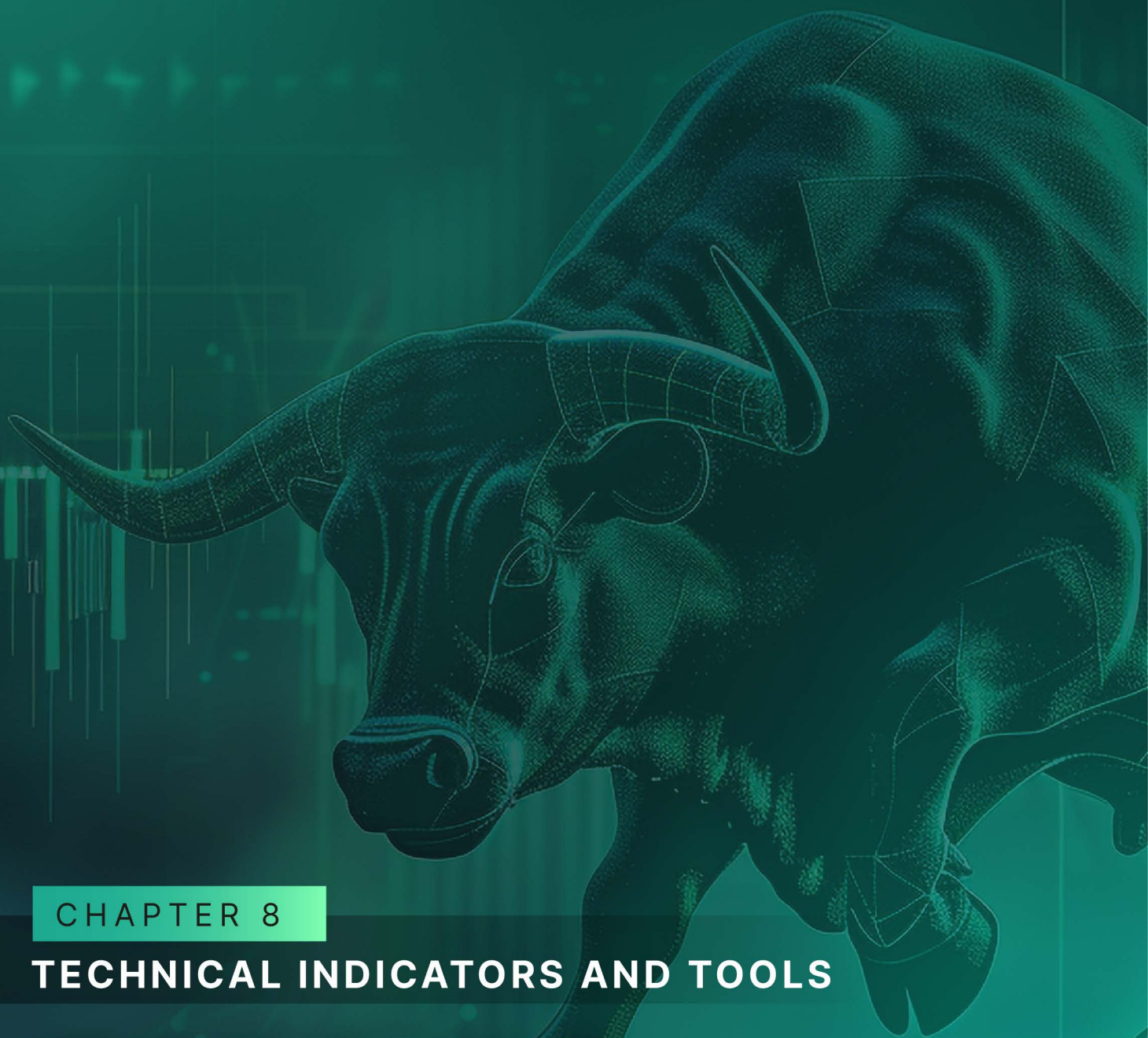


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CHAPTER 8

TECHNICAL INDICATORS AND TOOLS

TRADING STRATEGIES

FROM BEGINNER TO PROFESSIONAL

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Technical Indicators and Tools

Technical indicators are like the instrumentation of a pilot's flight: they do not replace skill and experience, but they provide crucial information to navigate in difficult conditions. The key is not to use all indicators, but to master perfectly a select set.

