection or personalized treatment plans?"

Finance: "How might predictive analytics mitigate financial risks and improve investment strategies?

What models could help detect fraudulent activities or predict market fluctuations?"



Key Takeaways

1. Establishing Foundations

- Craft a clear vision for IA analytics that aligns with business goals, emphasizing cross-functional collaborations and the role of audit as trusted advisors.
- Development of a repeatable and transparent methodology that encompasses data management and enhancement of auditor capabilities is achievable.

2. Becoming a Trusted Advisor Through Continuous Monitoring

- Leverage analytics to establish meaningful business partnerships in support of IA and the business.
- Recognize and address challenges proactively and allow your methodology to evolve.
- Maintain a balance between sustained upkeep of established continuous monitoring projects and consistent innovation.

3. Introduction to Predictive Analytics

- Continued advancement of auditor capabilities and collaboration with seasoned analytics teams within your organization is essential.
- Training data should reflect the real world, align features, create variables, balance your model between simplicity and accuracy.



Section Break

