

Lineup Cohesion Score

A Network-Based Metric for Optimal Starting Eleven Selection

NEU Sports Analytics Hackathon

Prompt A: Starting Eleven Lineup Construction

February 2026

Data: IMPECT Open Data • Bundesliga 2023/24 • 306 Matches

The Challenge: Optimal Lineup Selection

How do coaches select a starting eleven that maximizes team performance?

Key Questions

- Which players connect well together?
- Who are the critical hub players?
- What passing patterns lead to goals?

Our Approach

- Build player-to-player pass networks
- Weight edges by shot creation
- Quantify lineup "cohesion"

Key Insight

Elite teams are HUB-DEPENDENT: they funnel play through star players

(Counter to the intuition that "balanced" teams perform better)

Methodology: Lineup Cohesion Score

$$\text{Cohesion} = 0.50 \cdot \text{Connectivity} + 0.25 \cdot \text{Chemistry} + 0.15 \cdot \text{HubDep} + 0.10 \cdot \text{Progression}$$

Connectivity

(50%)

Network density +
avg clustering coefficient

*How interconnected
are the players?*

Chemistry

(25%)

Critical position pair
pass frequency

*Midfield→Wing,
Midfield→Striker links*

Hub Dependence

(15%)

Gini coefficient of
degree distribution

*Star player
reliance (GOOD)*

Progression

(10%)

Pre-shot pass ratio
(passes → shots)

*Attacking
effectiveness*

$$\text{Edge Weight} = \text{PassCount} \times (1 + \text{PreShotRatio})$$

Pipeline:

IMPECT Events → Filter Passes → Build DiGraph → Compute Metrics → Aggregate Season

* Weights optimized empirically: original "Balance" inverted to "Hub Dependence" based on correlation analysis showing elite teams are MORE centralized, not less.

Validation: Cohesion Predicts Season Performance

✓ Season-Level

$$r = 0.728, p < 0.001$$

Explains 53% of variance
in season points

✓ Match-Level

$$F = 36.64, p < 0.0001$$

Significant difference
win vs draw vs loss

Component Correlations with Season Points

Component	r	p-value	Sig
Connectivity	+0.785	0.0001	***
Hub Dependence	+0.714	0.0009	***
Chemistry	+0.448	0.0623	*
Progression	+0.133	0.5978	

□ Key Finding: Connectivity is the strongest predictor

Dense, well-clustered passing networks strongly associated with success

The "Balance Paradox": Star Players Matter

Original Assumption

"Balanced" teams perform better
(Even pass distribution)
 $r = -0.714 \text{ with points } \square$

Empirical Reality

Hub-dependent teams win more
(Star player centralization)
 $r = +0.714 \text{ with points } \checkmark$

Evidence from Top Teams:



Leverkusen (90 pts)

Xhaka orchestrates; Wirtz finishes



Bayern (72 pts)

Kimmich as central hub



Stuttgart (75 pts)

Clear passing hierarchy

\square Implication for Lineup Selection

Optimize for HUB CONNECTIVITY, not equal distribution.

Build lineups around your best playmaker.

Case Study: Leverkusen's Undefeated Season

2023/24 Season

28W - 6D - 0L

90 points | +63 GD

87 goals scored

Cohesion Score

0.560

League avg: 0.548

Hub Players

Granit Xhaka

(Volume hub: 558 passes)

Florian Wirtz

Top Attacking Connections (Pre-Shot Passes)

Connection	Pre-Shot	Conversion
Wirtz → Boniface	22	23.4%
Frimpong → Boniface	12	38.7%
Xhaka → Wirtz	11	6.1%
Palacios → Boniface	9	15.5%

▣ **Tactical Pattern: The Leverkusen Chain**

Defense → Xhaka (orchestrate) → Wirtz (create) → Boniface (finish)

Wirtz leads ALL pre-shot connections — the attacking fulcrum

Application: Lineup Optimization Framework

For Coaches

- Identify optimal player combinations
- Quantify impact of substitutions
- Plan for injuries: who maintains hub connectivity?

For Analysts

- Compare lineup alternatives
- Scout opposition networks
- Evaluate transfer targets' fit with existing passing structure

Example: Evaluating a Lineup Change

Lineup A (Current)

Cohesion: 0.52

Connectivity: 0.31

Lineup B (Proposed)

Cohesion: 0.58 (+12%)

Connectivity: 0.36



⚠ Limitations

- Requires historical pass data
- Single season validation
- Does not capture off-ball movement
- Opponent-specific effects not modeled

Conclusions & Future Work

Key Contributions

1. Validated Cohesion Metric

$r = 0.728$ with season points ($p < 0.001$)

2. Hub Dependence Insight

Elite teams centralize play through stars

3. Pre-Shot Weighting

Edges weighted by attacking contribution

4. Leverkusen Case Study

Xhaka → Wirtz → Boniface chain identified

Future Directions

- Incorporate opponent-specific adjustments
- Add temporal dynamics (fatigue, momentum)
 - Extend to multi-season validation
 - Build interactive lineup optimizer tool

Thank You!

Code: [github.com/\[your-repo\]](https://github.com/[your-repo]) | Questions?