Estimating the Number of Legal Chess Diagrams Using Monte Carlo Simulations

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Abstract

The total number of legal chess diagrams remains unknown due to the immense combinatorial complexity of chess and the difficulty of verifying legality (rule consistency) and reachability (arising from a legal game). This work presents a Monte Carlo sampling approach operating on the level of material classes—distinct configurations of chess pieces. For each class, random diagrams are generated and tested for legality using the python-chess library. By combining combinatorial enumeration of diagrams within each class with empirically estimated legal ratios, we derive an overall upper bound for the number of legal chess diagrams without promotions. Our results suggest an upper bound of approximately $(1.41 \pm 3\%) \times 10^{41}$ legal diagrams, corresponding to roughly one in every 800 random piece placements being legal. These findings highlight the distribution of legality across material classes and provide a foundation for more generalized estimates of legal chess diagrams in future work.

1 Introduction

1.1 Scientific Overview

Estimating the total number of possible chess diagrams or positions has long been a topic in chess mathematics. Claude Shannon [1] gave an early rough estimate of the number of possible chess positions. His statement reads:

"The number of possible positions, of the general order of $64!/(32!\,8!^2\,2!^6)$, or roughly 10^{43} ."

This statement originates from Shannons foundational 1950 paper. His work focused primarily on the complexity of chess as a game rather than the enumeration of strictly legal positions. Modern studies, such as Chinchalkar (1996) [2] established an upper bound for the number of legal chess positions by basic combinatoric arguments and basic legality restrictions. Steinerberger (2015) [3] introduced the color-complex restriction for bishops and conjectured that an upper bound of $\approx 10^{35}$ could be obtained if pawn structure legality were incorporated. He also suggested Monte Carlo methods for sharper estimates. Tromp (2021) [4] implemented such an approach and, based on one million samples, estimated an upper bound of $\approx 10^{44}$ positions (termed *urpositions*). It remains unclear whether his analysis excluded promotions.

1.2 Motivation

Despite decades of research, no exact enumeration of legal chess diagrams or positions exists as of 2025. The challenge primarily comes from two aspects: the complex verification of legality (ensuring positions obey chess rules such as valid pawn placement, single kings per side, castling and en passant rights, and promotions) and reachability (ensuring the position can arise from the initial setup by a series of legal moves).

In addition, many authors apply different restrictions and assumptions when defining or estimating the number of chess positions or diagrams. These include whether promotions are allowed, whether illegal pawn placements are excluded, or whether reachable positions are considered or not. Such differences lead to significantly varying numerical estimates. Table 1 summarizes typical definitions and includes representative values from the literature.

Definition	Description	Upper bounds
All diagrams	All possible placements of pieces on the board, ignoring any legality constraints.	$\sim 10^{43} [1]$
Legal diagrams	Diagrams satisfying chess rules (e.g., exactly one king per side, no pawns on rank 1 or 8) without promotions	
Legal positions	Legal diagrams with additional information including side to move, castling rights and en passant possibilities.	
Reachable positions	Legal positions that can be reached from the initial setup through valid move sequences, without promo- tions.	
Reachable positions with promotions	Reachable positions including all possible promotions (e.g., multiple queens).	

Table 1: Different definitions of chess diagrams and positions with corresponding approximate estimates from the literature.

In this paper, we adopt a similar approach to Tromp [4], however, we focus on counting legal diagrams rather than positions. In contrast to Tromp's work, we will sample multiple subspaces defined by single material classes with Monte Carlo and not the full space of diagrams at once.

2 Methods

All code used for material class enumeration, legality checking, and Monte Carlo sampling is available as open-source on GitHub [8].

2.1 Material Classes

We define a material class as a set of specific chess pieces. Each material class specifies the number of kings K, queens Q, rooks R, bishops B, knights N, and pawns P present for white and black.

