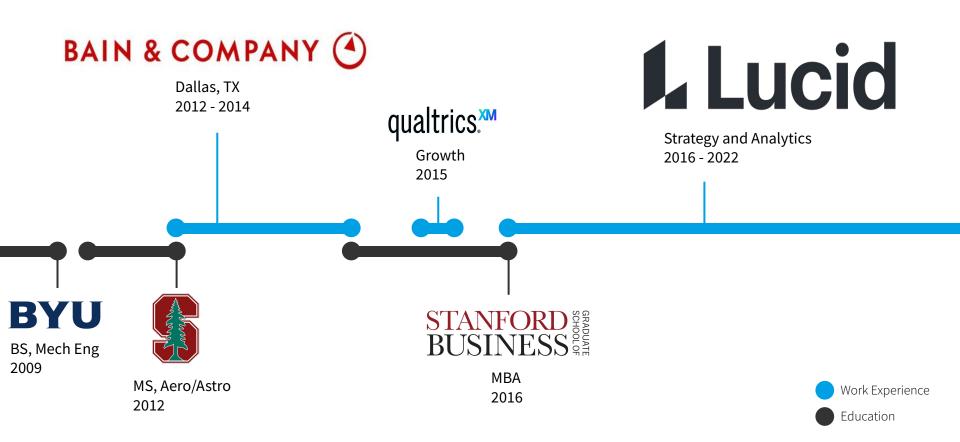
Building a Data Analytics Function

Tim Jenkins

My experience in Data Analytics



This presentation covers several common questions about building a data analytics function

- When is the **right time to hire** dedicated analytics resources? What should companies look for?
- How should companies handle data/analytics needs **before having a dedicated person**?
- What **tools/technology** do companies need, and when?
- What should the **org structure** look like for analytics?
- How should **analysts** be **spending their time**?
- What are some **common mistakes** to avoid?

FOR DISCUSSION

What are you most interested in learning about building a data analytics function?



COME OFF MUTE OR COMMENT IN THE CHAT

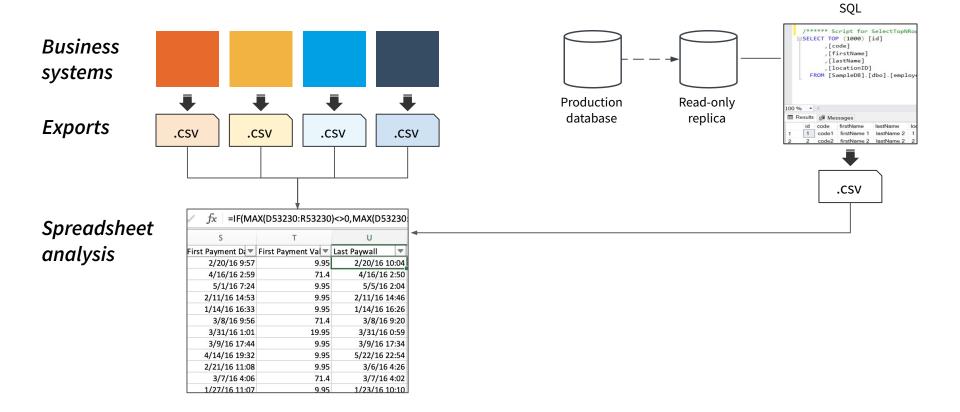
The primary purpose of analytics is to accelerate and improve decision-making

Primary purpose of analytics

analytics	Enable the business to make better decisions				
Secondary purposes	Necessary for robust reporting	Satisfy customer demand for data	Potential to productize data		

- People across the company make decisions every day
- When data is reliable, easy to access and use ⇒ better decisions
- When data isn't trusted, is difficult to access and use ⇒ poor decisions or acting on instinct

Early-stage companies can get by with out-of-the-box reporting, spreadsheet-based analytics, and basic SQL



When your company reaches 20-50 people, consider a dedicated analytics hire

Signs that you're ready for dedicated analytics resources

Analysis increasingly requires data from multiple systems



Reporting needs are known, stable, and ready to automate



Functions want to hire their own analysts

Your first hire should be someone who can provide value on day one but also scale and build a team



What to look for in an analytics lead

Business acumen

Capable of thought partnership with stakeholders and leaders across the company to solve business problems (likely finance/consulting background and/or MBA)

• Technical inclination

Modern analytics is a technical domain, with increasing similarities to software engineering

• Scalability

Eager to roll up sleeves but also capable of building and managing a team down the road

What should a new hire's first steps be?

