

COSC 419:
Mobile Educational Game
Development

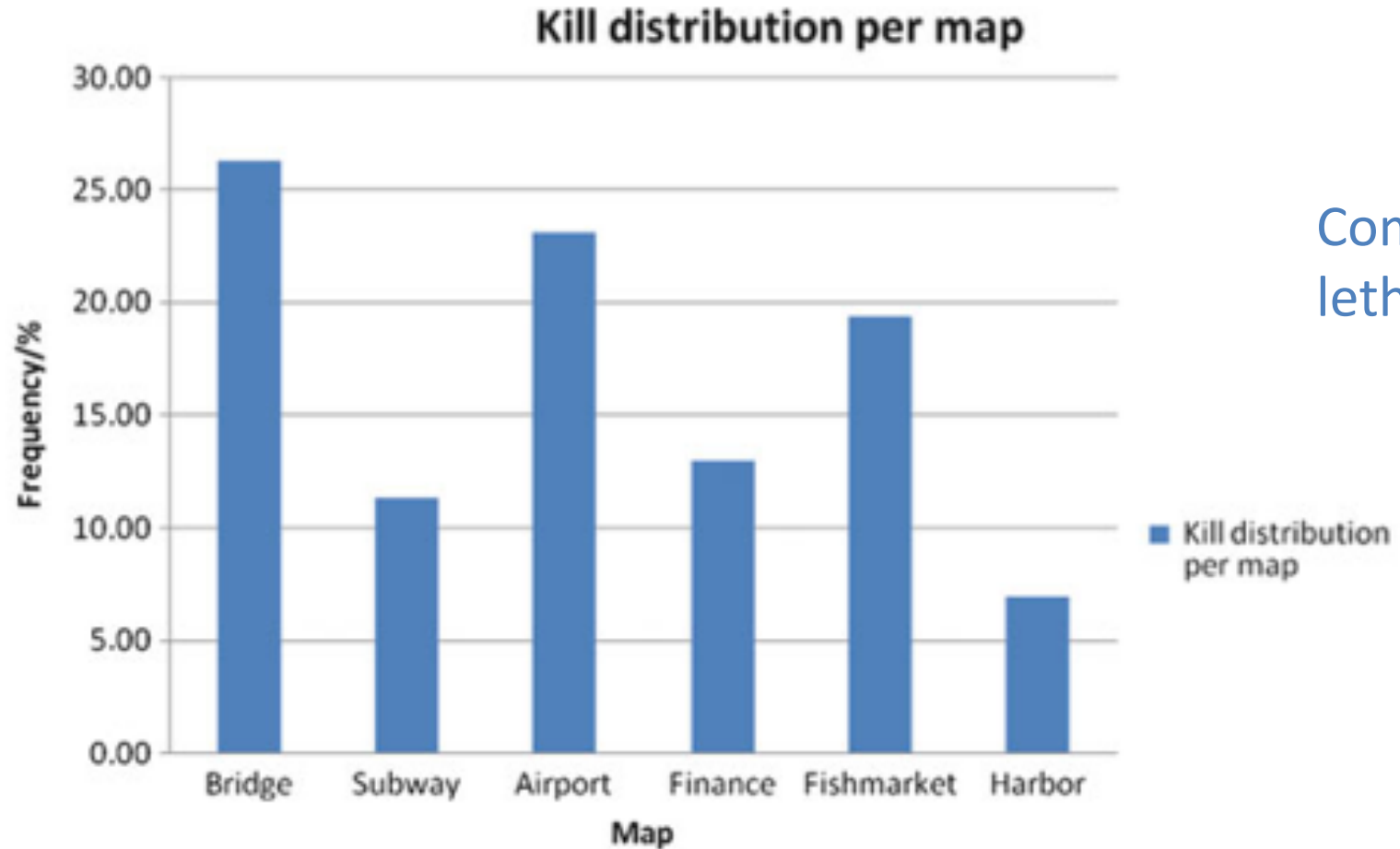
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University of British Columbia Okanagan

Game Metrics

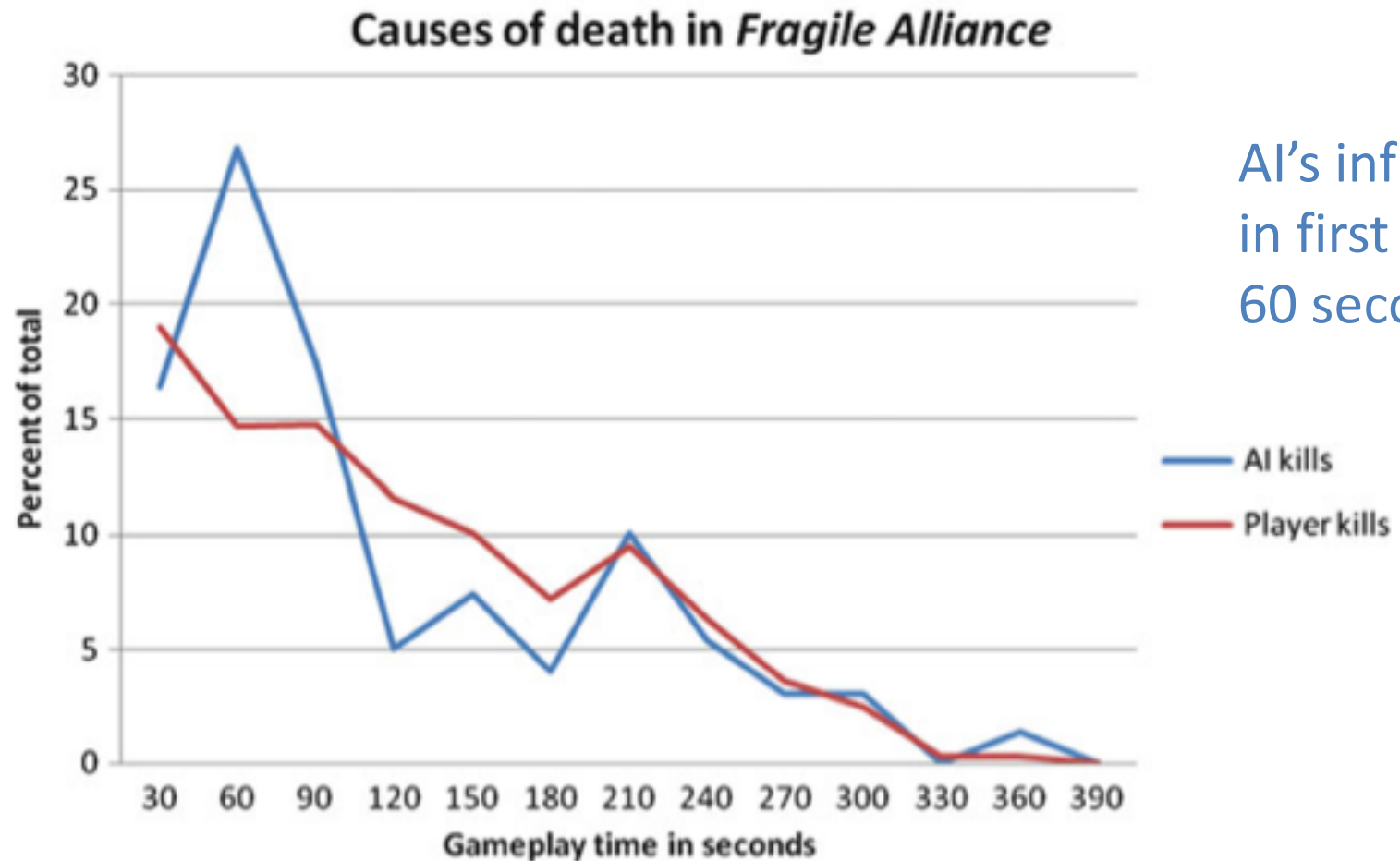
- What can you learn from game metrics?
- Once you have the metrics in place ...
 - What kind of data to collect?
 - How to analyze data?