The Pyramid of Indicators

Level 1: Price Indicators (Foundation)

- Pure Price Action
- Support & Resistance
- Trend Lines

Level 2: Volume Indicators

Volume Profile

On-Balance Volume (OBV)

Volume-Weighted Average Price (VWAP)

Level 3: Momentum Indicators

RSI, MACD, Stochastic

Rate of Change (ROC)

Williams %R

Level 4: Volatility Indicators

- Bollinger Bands
- Average True Range (ATR)
- Keltner Channels

Level 5: Composite Indicators

- Ichimoku Cloud
- Custom Multi-Factor Indicators

Price Indicators: The Base of Everything

Price Action: The Language of Price

Price Action is like reading the body language of the market. Each Japanese candle tells a story about the battle between buyers and sellers.

Master Japanese Candle Patterns:

1. Doji Family:

python

```
def identify_doji_patterns(ohlc_data):
    """
    Identifies doji patterns and their variations
    """
    o, h, l, c = ohlc_data['Open'], ohlc_data['High'], ohlc_data['Low'], ohlc_data['Close']
    body_size = abs(c - o)
    total_range = h - l
    upper_shadow = h - max(o, c)
    lower_shadow = min(o, c) - l

    # Criteria for different types of doji
    is_doji = body_size <= (total_range * 0.1)
    is_dragonfly = is_doji and (lower_shadow >= total_range * 0.6) and (upper_shadow
    <= total_range * 0.1)
    is_gravestone = is_doji and (upper_shadow >= total_range * 0.6) and
    (lower_shadow <= total_range * 0.1)
    is_long_legged = is_doji and (upper_shadow >= total_range * 0.3) and
    (lower_shadow >= total_range * 0.3)

    return {
        'doji': is_doji,
        'dragonfly_doji': is_dragonfly,
        'gravestone_doji': is_gravestone,
        'long_legged_doji': is_long_legged
    }
```

2. **Engulfing Patterns:** The engulfing candles are like a change of guard in the market: the new dominant force literally "devours" the previous one.

python

```
def bullish_engulfing(prev_candle, current_candle):
    """
    Identifies bullish engulfing pattern
    """
    # Previous candle must be bearish
    prev_bearish = prev_candle['Close'] < prev_candle['Open']

    # Current candle must be bullish
    current_bullish = current_candle['Close'] > current_candle['Open']

    # Current candle must completely engulf the previous one
    engulfs_body = (current_candle['Open'] < prev_candle['Close'] and
                    current_candle['Close'] > prev_candle['Open'])

    # Volume must be higher than the average
    volume_confirmation = current_candle['Volume'] > current_candle['Volume_MA_20']

    return prev_bearish and current_bullish and engulfs_body and volume_confirmation
```

python

```
class SupportResistanceDetector:
    def __init__(self, lookback_period=50, min_touches=3, zone_threshold=0.002):
        self.lookback_period = lookback_period
        self.min_touches = min_touches
        self.zone_threshold = zone_threshold

    def find_pivot_points(self, price_data):
        """
        Finds significant pivot points in the price
        """
        highs = price_data['High']
        lows = price_data['Low']

        # High pivots (resistance candidates)
        pivot_highs = []
        for i in range(2, len(highs) - 2):
            if (highs[i] > highs[i-1] and highs[i] > highs[i-2] and
                highs[i] > highs[i+1] and highs[i] > highs[i+2]):
```

```

        pivot_highs.append({'price': highs[i], 'index': i})

    # Low pivots (support candidates)
    pivot_lows = []
    for i in range(2, len(lows) - 2):
        if (lows[i] < lows[i-1] and lows[i] < lows[i-2] and
            lows[i] < lows[i+1] and lows[i] < lows[i+2]):
            pivot_lows.append({'price': lows[i], 'index': i})
    return pivot_highs, pivot_lows

def cluster_levels(self, pivots):
    """
    Groups nearby pivots into support/resistance zones
    """
    if not pivots:
        return []
    clusters = []
    for pivot in pivots:
        price = pivot['price']

        # Look for existing cluster
        assigned = False
        for cluster in clusters:
            if abs(price - cluster['avg_price']) / cluster['avg_price'] <=
self.zone_threshold:
                cluster['prices'].append(price)
                cluster['touches'] += 1
                cluster['avg_price'] = sum(cluster['prices']) / len(cluster['prices'])
                assigned = True
                break

        # Create new cluster if not assigned
        if not assigned:
            clusters.append({
                'prices': [price],
                'touches': 1,
                'avg_price': price,
                'strength': 1
            })

    # Filter clusters with sufficient touches