Stand up core tech

Transition reporting

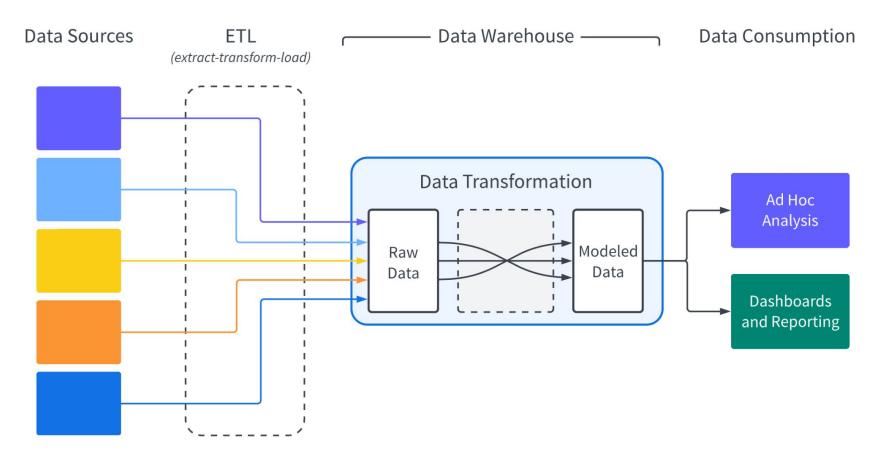
Learn while responding to ad-hoc questions



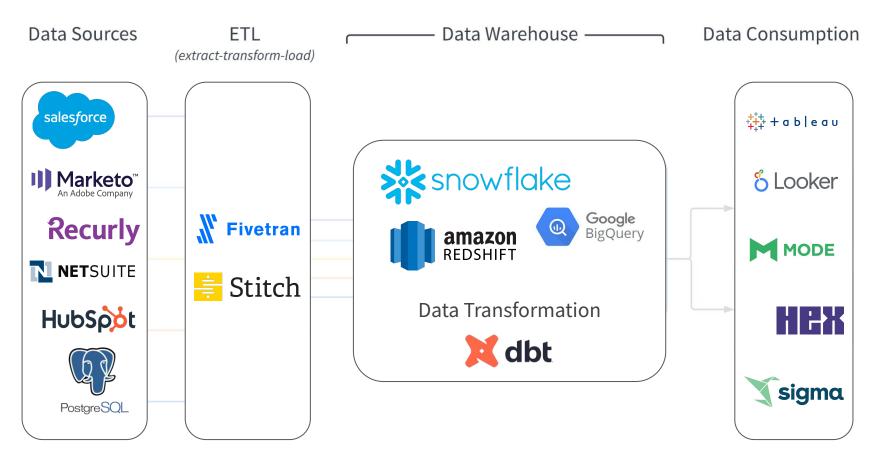




The modern data stack consists of a few core components



Various vendors exist for various parts of the stack



Depending on business needs, you may need to add components to your data stack over time

You may need	Event tracking and analytics	Orchestration and scheduling	Data science / machine learning	Reverse ETL / Operational analytics	Data Quality / Observability
	Track user behavior on your website and in your product	Schedule and manage complex data processing tasks	Build and deploy predictive models	Move data from your warehouse back into operational systems	Help monitor, alert, and troubleshoot data issues and anomalies
	Amplitude	Apache Airflow Cagster PREFECT	databricks	hightouch	MONTE CARLO great_expectations Datafold

Analytics teams can be structured in a variety of ways depending on priorities and business needs

Decentralized

Teams have their own analysts that provide dedicated support

Centralized

Analysts work together on a single team to support other teams within the company

Hybrid

Analytics is centralized, but most analysts are embedded within other teams

Decentralized analytics teams optimize for agility



Pros

- Speed Dedicated resources provide quicker turnaround
- Insight Because analysts are closer to the problem, they can come up with better solutions and more readily identify opportunities
- Specialization Analysts develop expertise in their area over time
- Impact People are more likely to listen to recommendations when analysts are part of their team