Multiplayer Shooter Game: Fragile Alliance 2



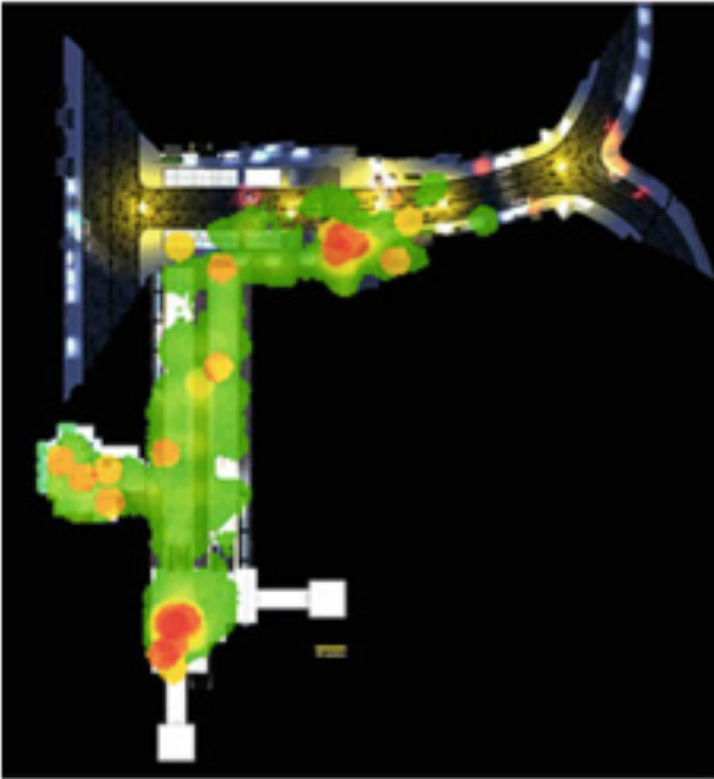
Compares
lethality

Multiplayer Shooter Game: Fragile Alliance 2

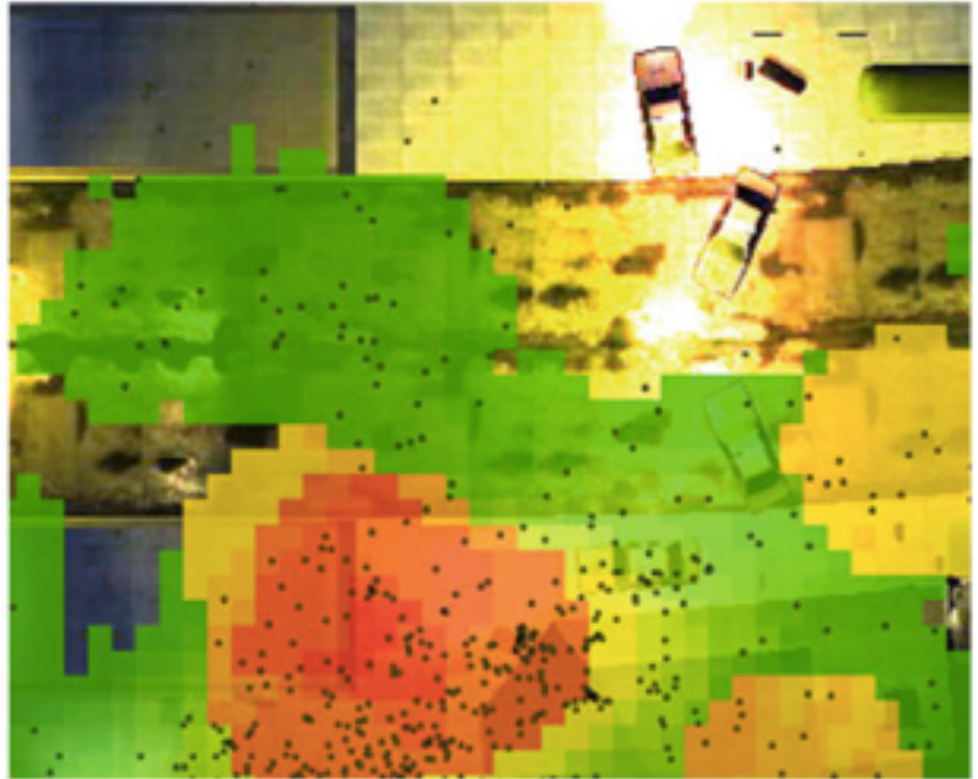


AI's influence
in first
60 seconds

Multiplayer Shooter Game: Fragile Alliance 2



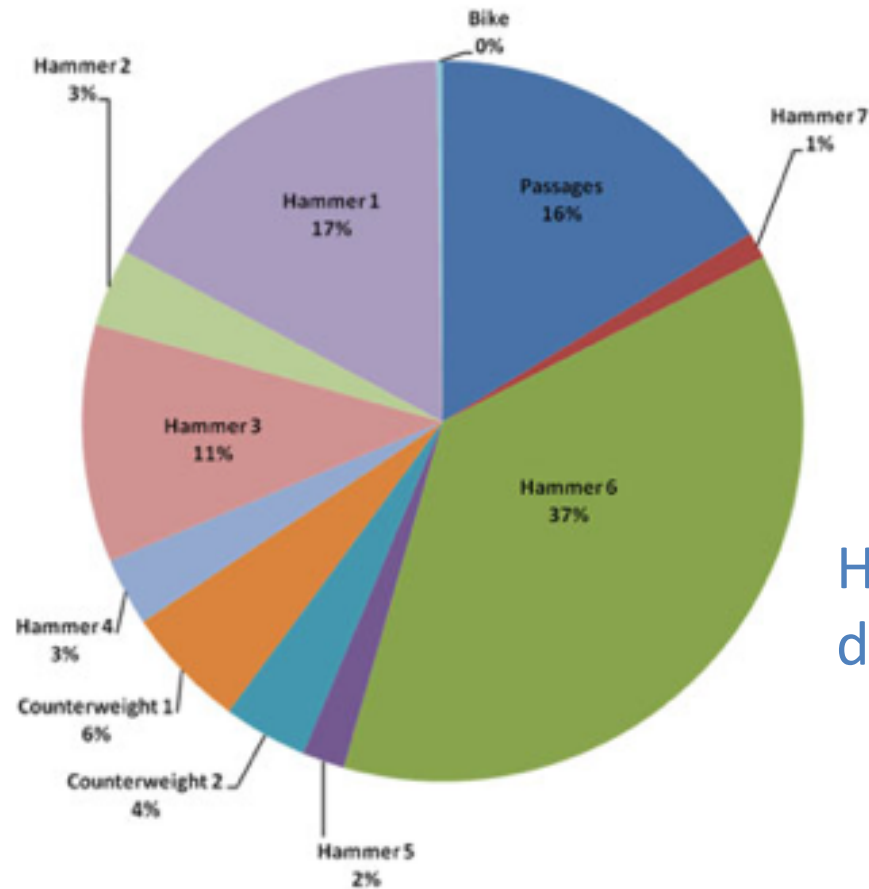