```

```
significant_levels = [cluster for cluster in clusters if cluster['touches'] >=
self.min_touches]
return sorted(significant_levels, key=lambda x: x['touches'], reverse=True)
```

Volume Indicators: The Fuel of Movement

Volume Profile: The Digital Footprint of the Market

The Volume Profile shows us where traders really put their money, revealing the most important levels of the market.

```
python
class VolumeProfileAnalyzer:
    def __init__(self, price_levels=100):
        self.price_levels = price_levels

    def calculate_volume_profile(self, ohlc_data):
        """
        Calculates the volume profile to identify POC and Value Area
        """
        high_price = ohlc_data['High'].max()
        low_price = ohlc_data['Low'].min()
        price_range = high_price - low_price
        tick_size = price_range / self.price_levels

        # Create price bins
        price_bins = {}
        for i in range(self.price_levels):
            price_level = low_price + (i * tick_size)
            price_bins[round(price_level, 4)] = 0

        # Distribute volume into bins
        for idx, row in ohlc_data.iterrows():
            candle_range = row['High'] - row['Low']
            if candle_range == 0:
                continue

        # Distribute volume proportionally across the candle's range
        for price_level in price_bins.keys():
```

```

        if row['Low'] <= price_level <= row['High']:
            weight = 1 - abs(price_level - ((row['High'] + row['Low']) / 2)) /
(candle_range / 2)
            price_bins[price_level] += row['Volume'] * weight

# Find POC (Point of Control)
poc_price = max(price_bins, key=price_bins.get)
poc_volume = price_bins[poc_price]

# Calculate Value Area (70% of total volume)
total_volume = sum(price_bins.values())
target_volume = total_volume * 0.70

# Expand from POC until reaching 70% of volume
sorted_levels = sorted(price_bins.items(), key=lambda x: x[1], reverse=True)
value_area_volume = poc_volume
value_area_levels = [poc_price]
for price, volume in sorted_levels[1:]:
    if value_area_volume >= target_volume:
        break
    value_area_levels.append(price)
    value_area_volume += volume

vah = max(value_area_levels) # Value Area High
val = min(value_area_levels) # Value Area Low
return {
    'poc': poc_price,
    'vah': vah,
    'val': val,
    'volume_profile': price_bins
}

```

VWAP: The Fair Price of the Day The VWAP (Volume Weighted Average Price) is like the "fair price" of the day, calculated considering the volume in each transaction.

python

```

def calculate_vwap_bands(ohlc_data, num_std=2):
    """
    Calculates VWAP and its standard deviation bands
    """
    # Basic VWAP

```

```
typical_price = (ohlc_data['High'] + ohlc_data['Low'] + ohlc_data['Close']) / 3
cumulative_tpv = (typical_price * ohlc_data['Volume']).cumsum()
cumulative_volume = ohlc_data['Volume'].cumsum()
vwap = cumulative_tpv / cumulative_volume
```

```
# Volume-weighted variance
```

```
squared_diff = (typical_price - vwap) ** 2
cumulative_squared_diff = (squared_diff * ohlc_data['Volume']).cumsum()
vwap_variance = cumulative_squared_diff / cumulative_volume
vwap_std = np.sqrt(vwap_variance)
```

```
# Bands
```

```
upper_band_1 = vwap + (vwap_std * 1)
upper_band_2 = vwap + (vwap_std * num_std)
lower_band_1 = vwap - (vwap_std * 1)
lower_band_2 = vwap - (vwap_std * num_std)
return {
    'vwap': vwap,
    'upper_1': upper_band_1,
    'upper_2': upper_band_2,
    'lower_1': lower_band_1,
    'lower_2': lower_band_2,
    'std_dev': vwap_std
}
```

Advanced RSI: Beyond 70/30

The traditional RSI is just the beginning. Advanced RSI includes divergences, failure swing patterns, and range analysis.