2.2 Definition of Structural Legal Diagrams

In this work we adopt a precise terminology to distinguish different levels of validity of chess positions:

- All diagrams: Pure geometric placements of the given pieces on the 64 squares, ignoring any chess rules.
- (Structural) legal diagrams: Diagrams that satisfy the structural constraints of chess (e.g., exactly one king per side, no pawns on ranks 1 or 8, no duplicate bishops on the same color complex without promotion). These are the objects studied in this paper.
- Legal positions: Structural legal diagrams enriched with meta-information such as side to move, castling rights, and en passant targets, all of which must be internally consistent.
- Reachable diagrams: A subset of structural legal diagrams that can in principle be constructed by a legal sequence of moves, but without specifying side to move or other position metadata.
- Reachable positions: A strict subset of legal positions, namely those that can actually arise from the initial starting position by a finite sequence of legal moves.

Remark. There exists a gray area between structural legal diagrams and reachable diagrams. For example, in our implementation we classify diagrams with six or more pawns of the same color on one file as "illegal". Strictly speaking, such diagrams are not forbidden by the chess rules, but they are unreachable in practice; hence they fall into this ambiguous intermediate zone. Our paper touches this boundary only lightly.

2.2.1 Analytical Calculation

The number of different material classes can be computed as follows: For a single side, the total number of possible piece count combinations with no promotions (limits: $K = 1, Q \le 1, R \le 2, B \le 2, N \le 2, P \le 8$) is:

num_combinations =
$$(K) \cdot (Q+1) \cdot (R+1) \cdot (B+1) \cdot (N+1) \cdot (P+1)$$

$$1 \cdot (1+1) \cdot (2+1) \cdot (2+1) \cdot (2+1) \cdot (8+1) = 486$$

Since white's and black's piece counts are independent, the total number of combined material classes is:

$$486^2 = 236.196$$

2.2.2 Python Implementation

The following Python code enumerates all material classes, verifying the piece count constraints and total number of pieces.

```
import itertools
limits = {"K": 1, "Q": 1, "R": 2, "B": 2, "N": 2, "P": 8}
def all_side_materials() -> List[Dict[str, int]]:
    """Generate all valid material configurations for one side.""
    side_classes = []
    for q in range(limits["Q"] + 1):
        for r in range(limits["R"] + 1):
             for b in range(limits["B"] + 1):
                 for n in range(limits["N"] + 1):
                    for p in range(limits["P"] + 1):
                        side_classes.append(
                             {"K": 1, "Q": q, "R": r, "B": b, "N":
                                n, "P": p}
                         )
   return side_classes
white_materials = all_side_materials()
black_materials = white_materials.copy()
classes = []
for w, b in itertools.product(white_materials, black_materials):
    total = sum(w.values()) + sum(b.values())
    # Limit total pieces on board
    if total <= 32 and len(w.values()) <=16 and len(b.values())</pre>
       <= 16:
         classes.append((w, b))
        # Stop early if max_classes reached
         if max_classes is not None and len(classes) >=
            max classes:
             break
                # 236196
return classes
```

Listing 1: Enumeration of all material classes in chess

Note: The check for total pieces per side \leq 16 (half the board) and total pieces \leq 32 have been added for enhanced correctness. They are only necessary when including pawn promotions.

Figure 1 shows the distribution of different material classes over the number of chess pieces.

2.3 Diagram Enumeration Within a Material Class

The total number of ways to distribute the given pieces of a material class over 64 squares (ignoring legality) can be expressed combinatorially as:

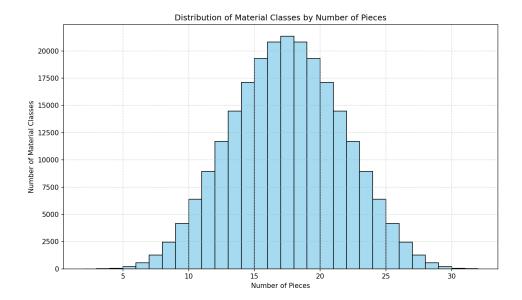


Figure 1: Distribution of the count of different material classes for chess pieces count of 2-32

Diagrams =
$$\frac{64!}{(64 - n_w - n_b)! \cdot \prod_i c_i!}$$

where:

- n_w and n_b denote the number of white and black pieces, respectively,
- c_i denotes the count of identical pieces (e.g., two white rooks, eight black pawns).