Subway map: overview



Subway map: close up

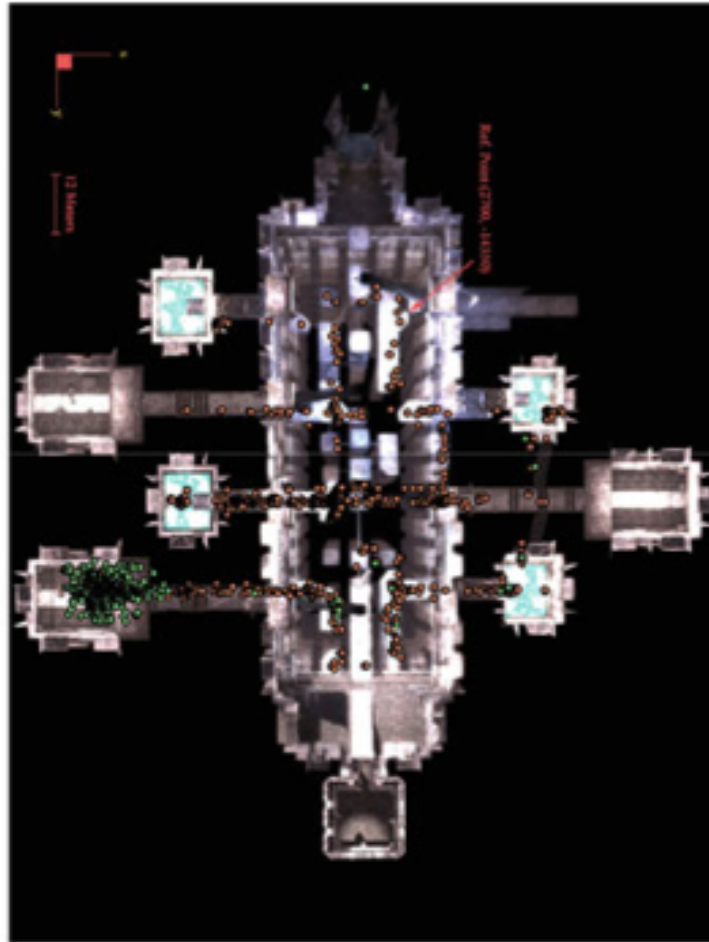
Tomb Raider: Underworld

Help on Demand frequencies - Valaskjalf



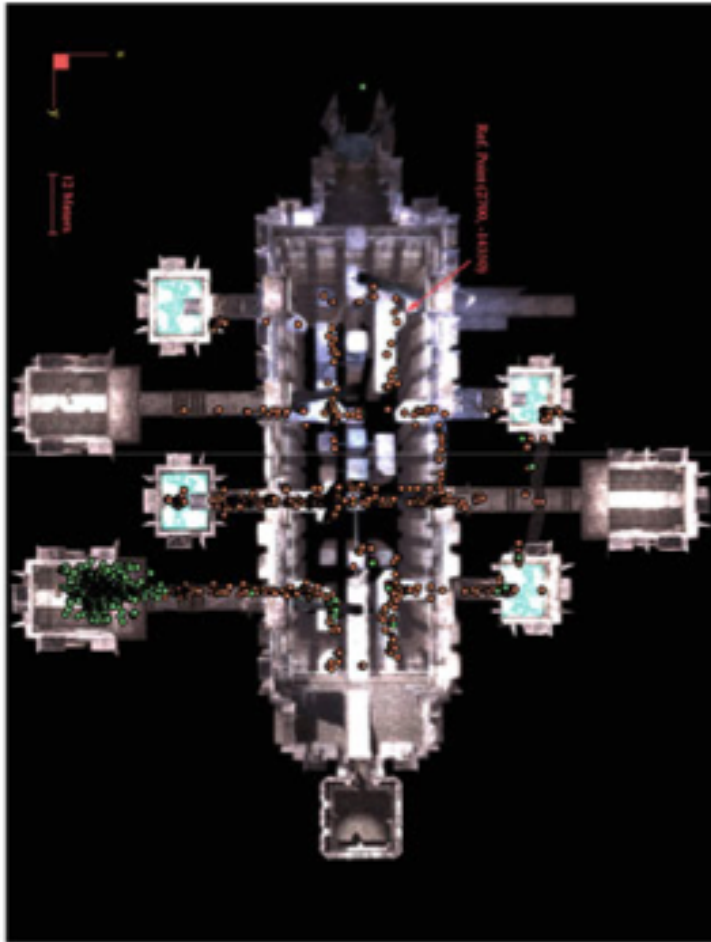
Help requests for different puzzles

Tomb Raider: Underworld

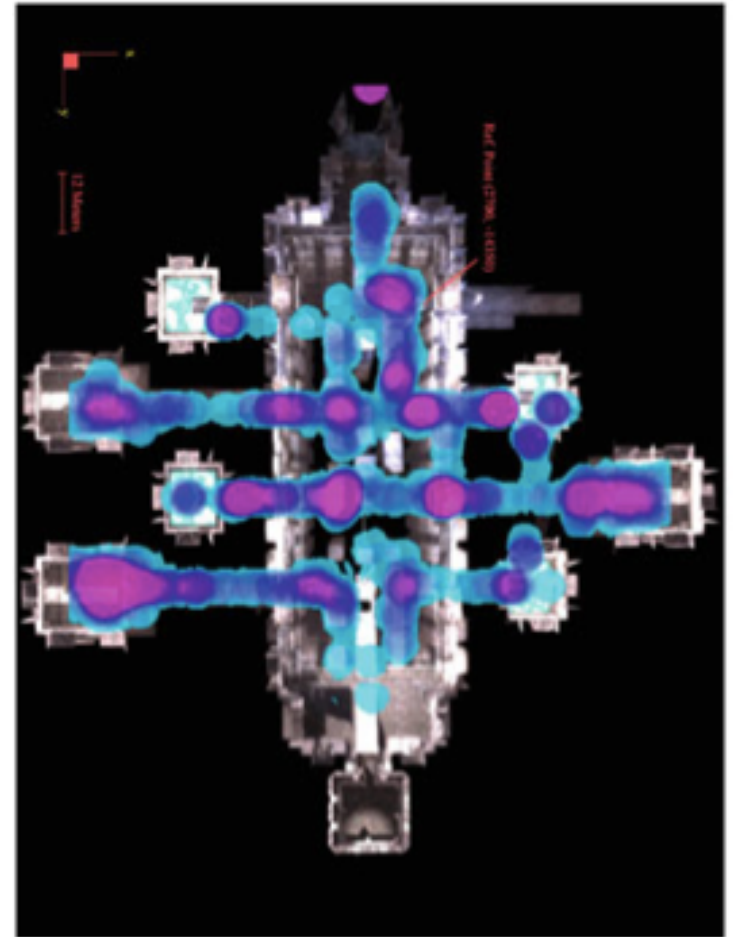


Positions when
help was asked