python

```
class AdvancedRSI:
```

```
    def __init__(self, period=14, overbought=70, oversold=30):
        self.period = period
        self.overbought = overbought
        self.oversold = oversold
```

```
def calculate_rsi(self, prices):
    """
    Calculates traditional RSI
    """
    deltas = prices.diff()
    gains = deltas.where(deltas > 0, 0)
    losses = -deltas.where(deltas < 0, 0)
    avg_gains = gains.rolling(window=self.period).mean()
    avg_losses = losses.rolling(window=self.period).mean()
    rs = avg_gains / avg_losses
    rsi = 100 - (100 / (1 + rs))
    return rsi

def detect_divergences(self, prices, rsi_values, lookback=20):
    """
    Detects divergences between price and RSI
    """
    divergences = []
    for i in range(lookback, len(prices) - lookback):
        # Look for peaks and troughs in price
        price_peak = (prices[i] > prices[i-1] and prices[i] > prices[i+1])
        price_trough = (prices[i] < prices[i-1] and prices[i] < prices[i+1])

        if price_peak:
            # Look for previous peak to compare
            for j in range(i - lookback, i - 2):
                if prices[j] > prices[j-1] and prices[j] > prices[j+1]:
                    # Compare price peaks vs RSI
                    if prices[i] > prices[j] and rsi_values[i] < rsi_values[j]:
                        divergences.append({
                            'type': 'bearish',
                            'price_indices': (j, i),
                            'strength': abs(rsi_values[i] - rsi_values[j])
                        })
                    break
```

```

if price_trough:
    # Similar for troughs (bullish divergence)
    for j in range(i - lookback, i - 2):
        if prices[j] < prices[j-1] and prices[j] < prices[j+1]:
            if prices[i] < prices[j] and rsi_values[i] > rsi_values[j]:
                divergences.append({
                    'type': 'bullish',
                    'price_indices': (j, i),
                    'strength': abs(rsi_values[i] - rsi_values[j])
                })
            break
    return divergences

def failure_swings(self, rsi_values):
    """
    Identifies failure swings - powerful reversal signals
    """
    swings = []
    for i in range(4, len(rsi_values)):
        # Bullish failure swing: RSI makes a higher low than previous < 30
        if (rsi_values[i-2] < 30 and
            rsi_values[i] < 30 and
            rsi_values[i] > rsi_values[i-2] and
            rsi_values[i-1] > rsi_values[i] and
            rsi_values[i-1] > rsi_values[i-3]):
            swings.append({
                'type': 'bullish_failure_swing',
                'index': i,
                'strength': 30 - min(rsi_values[i-2], rsi_values[i])
            })
        # Bearish failure swing: RSI makes a lower high than previous > 70
        elif (rsi_values[i-2] > 70 and
              rsi_values[i] > 70 and
              rsi_values[i] < rsi_values[i-2] and
              rsi_values[i-1] < rsi_values[i] and
              rsi_values[i-1] < rsi_values[i-3]):
            swings.append({

```

```
        'type': 'bearish_failure_swing',
        'index': i,
        'strength': min(rsi_values[i-2], rsi_values[i]) - 70
    })
return swings
```

MACD: The Ultimate Momentum Indicator

python

```
class MACDAnalyzer:
```

```
    def __init__(self, fast_period=12, slow_period=26, signal_period=9):
        self.fast_period = fast_period
        self.slow_period = slow_period
        self.signal_period = signal_period
```

```
    def calculate_macd(self, prices):
```

```
        """
```

```
        Calculates MACD line, Signal line, and Histogram
```

```
        """
```

```
        # EMAs
```

```
        ema_fast = prices.ewm(span=self.fast_period).mean()
```

```
        ema_slow = prices.ewm(span=self.slow_period).mean()
```

```
        # MACD Line
```

```
        macd_line = ema_fast - ema_slow
```

```
        # Signal Line
```

```
        signal_line = macd_line.ewm(span=self.signal_period).mean()
```

```
        # Histogram
```

```
        histogram = macd_line - signal_line
```

```
        return {
```

```
            'macd': macd_line,
```

```
            'signal': signal_line,
```

```
            'histogram': histogram
```

```
        }
```

```
    def detect_macd_signals(self, macd_data):
```

```
        """
```