Combinatorial Derivation. Let n denote the total number of pieces to be placed on the 64 squares of the chessboard. If all pieces were distinguishable (e.g., labeled), the number of possible placements would correspond to the number of injective mappings from an n-set to the 64 squares:

$$64 \cdot 63 \cdot 62 \cdots (64 - n + 1) = \frac{64!}{(64 - n)!}.$$

In practice, however, many pieces are identical (e.g., the eight white pawns or two black rooks). Let c_i denote the count of identical pieces of type i. Since permutations among identical pieces do not yield distinct arrangements, we must divide by the product of the corresponding factorials. Thus, the total number of distinct arrangements (ignoring legality) is:

Positions =
$$\frac{64!}{(64-n)! \prod_i c_i!}$$

This formula counts purely geometric placements of the given piecescommonly referred to as *diagrams* and ignores all legality constraints (e.g., pawns on the first or eighth rank, both kings in check, etc.).

Example. White: 1 King, 1 Queen, 3 Pawns $(n_w = 5)$

Black: 1 King, 2 Rooks $(n_b = 3)$

Total n = 8. The factorials for identical pieces are 1!, 1!, 3!, 1!, 2!. Hence:

Positions =
$$\frac{64!}{(64-8)! \, 1! \, 1! \, 3! \, 1! \, 2!} = \frac{64!}{56! \cdot 3! \cdot 2!} = 14,871,915,636,480 \approx 1.49 \times 10^{13}.$$

Note: This formula counts only board *diagrams*. A full *position* in chess additionally includes meta-information such as the side to move, castling rights, and possible enpassant targets. Therefore, the total number of possible positions is higher, depending on the material configuration. Moreover, further constraints must be applied to filter only *legal* positions (e.g., valid king placement, pawn locations, and consistency of castling and check states).

2.3.1 Legal Diagram Verification Using python-chess

We employ the python-chess library [5] to validate generated chess diagrams, ensuring that they satisfy all basic structural legality conditions. The library provides a comprehensive set of status flags to detect inconsistencies or impossible states in a given position. Among others, the following criteria are verified automatically:

- Exactly one king per side (no missing or duplicate kings)
- No pawns on the first or last rank
- Valid castling rights
- Valid en passant squares and consistency with pawn structure
- No king in check if it is that sides turn
- No more than two simultaneous checks (STATUS TOO MANY CHECKERS)
- No geometrically impossible double checks (STATUS_IMPOSSIBLE_CHECK), e.g., two attacking pieces aligned on the same ray with the king, or en passant positions that cannot have arisen from any legal move

If none of these error conditions are triggered, the position is marked as STATUS_VALID, meaning it satisfies the fundamental requirements of internal consistency. However, this does *not* guarantee that the diagram is reachable by a legal sequence of moves from the initial position—it merely ensures that it respects the structural rules of chess.

```
status = self.status()
       # STATUS_VALID == 0 everything is fine
       if status == STATUS_VALID:
           return True
       # Ignore only the back-rank error
       if status & ~STATUS_PAWNS_ON_BACKRANK == STATUS_VALID:
           return True
       return False
# Main function: is_position_legal()
# -----
def is_position_legal(board: chess.Board, no_promotion: bool =
  True) -> bool:
   0.00
   Checks whether a given position is legal according to chess
      rules.
   Args:
       board (chess.Board): The position to check.
       no_promotion (bool): If True, ignores
          STATUS_PAWNS_ON_BACKRANK
                           and assumes no pawn promotions.
   Returns:
       bool: True if the position is valid, otherwise False.
   # 1. Basic validation (with or without promotion rule)
   if no_promotion:
       if not isinstance(board, CustomBoard):
           board = CustomBoard(board.fen())
       if not board.is_valid_no_promotion():
          return False
   else:
       if not board.is_valid():
          return False
   # 2. Pawn rank check for no_promotion:
        - White pawns cannot be on rank 1 (rank == 0)
        - Black pawns cannot be on rank 8 (rank == 7)
        - Promotion ranks (White: 8 / Black: 1) are allowed here
   if no_promotion:
       for sq in board.pieces(chess.PAWN, chess.WHITE):
           if chess.square rank(sq) == 0:
              return False
       for sq in board.pieces(chess.PAWN, chess.BLACK):
```

```
if chess.square_rank(sq) == 7:
            return False
# 3. Pawn file check maximum 5 (or 6 without promotion)
   pawns per file
max_pawn_per_file = 6 if no_promotion else 5
for color in [chess.WHITE, chess.BLACK]:
    pawns_by_file = [0] * 8
    for sq in board.pieces(chess.PAWN, color):
        pawns_by_file[chess.square_file(sq)] += 1
    if any(count >= max_pawn_per_file for count in
       pawns_by_file):
        return False
# 4. Bishop color rule:
     - Always check if no_promotion=True
     - Otherwise only if all 8 pawns of that color are still
   present
for color in [chess.WHITE, chess.BLACK]:
    bishops = list(board.pieces(chess.BISHOP, color))
    num_pawns = len(board.pieces(chess.PAWN, color))
    check_bishop_colors = no_promotion or num_pawns == 8
    if check_bishop_colors and len(bishops) >= 2:
        # True = light square, False = dark square
        colors = [
            (chess.square_file(b) + chess.square_rank(b)) % 2
                == 0
            for b in bishops
        ]
        # All bishops on the same color invalid
        if all(colors) or not any(colors):
            return False
# 5. All checks passed
return True
```