Tomb Raider: Underworld



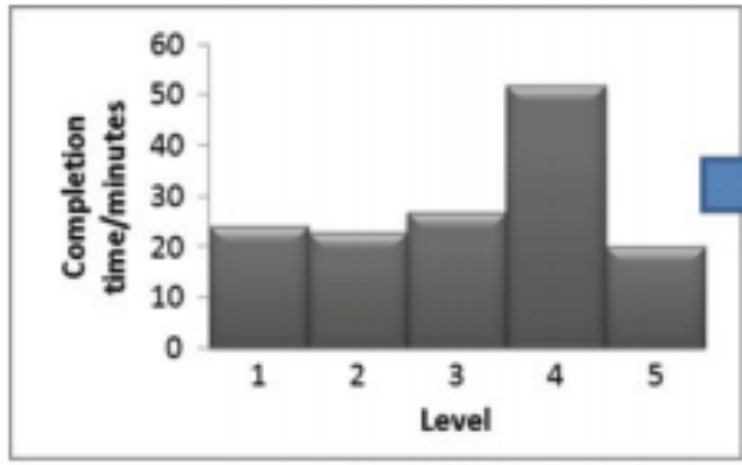
Correlate
to death
events?



Explorative vs. Hypothesis-Driven

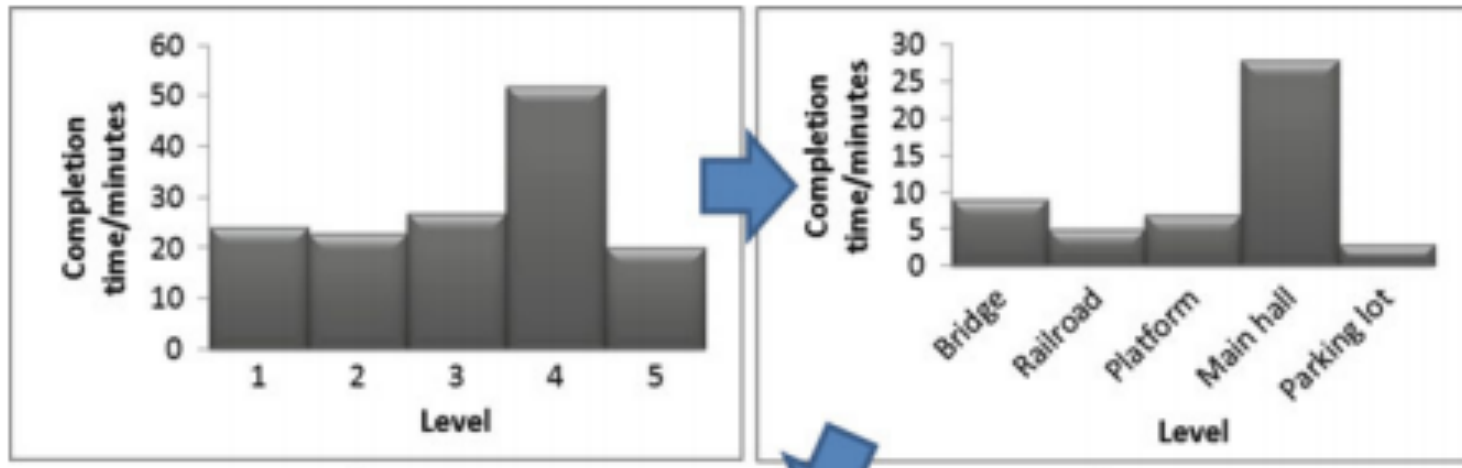
- Not to be confused with hypothesis testing
- Explorative:
 - See what the data can tell you
- Hypothesis-driven:
 - See how the data answers your question