Detects traditional and advanced MACD signals

```
"""
macd = macd_data['macd']
signal = macd_data['signal']
histogram = macd_data['histogram']
signals = []
for i in range(1, len(macd)):
    # Bullish crossover
    if macd[i] > signal[i] and macd[i-1] <= signal[i-1]:
        signals.append({
            'type': 'bullish_crossover',
            'index': i,
            'strength': abs(macd[i] - signal[i])
        })
    # Bearish crossover
    elif macd[i] < signal[i] and macd[i-1] >= signal[i-1]:
        signals.append({
            'type': 'bearish_crossover',
            'index': i,
            'strength': abs(macd[i] - signal[i])
        })
    # Zero line cross
    elif macd[i] > 0 and macd[i-1] <= 0:
        signals.append({
            'type': 'bullish_zero_cross',
            'index': i,
            'strength': macd[i]
        })
    elif macd[i] < 0 and macd[i-1] >= 0:
        signals.append({
            'type': 'bearish_zero_cross',
            'index': i,
            'strength': abs(macd[i])
        })
    # Histogram divergence (early warning)
    if i >= 3:
        if (histogram[i] > histogram[i-1] > histogram[i-2] and
```

```
        histogram[i-2] < 0):
    signals.append({
        'type': 'bullish_histogram_divergence',
        'index': i,
        'strength': abs(histogram[i-2])
    })
return signals
```

Volatility Indicators: Measuring Uncertainty Advanced Bollinger Bands

python

```
class BollingerBandsAdvanced:
    def __init__(self, period=20, std_dev=2.0):
        self.period = period
        self.std_dev = std_dev

    def calculate_bands(self, prices):
        """
        Calculates traditional Bollinger Bands
        """
        sma = prices.rolling(window=self.period).mean()
        std = prices.rolling(window=self.period).std()
        upper_band = sma + (std * self.std_dev)
        lower_band = sma - (std * self.std_dev)
        return {
            'sma': sma,
            'upper': upper_band,
            'lower': lower_band,
            'width': upper_band - lower_band,
            'percent_b': (prices - lower_band) / (upper_band - lower_band)
        }

    def squeeze_detection(self, bb_data, lookback=20):
        """
        Detects Bollinger Band Squeeze - low volatility before breakouts
        """
```

```
width = bb_data['width']
squeeze_signals = []
for i in range(lookback, len(width)):
    current_width = width[i]
    avg_width = width[i-lookback:i].mean()
    min_width = width[i-lookback:i].min()

    # Squeeze: current width is the minimum of the period
    if current_width == min_width and current_width < avg_width * 0.7:
        squeeze_signals.append({
            'index': i,
            'width': current_width,
            'avg_width': avg_width,
            'intensity': avg_width / current_width
        })
return squeeze_signals

def bb_percent_analysis(self, bb_data):
    """
    Analyzes %B to identify extreme conditions
    """
    percent_b = bb_data['percent_b']
    analysis = []
    for i in range(1, len(percent_b)):
        current_b = percent_b[i]
        prev_b = percent_b[i-1]

        # Walking the bands - trending conditions
        if current_b > 0.8 and prev_b > 0.8:
            analysis.append({
                'type': 'walking_upper_band',
                'index': i,
                'strength': current_b
            })
        elif current_b < 0.2 and prev_b < 0.2:
            analysis.append({
                'type': 'walking_lower_band',
```

```
        'index': i,
        'strength': 1 - current_b
    })

    # Reversal signals
    elif current_b > 1.0 and prev_b <= 1.0:
        analysis.append({
            'type': 'breakout_upper',
            'index': i,
            'strength': current_b - 1.0
        })
    elif current_b < 0.0 and prev_b >= 0.0:
        analysis.append({
            'type': 'breakdown_lower',
            'index': i,
            'strength': abs(current_b)
        })
    return analysis
```