Listing 2: Validation of chess positions using python-chess

Bishop Color Rule. The default python-chess legality checks do not enforce the constraint that two bishops of the same color on a side are only possible through pawn promotion. To address this, our implementation (see Listing 2) adds an explicit bishop color diversity check whenever the no_promotion flag is enabled. This guarantees that each side may have at most one bishop on each color complex, which reflects the structure of a standard chess game without promotions. The heuristic ensures that diagrams with "illegal twin bishops" on the same color are excluded from the dataset.

Pawn File Rule. A further extension of the validation routine concerns pawn distribution across files. Our function explicitly rejects any diagram containing six or more pawns of the same color on a single file. For the case of allowed promotions the limit for

the maximum pawn count per color per file is five. A rigourous proof for these limits is missing in this work ans should be provided in future works.

Pawn Back Rank Rule. In the case of no allowed promotions, the *python-chess* library will return false validations, because it will consider white pawns as illegal in the last rank (and black pawns in the first rank). Therefore we implemented our own legality test for this case.

2.4 Monte Carlo Sampling and Legal Ratio Estimation

For each material class, we estimate the fraction of legal diagrams by Monte Carlo sampling. Random diagrams are generated by placing the pieces of the class uniformly at random on the 64 squares, and each diagram is tested for legality using the python-chess library (see Section 2.2.1).

Let n denote the number of sampled diagrams and l the number among them that pass the legality check. The legal ratio for the material class is then estimated as

$$\hat{r} = \frac{l}{n}.\tag{1}$$

2.4.1 Adaptive Sample Size Determination

To obtain stable estimates without oversampling easy classes, we use an adaptive sampling scheme that increases the sample size only as needed. The procedure implemented in our code is:

- 1. **Initialization.** Start with an initial sample size $n_0 = 1000$ and draw n_0 random diagrams for the class. Record the number l_0 of legal diagrams and compute the initial estimate $\hat{r}_0 = l_0/n_0$.
- 2. Incremental refinement (accumulating samples). If higher precision is required, we do not discard previous samples. Instead, we double the target sample size $n \leftarrow \min(2n, 128,000)$ and draw only the additional diagrams needed to reach n. Let k be the cumulative number of legal diagrams out of the cumulative n samples; update $\hat{r} = k/n$.
- 3. Error estimation per iteration. After each update we compute the binomial standard error

$$SE(\hat{r}) = \sqrt{\frac{\hat{r}(1-\hat{r})}{n}}, \qquad (2)$$

and its relative form $rSE(\hat{r}) = SE(\hat{r})/\hat{r}$ (defined as 1 if $\hat{r} = 0$).

4. Stopping rule. We stop when the relative standard error is below the threshold

$$\frac{\mathrm{SE}(\hat{r})}{\hat{r}} < 0.10, \tag{3}$$

or when the cap $n \leq 128,000$ is reached. Thus, per-class relative uncertainty is at most 10% by construction, and often substantially smaller.