Explorative Example



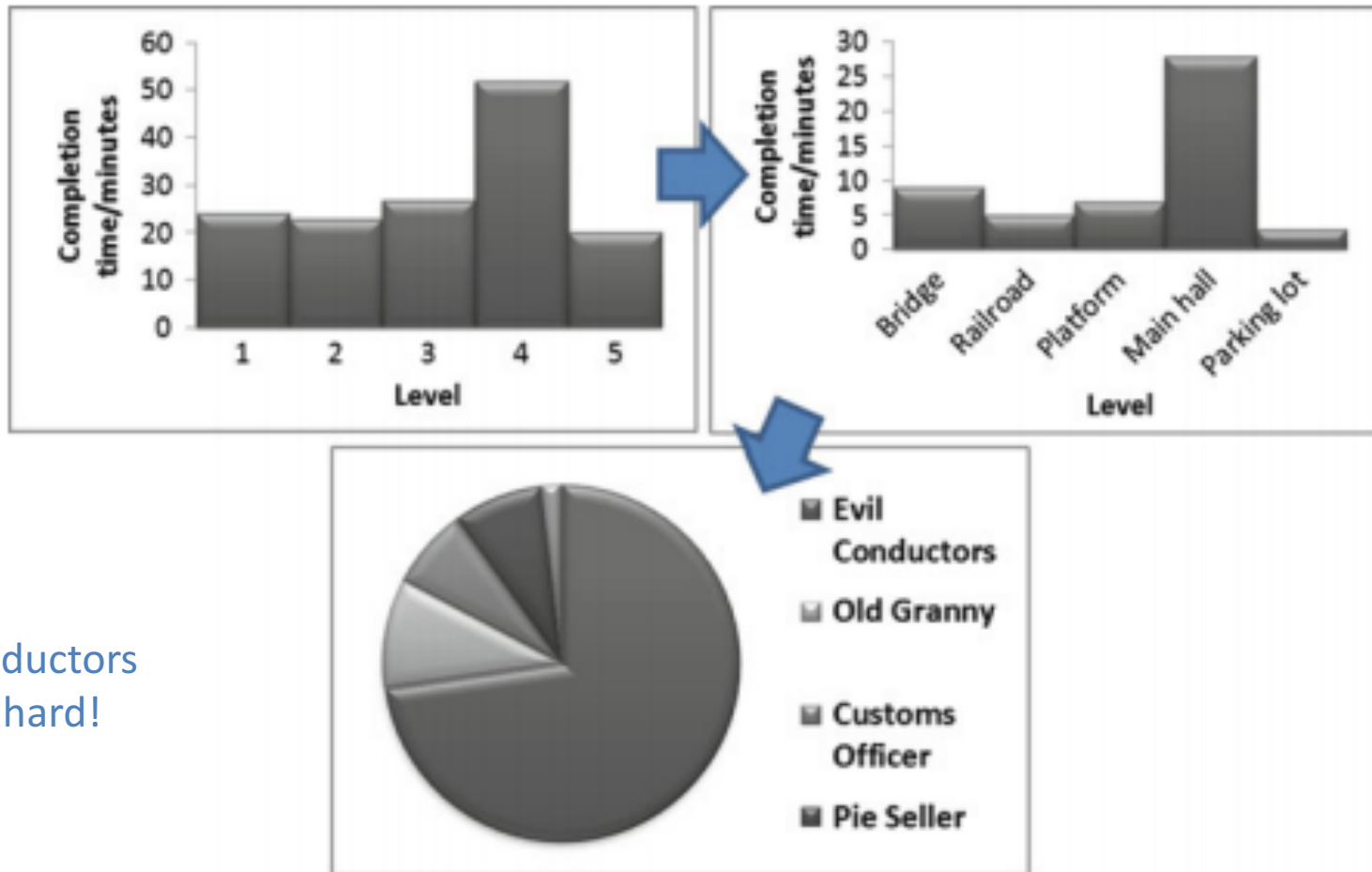
unintended design:
level 4 takes too long

Explorative Example



drill down into level 4:
problem in main hall

Explorative Example



evil conductors
are too hard!

Case 1: Weapon usage metrics in Kane & Lynch: Dog Days

- Explorative approach: how weapon attributes affect player behaviour?
- Short range weapons
 - Drive players forward if enemies are far away
 - More intense experience
- Long range weapons
 - Sniper-like effect
 - Less roaming behaviour
 - Gives marksman-like experience



Fig. 14.3: Green dots are the player positions at the time killing shots were made; the *colored lines* represent the lines of fire (different colors are different weapons), and *red dots* are locations of enemies at the time of death

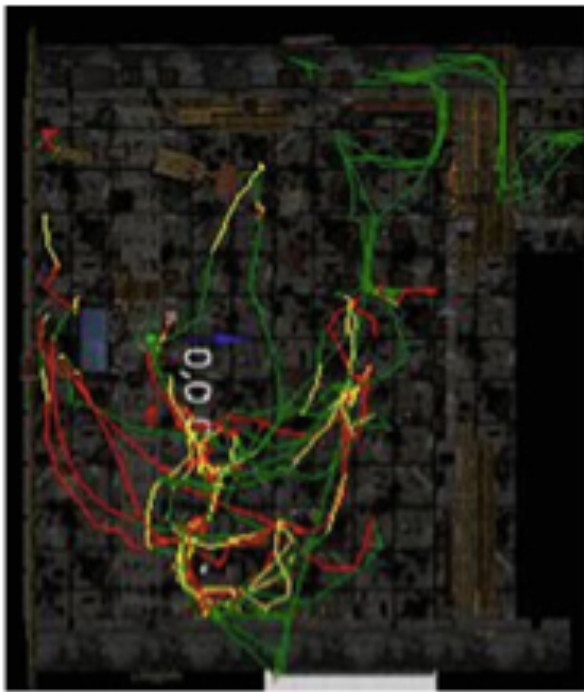
Sector 1



Fig. 14.4: (left) location vs. health
green is strong
red is close to death



Fig. 14.4: (right) locations of death



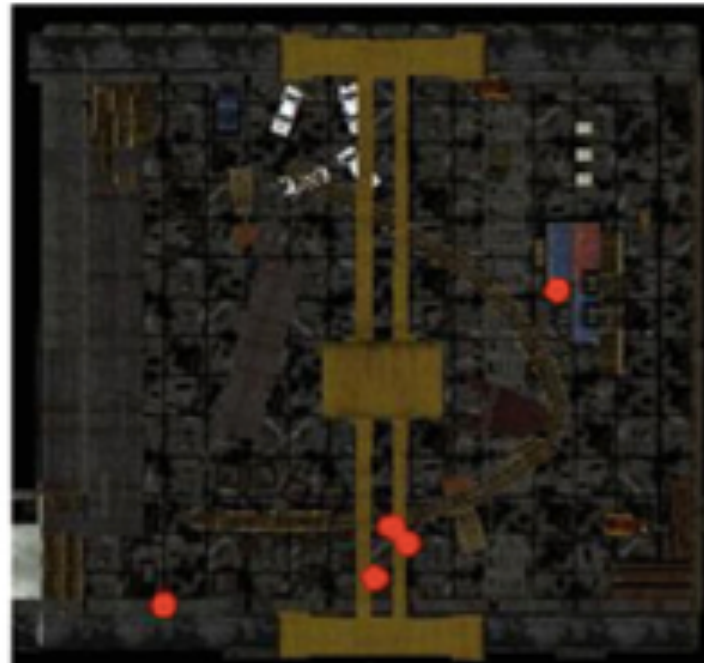
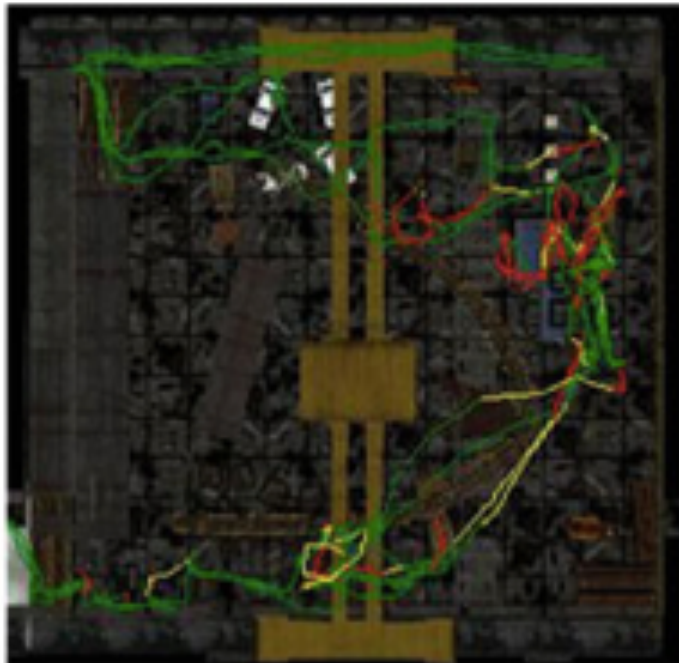
Sector 1:

- player under more pressure
- more deaths

Sector 1 (top)

vs.