Advanced Trading Tools Market Microstructure Analysis

python

```
class MarketMicrostructure:
    def __init__(self):
        pass

    def calculate_bid_ask_spread(self, level1_data):
        """
        Analyzes bid-ask spreads to assess liquidity
        """
        spreads = level1_data['ask'] - level1_data['bid']
        relative_spreads = spreads / level1_data['mid_price']
        return {
            'absolute_spread': spreads,
            'relative_spread': relative_spreads,
```

```

        'avg_spread': spreads.mean(),
        'spread_volatility': spreads.std()
    }

def order_flow_imbalance(self, level2_data):
    """
    Calculates order flow imbalance
    """
    bid_volume = level2_data['bid_volume'].sum()
    ask_volume = level2_data['ask_volume'].sum()
    total_volume = bid_volume + ask_volume
    imbalance = (bid_volume - ask_volume) / total_volume if total_volume > 0 else 0
    return {
        'imbalance': imbalance,
        'bid_volume': bid_volume,
        'ask_volume': ask_volume,
        'interpretation': 'bullish' if imbalance > 0.1 else 'bearish' if imbalance < -0.1 else
'neutral'
    }

def tick_by_tick_analysis(self, tick_data):
    """
    Analyzes tick-by-tick data to detect institutional activity
    """
    # Classify ticks as buyer/seller initiated
    tick_data['trade_type'] = 'unknown'
    for i in range(len(tick_data)):
        price = tick_data.iloc[i]['price']
        bid = tick_data.iloc[i]['bid']
        ask = tick_data.iloc[i]['ask']
        if price >= ask:
            tick_data.iloc[i, tick_data.columns.get_loc('trade_type')] = 'buyer_initiated'
        elif price <= bid:
            tick_data.iloc[i, tick_data.columns.get_loc('trade_type')] = 'seller_initiated'
        else:
            tick_data.iloc[i, tick_data.columns.get_loc('trade_type')] = 'inside_spread'

```

```
# Aggregate analysis
buyer_volume = tick_data[tick_data["trade_type"] ==
'buyer_initiated']['volume'].sum()
seller_volume = tick_data[tick_data["trade_type"] ==
'seller_initiated']['volume'].sum()
net_flow = buyer_volume - seller_volume
return {
    'buyer_volume': buyer_volume,
    'seller_volume': seller_volume,
    'net_flow': net_flow,
    'flow_ratio': buyer_volume / seller_volume if seller_volume > 0 else float('inf')
}
```

Sentiment Analysis Tools

python

```
class MarketSentimentAnalyzer:
```

```
    def __init__(self):
        self.vix_levels = {
            'low': 15,
            'normal': 25,
            'high': 35,
            'extreme': 50
        }
```

```
    def vix_sentiment(self, vix_value):
        """
        Interprets VIX levels for sentiment
        """
        if vix_value < self.vix_levels['low']:
            return {
                'sentiment': 'extremely_bullish',
                'signal': 'caution_complacency',
                'contrarian_view': 'potential_top'
            }
        elif vix_value < self.vix_levels['normal']:
```

```
    return {
        'sentiment': 'bullish',
        'signal': 'low_fear',
        'contrarian_view': 'neutral'
    }
elif vix_value < self.vix_levels['high']:
    return {
        'sentiment': 'neutral_bearish',
        'signal': 'elevated_concern',
        'contrarian_view': 'neutral'
    }
elif vix_value < self.vix_levels['extreme']:
    return {
        'sentiment': 'fearful',
        'signal': 'high_fear',
        'contrarian_view': 'buying_opportunity'
    }
else:
    return {
        'sentiment': 'panic',
        'signal': 'extreme_fear',
        'contrarian_view': 'strong_buy_contrarian'
    }

def put_call_ratio(self, options_data):
    """
    Calculates and interprets Put/Call ratio
    """
    put_volume = options_data['put_volume'].sum()
    call_volume = options_data['call_volume'].sum()
    pc_ratio = put_volume / call_volume if call_volume > 0 else float('inf')

    # Contrarian interpretation
    if pc_ratio > 1.2:
        sentiment = 'extremely_bearish'
        contrarian_signal = 'bullish_contrarian'
    elif pc_ratio > 0.9:
```

```
sentiment = 'bearish'  
contrarian_signal = 'neutral_bullish'  
elif pc_ratio > 0.7:  
    sentiment = 'neutral'  
    contrarian_signal = 'neutral'  
else:  
    sentiment = 'bullish'  
    contrarian_signal = 'bearish_contrarian'  
return {  
    'pc_ratio': pc_ratio,  
    'sentiment': sentiment,  
    'contrarian_signal': contrarian_signal  
}
```



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