This adaptive, cumulative design concentrates computation on difficult classes (small \hat{r} or high variance), while simple classes terminate early.

2.4.2 Statistical Properties and Error Propagation

Assuming independence between classes, the total number of legal diagrams is estimated as

$$\hat{N}_{\text{legal}} = \sum_{i} d_i \, \hat{r}_i \,, \tag{4}$$

where d_i is the (theoretical) diagram count of class i and \hat{r}_i its estimated legal ratio. For each class i, the binomial standard error is

$$SE(\hat{r}_i) = \sqrt{\frac{\hat{r}_i(1-\hat{r}_i)}{n_i}}.$$

By linear error propagation, the variance of the global estimate is

$$\operatorname{Var}(\hat{N}_{\text{legal}}) = \sum_{i} d_{i}^{2} \operatorname{SE}(\hat{r}_{i})^{2} \quad \Rightarrow \quad \operatorname{SE}(\hat{N}_{\text{legal}}) = \sqrt{\sum_{i} d_{i}^{2} \operatorname{SE}(\hat{r}_{i})^{2}}.$$
 (5)

A convenient global relative uncertainty is then

$$rSE(\hat{N}_{legal}) = \frac{SE(\hat{N}_{legal})}{\hat{N}_{legal}}.$$
 (6)

Observed precision. Because each class is sampled until its *per-class* relative standard error is at most 10%, the final per-class uncertainties lie at or below this threshold (many classes finish below it). Aggregating across all material classes with the variance formula above, we obtained a *global* relative standard error of approximately 4.1% for \hat{N}_{legal} , reflecting averaging across a very large number of independent class estimates.

3 Results

Based on the Monte Carlo sampling approach described above, the total number of *legal* chess diagrams (i.e., piece arrangements satisfying the structural rules of chess but not necessarily reachable by legal play) was estimated.

For each material class m, the total number of diagrams was computed combinatorially, and multiplied by the empirically estimated valid ratio r_m obtained from the Monte Carlo simulations. The overall estimate across all material classes is therefore given by:

$$N_{\text{legal}} = \sum_{m \in \mathcal{M}} r_m \cdot N_m,$$

where:

- \mathcal{M} denotes the set of all material classes (possible distributions of pieces by type and color),
- N_m is the total number of diagrams for material class m,
- r_m is the fraction of those diagrams that are structurally legal according to python-chess.

Using this procedure, the estimated upper bound for the total number of legal diagrams is:

$$N_{\text{legal}} \approx 1.41 \times 10^{41} \pm 4.1^{39} (2.9\%).$$

To compute this bound we did calculate approximately 1000 samples per material class (median) and checked the legality: In total about 441×10^6 sample diagrams. The minimum sample size was 1000 for some classes and the maximum 16000 for others. As stated above we used a sampling method with adaptive sizes per material class. When compared to the total number of all possible diagrams (i.e., all unconstrained piece placements across all material classes), the global ratio of legal to total diagrams was found to be approximately:

$$\frac{N_{\text{legal}}}{N_{\text{all}}} \approx 1.19 \times 10^{-2}.$$

This indicates that fewer than one in every 800 random piece configurations satisfies even the basic structural rules of chess, stated above. The legal ratio r_m is, however, not constant across material classes: positions with a small number of pieces (e.g., king and pawns only) or heavily asymmetrical material distributions exhibit a higher probability of structural legality, while classes with many pieces or numerous pawns show significantly lower ratios due to increased likelihood of overlapping or invalid pawn placements.

These results represent an upper bound, as the legality checks performed do not test for more complex unreachable pawn patterns like side-pawn patterns [7]. Additionally, the diagrams are not checked for reachability.

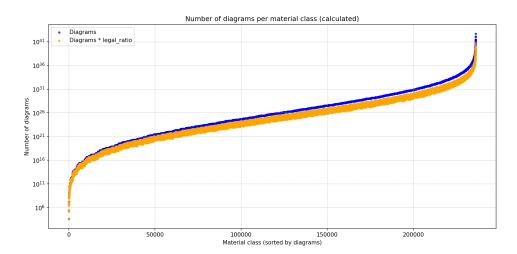


Figure 2: Count of diagrams and legal diagrams over the rank of each material class. The rank of a classes is defined by sorting by the number of all diagrams of the classes.