Sector 2 (bottom)



Deriving Analysis Metrics

- Create kills/location ratio
 - If k/l is high, then experience is more like Sector 1
- Add kill distance metric
 - If long distances, then more sniper-like experiences
- Further analysis and data collection to confirm these relationships

Case 2: Gameplay analysis in Fragile Alliance 2

- Hypothesis-driven: did players die in the right locations with the right frequencies as intended by the design?
- Mercenaries try to steal \$, cops (AI bots and players) try to catch them, mercenaries can kill each other (to steal \$) and become traitors
- Focus data analysis on death events: where, when, why

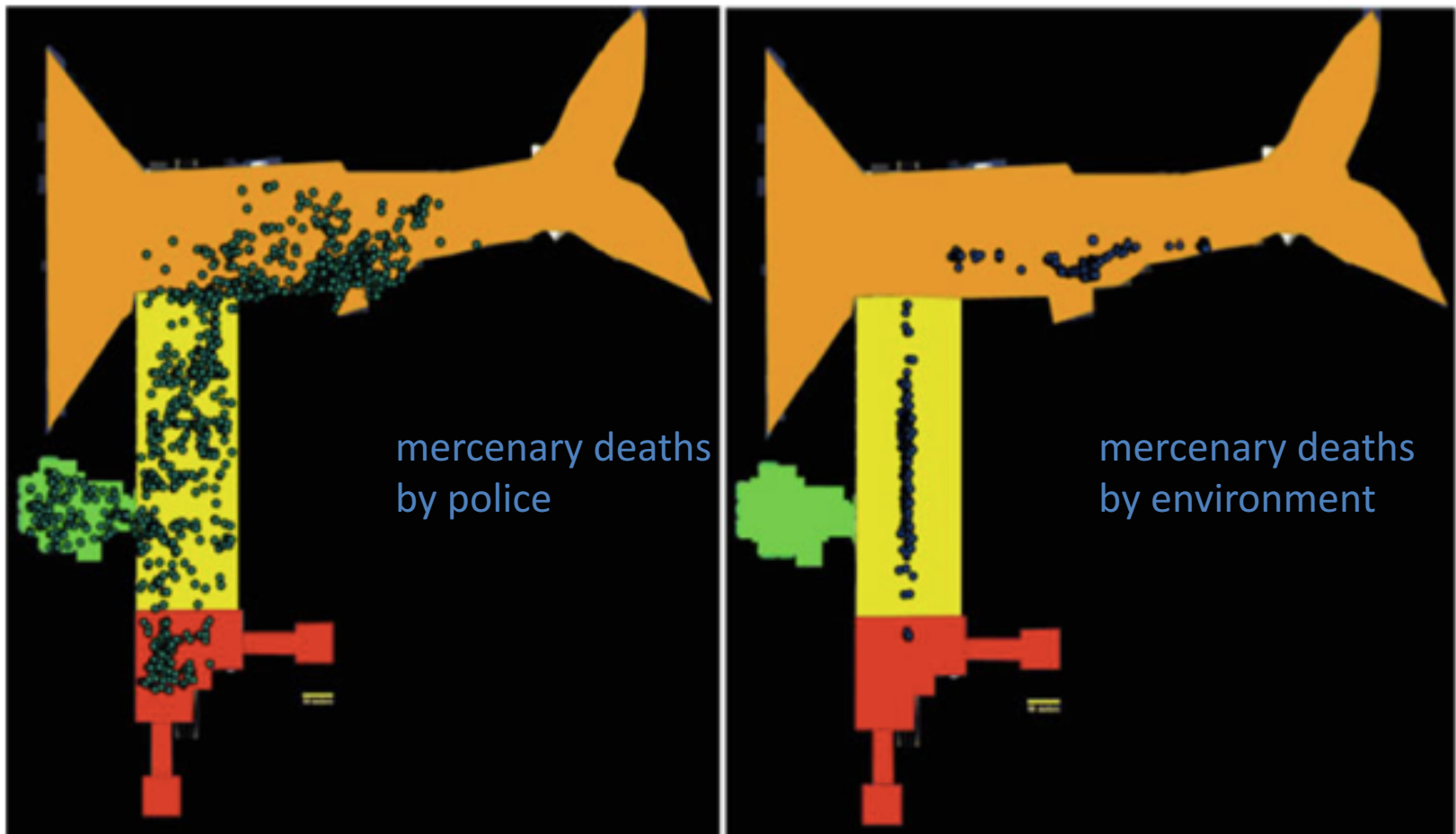


Fig. 14.8: The “Subway” map divided as: *Red* = spawning area for mercenary players; *Yellow* = subway; *Green* = vault area; *Orange* = road/exit area (+ spawning area for police. *(left)*: The locations where police officers were the cause of death. A broad distribution is apparent indicating that AI police officers can reach the entire map. *(right)*: Locations of environment-caused death events. *Yellow* sector events are caused by players being run over by a subway train while crossing a set of tracks, while the orange sector events are caused by exploding scenery (e.g. cars that explode).