Cons

- Inefficiency Each group comes up with its own solutions, creating redundancy in work, tools, processes
- X Inconsistency Metrics won't match across the org, which can cause people to lose trust in the data
- **Recruiting** Career progression is limited, work gets stale, and analysts can feel isolated and buried in silos
- **X** Tactical focus Analysts can get bogged down in reporting; nobody focuses on bigger strategic issues
- **Resilience** Risk if analysts leave because of their unique domain knowledge

Centralized teams optimize for consistency



Pros

- Consistency Metrics are calculated the same way throughout the org
- Credibility Analysis is less biased because people aren't "grading their own work"
- Efficiency Able to make company-wide investments to standardize tools/processes
- Autonomy Team can prioritize high-value projects from across the company
- Culture Easier to develop a holistic analytics strategy; signals that the company values analytics
- Career Path Project variety, learning from other analysts, and opportunity to manage within team

Cons

- **Reactivity** Team can easily devolve into "order takers" for the rest of the company instead of thinking proactively
- **Responsiveness** Can be slower and less responsive to needs of the business
- **Detachment** Analytics team can grow out of touch with the business; other teams may view analytics as someone else's problem
- **Rivalry** People dislike relying on resources they don't control, and teams may feel like they're competing with each other for support

Hybrid teams aim for the best of both worlds

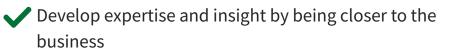


Pros

Centralization...

- Allows for standardization of metrics, tools, and processes
- Enables work on strategic, company-wide initiatives
- Facilitates learning and sharing of best practices between analysts

Embedded Analysts...



Gain broad experience by rotating through different functions

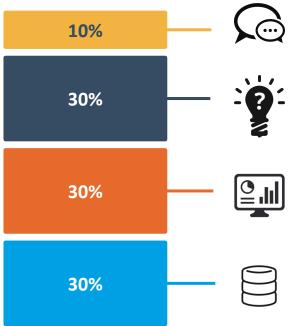
Cons

- **Control** Analysts report through analytics team but sit with the teams they support, which can be challenging to manage
- X Integration Analysts may have trouble integrating into other teams and may be treated like second-class citizens

Analysts split their time between data management, reporting, ad hoc questions, and strategic projects

Analyst time breakdown

(rough guidelines)



- Answering **ad hoc questions**, diagnosing issues, and pulling data
- Answering important **strategic questions**, providing recommendations, and implementing solutions

Building dashboards and automating processes

Managing the data warehouse and **transforming raw data** to be usable by the analytics team and others

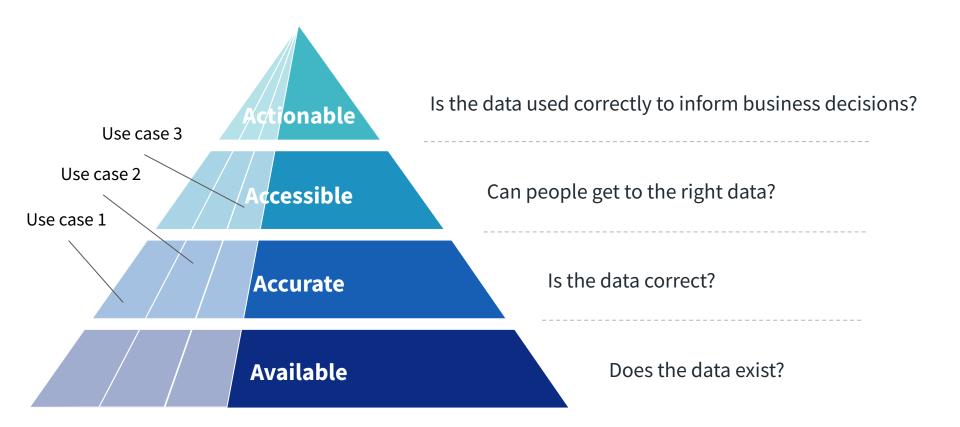
In standing up analytics, remember to create trust, provide supporting resources, and build use case by use case



If people don't trust the data, they aren't going to follow your team's recommendations and will rely on their own instinct As much as the modern data stack has empowered analysts, you will still need engineering support in some cases

Pick a particular analytics use case and solve it from end to end (from raw data to insights and operational decisions)

Build your capabilities in slices, not layers



Questions