In figure 2 one may observe, that only very few of material classes have a huge impact on the total number of possible (legal) diagrams. Therefore it would be sufficient to only analyze these classes to come up with a good approximation of the stated upper bound for legal diagrams. The visible noise in the legal diagrams line, is due to the statistical approach.

The figure 3 shows the legal ratio over all classes, with classes sorted in the same way, as in figure 2. One may observe a noisy variation of the legal ratio in dependence of the material class. Only a part of the observed variation is explainable by the statistical

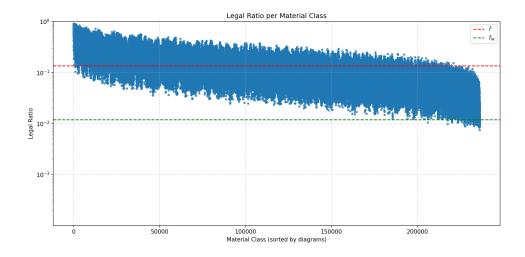


Figure 3: Legal ratio over the rank of material classes.

approach of measuring the legal ratio. There is also a systematic drift, with low ratio for classes with high rank. Meaning the fraction of illegal diagrams is higher for classes with large counts of different diagrams.

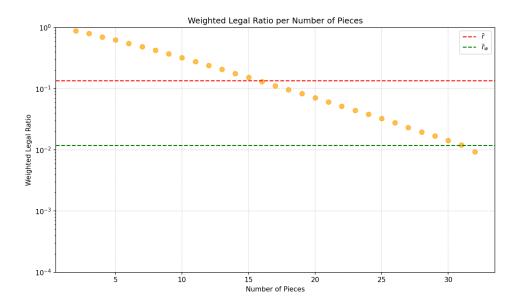


Figure 4: Average Legal ratio over number of chess pieces

Figure 4 shows the average legal ratio over the number of chess pieces. One observes, that the lowest fraction of legal diagrams can be found for 32 pieces. This indicates again, that for solving the problem of finding an upper bound for legal diagrams or positions, analyzing games with high number of pieces is most important.

Finally figure 5 shows the number of total and legal diagrams over the number of chess pieces. Interestingly we find the maximum of legal diagrams for 31 pieces and not 32, like for the total number of diagrams.

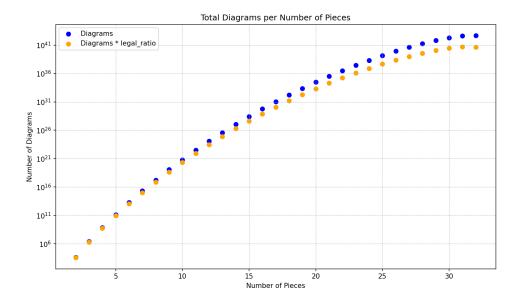


Figure 5: Average (legal) diagrams over number of chess pieces

4 Summary and Conclusion

This study demonstrates that Monte Carlo simulations can provide meaningful statistical estimates for the otherwise intractable problem of enumerating legal chess diagrams (without promotions). By decomposing the search space into well-defined material classes and applying legality checks on sampled diagrams, we approximate the total number of structurally valid configurations to be on the order of 10^{41} .

The data reveal strong variations in the legality ratio across material classes and a systematic decline for positions with a high total number of pieces, especially near the full 32-piece configuration. This indicates that illegalities such as invalid pawn placements and conflicting king states dominate in dense positions.

Our findings also highlight that only a small subset of material classes significantly contributes to the global count primarily those with near-complete sets of pieces. This observation suggests a strategy for future refinements: focus computational effort on high-impact material classes while applying statistical interpolation for the rest.

While this work excludes promotions and reachability constraints, it provides a scalable foundation for estimating the space of *legal* (but not necessarily *reachable*) chess diagrams. Extending the method to include promoted pieces, bishop color and pawn structure constraints represents a promising direction for further research.

5 References

References

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