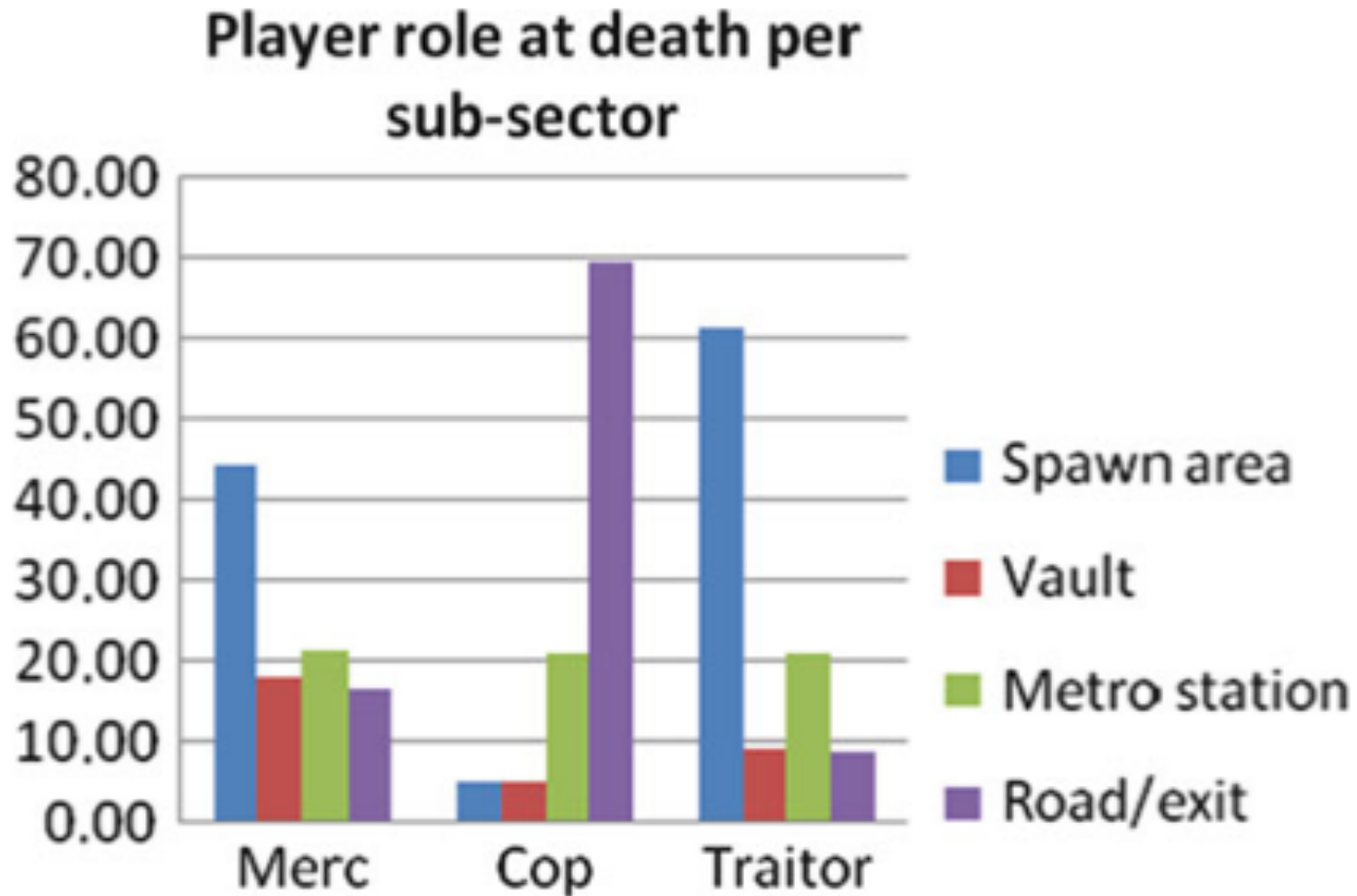
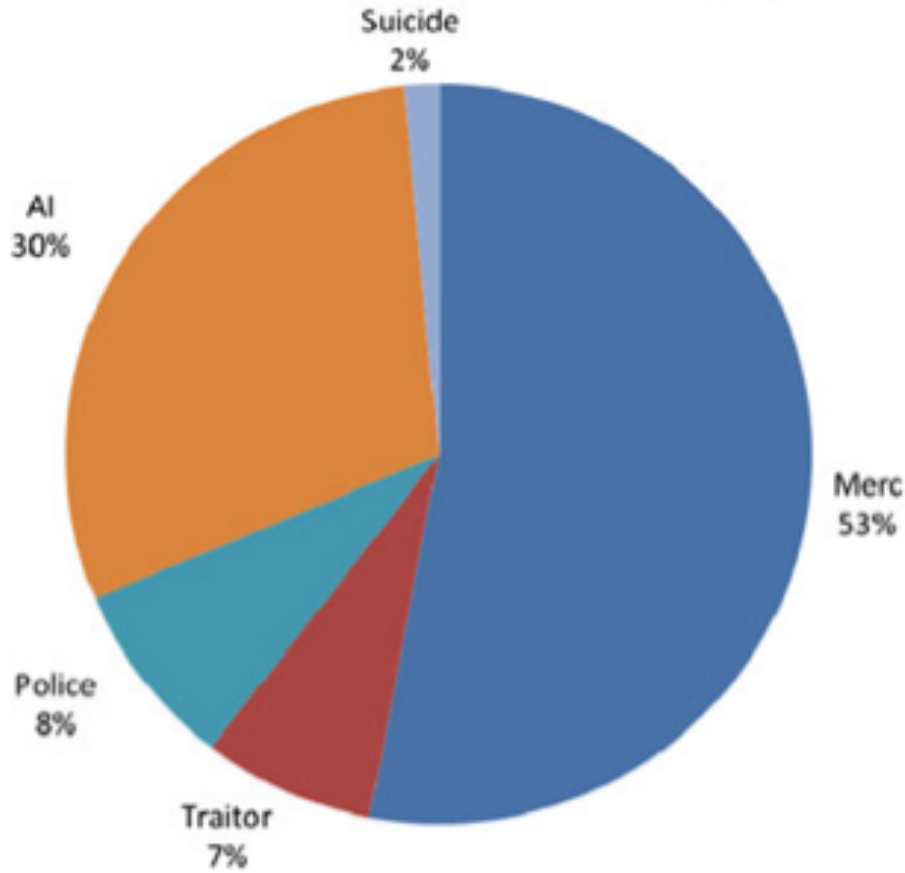


Fig. 14.10:

- Merc mostly killed at beginning but spread throughout the map
- Cop mostly killed at exit (where they enter), some cops penetrate to beginning
- Traitor mostly killed at beginning (unexpected)

Killer roles - first 45 seconds of play



Killer roles - following 90 seconds of play

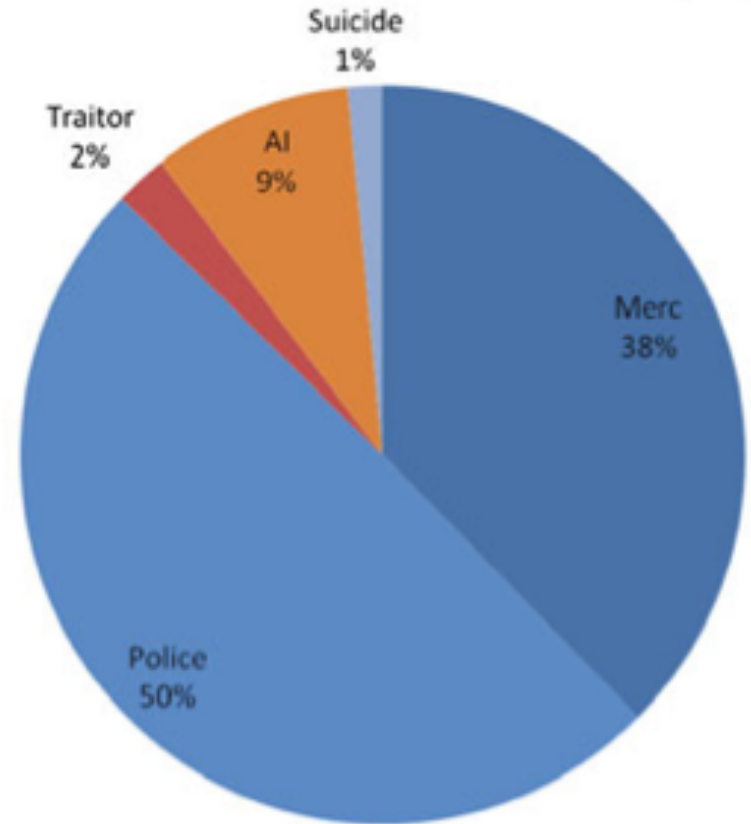


Fig. 14.11:

- how shift in power manifests
- mercenaries are initially powerful (when they die, they become police players)
- comparison of dominant kills across time

Alternative view: line graph of kills per role over time
can see when power “crosses over”

Case 3: Frustration analysis in Kane & Lynch: Dog Days

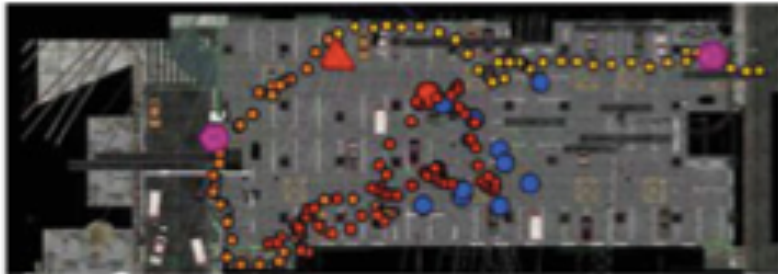
- Explorative approach
- Frustration:
 - Groaning, throwing the controller away, agitated, angry, etc.
- Tries to identify which behaviours led to exhibited frustration
- Problem:
 - Relationship between **observable behaviours** and **hidden variables** (causes)
 - Solution: Couple metrics analysis with user playtesting and questionnaires

Initial Analysis

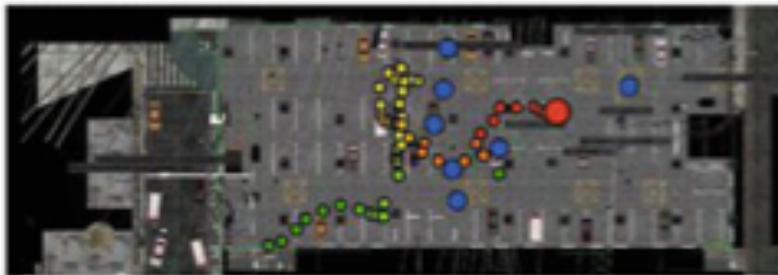
- Use insights from one frustrated player
 - Video recording, game activity, in-person feedback
- Exhibited behaviour include:
 - Body movements, facial expressions, verbalizations
- Corresponding game behaviour:
 - Inability to advance (e.g., repeated deaths in same location)



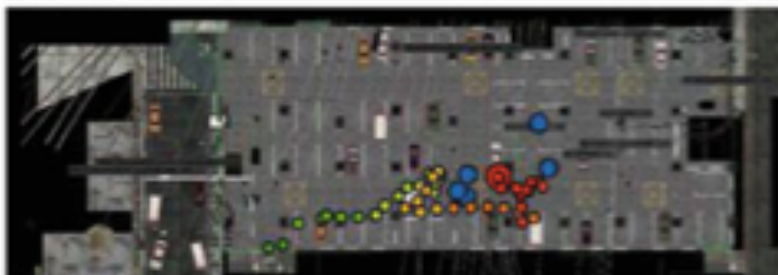
Fig. 14.14 Visualization of the spatial behavior of a player through the environment of a *Kane & Lynch: Dog Days* level, between two death events. The location of the player is plotted at each second of playtime, and a color scale applied to show the dimension of time along the path. Various events are plotted as symbols: enemy kills (*blue dots*), weapon pickups (*red triangle*) and taking cover (*green squares*). Spatial metrics visualizations such as this one are highly useful for the



First attempt



Second attempt



Third attempt



Fourth attempt



Fifth attempt (success)

Figure
 Medium
 Large
 Orange
 Hexag
 Triang
 Light g
 Dark g

 The su
 an inte
 red, co
 to later

Possible Indicators

- Player dies repeatedly in same location
- Decrease in progress in successive attempt
- Number of enemy kills decreases in successive attempt
- Pace of movement increases in each attempt
- Similar route taken each time
- Minimal use of special abilities or weapon pickups
- Camera angle coincides with character movement (less interest in exploring environment)

Confirming Frustration Indicators

- Sample of 22 players
 - 6 players exhibited similar behaviours
- Follow-up interviews with animated replays to confirm frustration behaviours



Model Creation

- Collectively generalize data to model:
 - Decrease in progress in successive attempt
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What about the other players?

Six Myths about Game Metrics

- 1. With game data mining, we can fire our designers and testers!

Six Myths about Game Metrics

- 1. With game data mining, we can fire our designers and testers!
- Wrong!
 - Mining is incredibly useful for evaluating design
 - But data cannot tell us what good design is, or how users feel, or what makes a good game experience

Six Myths about Game Metrics

- 2. With game data mining, we do not need user playtesting!

Six Myths about Game Metrics

- 2. With game data mining, we do not need user playtesting!
- Wrong!
 - Results compliment each other, not replace
 - Each have different advantages

Six Myths about Game Metrics

- 3. Game analytics is autonomous, requiring little human oversight!

Six Myths about Game Metrics

- 3. Game analytics is autonomous, requiring little human oversight!
- Wrong!
 - Analytics is a process
 - Need to identify goal, problems, relevant patterns, possible indicators, combined use of other info sources, etc.
 - Results require interpretation

Six Myths about Game Metrics

- 4. Game data mining pays for itself quickly!

Six Myths about Game Metrics

- 4. Game data mining pays for itself quickly!
- Maybe!
 - Depends on the problems identified
 - Investment to develop or purchase game analytic tools can be costly

Six Myths about Game Metrics

- 5. Game data mining will solve all our problems!

Six Myths about Game Metrics

- 5. Game data mining will solve all our problems!
- Wrong!
 - Game data can uncover patterns of behaviour: both game play and business oriented patterns
 - Patterns still require human interpretation
 - Still need to identify (and possibly fix) causes of patterns

Six Myths about Game Metrics

- 6. We need to obtain data on everything!

Six Myths about Game Metrics

- 6. We need to obtain data on everything!
- Wrong!
 - Need the *right* data
 - Don't waste resources
 - Too much data = shooting in the dark